# Lecture Notes in Control and Information Sciences 235 

## Ben M. Chen

## $H_{\infty}$ Control

## and Its Applications



Springer

# Lecture Notes <br> in Control and Information Sciences 

Editor: M. Thoma

## Ben M. Chen

## $H_{\infty}$ Control and Its Applications

## Series Advisory Board

A. Bensoussan • M.J. Grimble - P. Kokotovic • H. Kwakernaak J.L. Massey • Y.Z. Tsypkin

## Editors

Dr Ben M. Chen<br>Department of Electrical Engineering, National University of Singapore, Singapore 119260, Singapore

ISBN 978-1-85233-026-2

British Library Cataloguing in Publication Data
Chen, Ben M., 1963-
$\mathrm{H}_{\infty}$ control and its applications. - (Lecture notes in control
and information sciences ; 235)
1.Control theory 2.Mathematical optimization
1.Title
629.8'312

ISBN 978-1-85233-026-2 ISBN 978-1-84628-529-5 (eBook)
DOI 10.1007/978-1-84628-529-5
Library of Congress Cataloging-in-Publication Data
A catalog record for this book is available from the Library of congress
Apart from any fair dealing for the purposes of research or private study, or criticism or review, as permitted under the Copyright, Designs and Patents Act 1988, this publication may only be reproduced, stored or transmitted, in any form or by any means, with the prior permission in writing of the publishers, or in the case of reprographic reproduction in accordance with the terms of licences issued by the Copyright Licensing Agency. Enquiries concerning reproduction outside those terms should be sent to the publishers.
© Springer-Verlag Berlin Heidelberg 1998
Originally published by Springer-Verlag Berlin Heidelberg New York in 1998
The use of registered names, trademarks, etc. in this publication does not imply, even in the absence of a specific statement, that such names are exempt from the relevant laws and regulations and therefore free for general use.

The publisher makes no representation, express or implied, with regard to the accuracy of the information contained in this book and cannot accept any legal responsibility or liability for any errors or omissions that may be made.

This work is dedicated to

My Grand Uncle, Very Reverend Paul Chan,
My Parents, and
Feng, Andy, Jamie and Wen

## Preface

The formulation of the optimization theory has certainly become one of the mile stones of modern control theory. In a typical analytical design of control systems, the given specifications are first transformed into a performance criterion, and then control laws which would minimize the performance criterion are sought. Two important and well-known criteria are the $H_{2}$ norm and the $H_{\infty}$ norm of a transfer matrix from an exogenous disturbance to a pertinent controlled output of a given linear time invariant plant. This book aims to study the $H_{\infty}$ control wherein the control design problem is modeled as a problem of minimizing the $H_{\infty}$ norm of a certain closed-loop transfer matrix under appropriate feedback control laws. Our aim is to examine both the theoretical and practical aspects of $H_{\infty}$ control from the angle of the structural properties of linear systems. Our objectives are to provide constructive algorithms for finding solutions to general singular $H_{\infty}$ control problems, and to general $H_{\infty}$ almost disturbance decoupling problems, as well as to apply these techniques to solve some real life problems. Two practical problems are presented in the book. The first one is about a piezoelectric bimorph actuator system, which has a potential application in forming a dual actuator system for the hard disk drives of the next generation. The second problem is about a gyro-stabilized mirror targeting system, which has some crucial military applications.

The intended audience of this manuscript includes practicing control engineers and researchers in areas related to control engineering. An appropriate background for this monograph would be some first year graduate courses in linear systems and multivariable control. A little bit of knowledge of the geometrical theory of linear systems would certainly be helpful.

I have been fortunate to have the benefit of the cooperation of many coworkers. Foremost, I am indebted to Zongli Lin of the University of Virginia, formerly a fellow classmate at Xiamen University and Washington State University. Many parts of this monograph were born as the result of our continuing collaboration and our numerous discussions over the past few years. In general,

I would like to thank Professors Ali Saberi of Washington State University, Yacov Shamash of the State University of New York at Stony Brook, Chang C. Hang and Tong H. Lee of the National University of Singapore, Uy-Loi Ly of the University of Washington, Yaling Chen of Xiamen University, Anton Stoorvogel of the Eindhoven University of Technology, Steve Weller of the University of Newcastle, and Dr. Siri Weerasooriya of Seagate, Silicon Valley, for their various contributions to this book. Also, I am thankful to Professors Pedda Sannuti of Rutgers University, Dazhong Zheng of Tsinghua University, Yufan Zheng of East-China Normal University, Cishen Zhang of the University of Melbourne, and Shuzhi S. Ge of the National University of Singapore for many beneficial discussions over the past few years, and to Drs. Teck-Seng Low, Tony Huang and Wei Guo of the Data Storage Institute of Singapore for their generous support to my project on the dual actuator systems of hard disk drives.

I am particularly thankful to my current and former graduate students, especially Boon-Choy Siew, Yi Guo, Jun He, Lan Wang and Teck-Beng Goh, for their contributions and for applying and testing parts of the results of this book to real life problems such as gyro-stabilized mirror platform, piezoelectric actuator, and dual actuator systems of hard disk drives. I am also indebted to Andra Leo, my good friend and English teacher at the National University of Singapore, for her kindest help in correcting English errors throughout the original draft of this manuscript.

This work was completed mainly using my 'spare' time, i.e., holidays, weekends, evenings and many sleepless nights. I owe a debt of deepest gratitude to my parents, my wife Feng, and my children Andy, Jamie and Wen, for their sacrifice, patience, understanding and encouragement. Last but certainly not the least, I would like to express my gratitude to my grand uncle, the Very Reverend Paul Chan, and to his Sino-American Amity Fund and Chinese Catholic Information Center, New York. It would not have been possible for me to build my academic career without the spiritual and financial support that I received from them during my course of studies at Gonzaga University and Washington State University. It is natural that I dedicate this work to all of them.

This manuscript was typeset by the author using $\mathrm{IAT}_{\mathrm{E}} \mathrm{X}$. All simulations and numerical computations were carried out in MATLAB and its simulation package, SIMULINK. The real-time implementations were done using $\mathrm{C}_{++}$.

Ben M. Chen

## Contents

1 Introduction ..... 1
1.1 Introduction ..... 1
1.2 Notations and Terminology ..... 3
1.3 Statement of $H_{\infty}$ Optimization Problem ..... 4
1.4 Preview of Chapters ..... 8
2 Linear System Tools ..... 11
2.1 Jordan and Real Jordan Canonical Forms ..... 12
2.2 Brunovsky and Block Diagonal Controllability Forms ..... 14
2.3 Special Coordinate Basis ..... 20
2.4 Proofs of Properties of Special Coordinate Basis ..... 37
3 Existence Conditions of $H_{\infty}$ Suboptimal Controllers ..... 49
3.1 Continuous-time Systems ..... 49
3.2 Discrete-time Systems ..... 54
4 Bilinear Transformations and Discrete Riccati Equations ..... 59
4.1 Structural Mappings of Bilinear Transformations ..... 60
4.1.1 Continuous-time to Discrete-time ..... 61
4.1.2 Discrete-time to Continuous-time ..... 67
4.1.3 Proof of Theorem 4.1.1 ..... 72
4.2 Solutions to Discrete-time Riccati Equations ..... 86
4.2.1 Solution to a General DARE ..... 86
4.2.2 Solution to an $H_{\infty}$-DARE ..... 90
5 Infima in Continuous-time $H_{\infty}$ Optimization ..... 101
5.1 Full Information Feedback Case ..... 102
5.2 Output Feedback Case ..... 111
5.3 Plants with Imaginary Axis Zeros ..... 121
6 Solutions to Continuous-time $H_{\infty}$ Problem ..... 131
6.1 Full State Feedback ..... 132
6.2 Full Order Output Feedback ..... 140
6.3 Reduced Order Output Feedback ..... 146
6.4 Proofs of Main Results ..... 153
7 Continuous-time $H_{\infty}$ Almost Disturbance Decoupling ..... 165
7.1 Introduction ..... 165
7.2 Solvability Conditions ..... 168
7.3 Solutions to Full State Feedback Case ..... 172
7.4 Solutions to Output Feedback Case ..... 179
7.4.1 Full Order Output Feedback ..... 180
7.4.2 Reduced Order Output Feedback ..... 182
7.5 Proofs of Main Results ..... 186
8 Infima in Discrete-time $H_{\infty}$ Optimization ..... 203
8.1 Full Information Feedback Case ..... 203
8.2 Output Feedback Case ..... 215
8.3 Plants with Unit Circle Zeros ..... 220
9 Solutions to Discrete-time $H_{\infty}$ Problem ..... 223
9.1 Full Information and State Feedbacks ..... 224
9.2 Full Order Output Feedback ..... 229
9.3 Reduced Order Output Feedback ..... 234
10 Discrete-time $H_{\infty}$ Almost Disturbance Decoupling ..... 241
10.1 Introduction ..... 241
10.2 Solvability Conditions ..... 243
10.3 Solutions to State and Full Information Feedback Cases ..... 246
10.4 Solutions to Measurement Feedback Case ..... 254
10.4.1 Full Order Output Feedback ..... 255
10.4.2 Reduced Order Output Feedback ..... 258
10.5 Proofs of Main Results ..... 262
11 A Piezoelectric Actuator System Design ..... 283
11.1 Introduction ..... 283
11.2 Linearization of the Nonlinear Hysteretic Dynamics ..... 287
11.3 Formulation of the Problem as an $H_{\infty}$-ADDPMS ..... 289
11.4 Final Controller and Simulation Results ..... 297
12 A Gyro-stabilized Mirror Targeting System Design ..... 309
12.1 Introduction ..... 309
12.2 The Free Gyro-stabilized Mirror System ..... 310
12.3 Controller Design Using an $H_{\infty}$ Approach ..... 316
12.4 Simulation and Implementation Results ..... 321
13 An Open Problem ..... 333
Bibliography ..... 337
Index ..... 349

## Chapter 1

## Introduction

### 1.1. Introduction

The ultimate goal of a control system designer is to build a system that will work in a real environment. Since the real environment may change and operating conditions may vary from time to time, the control system must be able to withstand these variations. Even if the environment does not change, other factors of life are the model uncertainties as well as noises. Any mathematical representation of a system often involves simplifying assumptions. Nonlinearities are either unknown and hence unmodeled, or are modeled and later ignored in order to simplify analysis. High frequency dynamics are often ignored at the design stage as well. In consequence, control systems designed based on simplified models may not work on real plants in real environments. The particular property that a control system must possess for it to operate properly in realistic situations is commonly called robustness. Mathematically, this means that the controller must perform satisfactorily not just for one plant, but for a family of plants. If a controller can be designed such that the whole system to be controlled remains stable when its parameters vary within certain expected limits, the system is said to possess robust stability. In addition, if it can satisfy performance specifications such as steady state tracking, disturbance rejection and speed of response requirements, it is said to possess robust performance. The problem of designing controllers that satisfy both robust stability and performance requirements is called robust control. Optimization theory is one of the cornerstones of modern control theory and was developed in an attempt to solve such a problem. In a typical control system design, the given specifications are at first transformed into a performance index, and then control laws which
would minimize some norm, say $H_{2}$ or $H_{\infty}$ norm of the performance index are sought. This book focusses on the $H_{\infty}$ optimal control theory.

Over the past decades we have witnessed a proliferation of literature on $H_{\infty}$ optimal control since it was first introduced by Zames [114]. The main focus of the work has been and continues to be on the formulation of the problem for robust multivariable control and its solution. Since the original formulation of the $H_{\infty}$ problem in Zames [114], a great deal of work has been done on finding the solution to this problem. Practically all the research results of the early years involved a mixture of time-domain and frequency-domain techniques including the following: i) Interpolation approach (see e.g., Limbeer and Anderson [58]); ii) Frequency domain approach (see e.g., Doyle [37], Francis [42] and Glover [45]); iii) Polynomial approach (see e.g., Kwakernaak [52]); and iv) $J$-spectral factorization approach (see e.g., Kimura [50]). Recently, considerable attention has been focussed on purely time-domain methods based on algebraic Riccati equations (ARE) (see e.g., Chen, Guo and Lin [17], Chen, Saberi and Ly [24], Doyle and Glover [38], Doyle, Glover, Khargonekar and Francis [39], Khargonekar, Petersen and Rotea [49], Petersen [79], Saberi, Chen and Lin [86], Sampei, Mita and Nakamichi [92], Scherer [94-96], Stoorvogel [100], Stoorvogel, Saberi and Chen [102], Tadmor [105], Zhou, Doyle and Glover [115], and Zhou and Khargonekar [116]). Along this line of research, connections are also made between $H_{\infty}$ optimal control and differential games (see e.g., Başar and Bernhard [4], and Papavassilopoulos and Safonov [76]).

Most of the results in the literature are restricted to the so-called regular $H_{\infty}$ control problem (see Definition 1.3.8). Unfortunately, many real life problems do not satisfy these conditions and must be formulated in terms of the regular case by adding some dummy controlled outputs and/or disturbances in order to apply the theory that deals with only the regular problem. The problem we treat in this book is general, i.e., it does not necessarily satisfy the regularity assumptions. The existence conditions for $H_{\infty}$ suboptimal controllers for this type of problem are well studied in Stoorvogel [100] and Scherer [96]. The main focus of this book is, however, very different. We concentrate on 1) the computation of infimum of $H_{\infty}$ optimization problem, which must be known before one can carry out any meaningful designs; 2) solutions to general $H_{\infty}$ optimization problem; 3) solutions to general $H_{\infty}$ disturbance decoupling problem, which itself is a very important subject; and 4) the practical applications of $H_{\infty}$ control.

Most of the results presented in this book are from research carried out by the author and his co-workers over the last six or seven years. The purpose of this book is to discuss various aspects of the subject under a single cover.

### 1.2. Notations and Terminology

Throughout this book we shall adopt the following abbreviations and notations:

```
        \mathbb{R}:= the set of real numbers,
        C}:= the entire complex plane
    \mp@subsup{\mathbb{C}}{}{-}}:== the open left-half complex plane
    \mp@subsup{\mathbb{C}}{}{+}}:= the open right-half complex plane
    \mp@subsup{\mathbb{C}}{}{0}}:=\mathrm{ the imaginary axis in the complex plane,
    \mp@subsup{\mathbb{C}}{}{\odot}}:= the set of complex numbers inside the unit circle
    \mp@subsup{\mathbb{C}}{}{\otimes}}:= the set of complex numbers outside the unit circle
    \mp@subsup{\mathbb{C}}{}{0}}:=\mathrm{ the unit circle in the complex plane,
        I:= an identity matrix,
    I
    X':= the transpose of X,
    X ^ { \mathrm { H } } : = ~ t h e ~ c o m p l e x ~ c o n j u g a t e ~ t r a n s p o s e ~ o f ~ X ,
    det}(X):= the determinant of X
rank}(X):= the rank of X
    X }\mp@subsup{}{}{\dagger}:= the Moore-Penrose (pseudo) inverse of X
    \lambda(X):= the set of eigenvalues of X,
\lambdamax (X):= the maximum eigenvalues of X where }\lambda(X)\subset\mathbb{R}
\sigmamax
    \rho(X):= the spectral radius of X which is equal to max }\mp@subsup{\operatorname{max}}{i}{}|\mp@subsup{\lambda}{i}{}(X)|
        \Sigma* := a linear system characterized by ( }\mp@subsup{A}{*}{},\mp@subsup{B}{*}{},\mp@subsup{C}{*}{},\mp@subsup{D}{*}{})
```



```
Ker (X):= the null space of X,
    Im}(X):= the range space of X
    dim}(\mathcal{X}):= the dimension of a subspace \mathcal{X}
    \mathcal{X}
C}\mp@subsup{}{}{-1}{\mathcal{X}}:={x|Cx\in\mathcal{X}},\mathrm{ where }\mathcal{X}\mathrm{ is a subspace and C is a matrix,
    & := the end of an algorithm or assumption,
        \square := the end of a corollary,
        @ := the end of a definition,
```

国：＝the end of an example，
$\square:=$ the end of a lemma，
＠：＝the end of an observation，
回：＝the end of a property or proposition，
圆：＝the end of a remark，
T $:=$ the end of a theorem，
区：＝the end of the proof of an interim result，
因：＝the end of a proof，
ADDPMS $:=$ almost disturbance decoupling problem with measurement feedback and with internal stability，

ADDPS $:=$ almost disturbance decoupling problem with state feedback and with internal stability，

ARE ：＝algebraic Riccati equation，
CARE：＝continuous－time algebraic Riccati equation，
DARE $:=$ discrete－time algebraic Riccati equation， SCB $:=$ special coordinate basis．

Finally，we denote normrank $\{X(\varsigma)\}$ the rank of $X(\varsigma)$ with entries in the field of rational functions of $\varsigma$ ．

## 1．3．Statement of $H_{\infty}$ Optimization Problem

We consider a generalized system $\Sigma$ with a state－space description，

$$
\Sigma:\left\{\begin{align*}
\delta(x) & =A x+B u+E w  \tag{1.3.1}\\
y & =C_{1} x+D_{1} w \\
h & =C_{2} x+D_{2} u+D_{22} w
\end{align*}\right.
$$

where $\delta(x)=\dot{x}(t)$ if $\Sigma$ is a continuous－time system，or $\delta(x)=x(k+1)$ if $\Sigma$ is a discrete－time system．As usual，$x \in \mathbb{R}^{n}$ is the state，$u \in \mathbb{R}^{m}$ is the control input，$w \in \mathbb{R}^{k}$ is the external disturbance input，$y \in \mathbb{R}^{p}$ is the measurement output，and $h \in \mathbb{R}^{\ell}$ is the controlled output of $\Sigma$ ．They represent $x(t), u(t)$ ， $w(t), y(t)$ and $h(t)$ ，respectively，if $\Sigma$ is of continuous－time，or represent $x(k)$ ， $u(k), w(k), y(k)$ and $h(k)$ ，respectively，if $\Sigma$ is of discrete－time．For the sake of simplicity in future development，throughout this book，we let $\Sigma_{\mathrm{P}}$ be the subsystem characterized by the matrix quadruple $\left(A, B, C_{2}, D_{2}\right)$ and $\Sigma_{\mathrm{Q}}$ be the subsystem characterized by the matrix quadruple $\left(A, E, C_{1}, D_{1}\right)$ ．


Figure 1.3.1: The standard $H_{\infty}$-optimization problem.

The $H_{\infty}$ optimal control problem is to find an internally stabilizing proper measurement feedback control law,

$$
\Sigma_{\mathrm{cmp}}:\left\{\begin{align*}
\delta(v) & =A_{\mathrm{cmp}} v+B_{\mathrm{cmp}} y  \tag{1.3.2}\\
u & =C_{\mathrm{cmp}} v+D_{\mathrm{cmp}} y
\end{align*}\right.
$$

such that the $H_{\infty}$-norm of the overall closed-loop transfer matrix function from $w$ to $h$ is minimized (see also Figure 1.3.1). To be more specific, we will say that the control law $\Sigma_{\mathrm{cmp}}$ of (1.3.2) is internally stabilizing when applied to the system $\Sigma$ of (1.3.1), if the following matrix is asymptotically stable:

$$
A_{\mathrm{cl}}:=\left[\begin{array}{cc}
A+B D_{\mathrm{cmp}} C_{1} & B C_{\mathrm{cmp}}  \tag{1.3.3}\\
B_{\mathrm{cmp}} C_{1} & A_{\mathrm{cmp}}
\end{array}\right]
$$

i.e., all its eigenvalues lie in the open left-half complex plane for a continuoustime system or in the open unit disc for a discrete-time system. It is straightforward to verify that the closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$ is given by the linear fraction map

$$
\begin{equation*}
T_{h w}(\varsigma)=C_{e}\left(\varsigma I-A_{e}\right)^{-1} B_{e}+D_{e} \tag{1.3.4}
\end{equation*}
$$

where $\varsigma=s$, the Laplace transform operator, if $\Sigma$ is a continuous-time system, or $\varsigma=z$, the $z$-transform operator, if $\Sigma$ is a discrete-time one, and

$$
\left.\begin{array}{rl}
A_{e} & :=A_{\mathrm{cl}}=\left[\begin{array}{cc}
A+B D_{\mathrm{cmp}} C_{1} & B C_{\mathrm{cmp}} \\
B_{\mathrm{cmp}} C_{1} & A_{\mathrm{cmp}}
\end{array}\right]  \tag{1.3.5}\\
B_{e}:=\left[\begin{array}{c}
E+B D_{\mathrm{cmp}} D_{1} \\
B_{\mathrm{cmp}} D_{1}
\end{array}\right], \\
C_{e}:=\left[\begin{array}{ll}
C_{2}+D_{2} D_{\mathrm{cmp}} C_{1} & D_{2} C_{\mathrm{cmp}}
\end{array}\right] \\
D_{e} & :=D_{2} D_{\mathrm{cmp}} D_{1}+D_{22}
\end{array}\right\}
$$

It is simple to note that if $\Sigma_{\mathrm{cmp}}$ is a static state feedback law, i.e., $u=F x$, then the closed-loop transfer matrix from $w$ to $h$ is given by

$$
\begin{equation*}
T_{h w}(\varsigma)=\left(C_{2}+D_{2} F\right)(\varsigma I-A-B F)^{-1} E+D_{22} \tag{1.3.6}
\end{equation*}
$$

Similarly, if $\Sigma_{\mathrm{cmp}}$ is given by $u=F_{1} x+F_{2} w$, i.e., a static full information feedback control law, then we have

$$
\begin{equation*}
T_{h w}(\varsigma)=\left(C_{2}+D_{2} F_{1}\right)\left(\varsigma I-A-B F_{1}\right)^{-1}\left(E+B F_{2}\right)+\left(D_{22}+D_{2} F_{2}\right) \tag{1.3.7}
\end{equation*}
$$

The following definitions will be convenient in our future development.
Definition 1.3.1. ( $l_{2}$-norm). The $l_{2}$-norm of a continuous-time signal $y(t)$ is defined by

$$
\begin{equation*}
\|y\|_{2}:=\left(\int_{0}^{\infty} y(t)^{\prime} y(t) d t\right)^{\frac{1}{2}} \tag{1.3.8}
\end{equation*}
$$

Similarly, for a discrete-time signal $y(k)$, we have

$$
\begin{equation*}
\|y\|_{2}:=\left(\sum_{k=0}^{\infty} y(k)^{\prime} y(k)\right)^{\frac{1}{2}} \tag{1.3.9}
\end{equation*}
$$

The square of the $l_{2}$-norm of $y(t)$ or $y(k)$ is commonly termed the total energy in the signal $y(t)$ or $y(k)$. In many areas of engineering, energy or $l_{2}$-norm is used as a measure of the size of a transient signal $y(t)$ or $y(k)$ which decays to zero as time $t$ or shift $k$ progresses towards infinity. By Parseval's theorem, $\|y\|_{2}$ can also be computed in the frequency domain as follows: for the continuous-time case,

$$
\begin{equation*}
\|y\|_{2}=\left(\frac{1}{2 \pi} \int_{-\infty}^{\infty} Y(j \omega)^{\mathrm{H}} Y(j \omega) d \omega\right)^{\frac{1}{2}} \tag{1.3.10}
\end{equation*}
$$

where $Y(j \omega)$ is the Fourier transform of $y(t)$; similarly, for the discrete-time case,

$$
\begin{equation*}
\|y\|_{2}=\left(\frac{1}{2 \pi} \int_{-\pi}^{\pi} Y\left(e^{j \omega}\right)^{\mathrm{H}} Y\left(j e^{\omega}\right) d \omega\right)^{\frac{1}{2}} \tag{1.3.11}
\end{equation*}
$$

where $Y(z)$ is the $z$-transform of $y(k)$.
Definition 1.3.2. ( $H_{\infty}$-norm). The $H_{\infty}$-norm of a stable continuous-time transfer matrix $T_{h w}(s)$ is defined as

$$
\begin{equation*}
\left\|T_{h w}\right\|_{\infty}:=\sup _{w \in[0, \infty)} \sigma_{\max }\left[T_{h w}(j \omega)\right]=\sup _{\|w\|_{2}=1} \frac{\|h\|_{2}}{\|w\|_{2}} \tag{1.3.12}
\end{equation*}
$$

where $w$ and $h$ are respectively the input and output of $T_{h w}$. Similarly, the $H_{\infty}$-norm of a stable discrete-time transfer matrix $T_{h w}(z)$ is defined as

$$
\begin{equation*}
\left\|T_{h w}\right\|_{\infty}:=\sup _{w \in[0,2 \pi]} \sigma_{\max }\left[T_{h w}\left(e^{j \omega}\right)\right]=\sup _{\|w\|_{2}=1} \frac{\|h\|_{2}}{\|w\|_{2}} \tag{1.3.13}
\end{equation*}
$$

where $w$ and $h$ are respectively the input and output of $T_{h w}$.
Definition 1.3.3. ( $\gamma$-Suboptimal Controller). Consider the given system $\Sigma$ of (1.3.1) and the controller $\Sigma_{\mathrm{cmp}}$ of (1.3.2). $\Sigma_{\mathrm{cmp}}$ is said to be an $H_{\infty} \gamma$ suboptimal controller, or in short a $\gamma$-suboptimal controller, for $\Sigma$ if when $\Sigma_{\text {cmp }}$ is applied to $\Sigma$, the resulting closed-loop is internally stable and the $H_{\infty}$-norm of the closed-loop transfer matrix is less than $\gamma$.

Definition 1.3.4. (Infimum $\gamma^{*}$ ). Consider the given system $\Sigma$ of (1.3.1) and the controller $\Sigma_{\text {cmp }}$ of (1.3.2). The infimum of the $H_{\infty}$-norm of the closed-loop transfer matrix $T_{h w}\left(\Sigma \times \Sigma_{\mathrm{cmp}}\right)$ over all stabilizing controllers $\Sigma_{\mathrm{cmp}}$ is denoted by $\gamma^{*}$, namely

$$
\begin{equation*}
\gamma^{*}:=\inf \left\{\left\|T_{h w}\left(\Sigma \times \Sigma_{\mathrm{cmp}}\right)\right\|_{\infty} \mid \Sigma_{\mathrm{cmp}} \text { internally stabilizes } \Sigma\right\} \tag{1.3.14}
\end{equation*}
$$

Obviously, $\gamma^{*} \geq 0$. Occasionally, when it is clear in the context, we may also say that $\gamma^{*}$ is the infimum of the given system $\Sigma$.

Definition 1.3.5. ( $H_{\infty}$ Optimal Controller). Consider the given system $\Sigma$ of (1.3.1) and the controller $\Sigma_{\mathrm{cmp}}$ of (1.3.2). $\Sigma_{\mathrm{cmp}}$ is said to be an $H_{\infty}$ optimal controller for $\Sigma$ if when $\Sigma_{\text {cmp }}$ is applied to $\Sigma$, the resulting closed-loop is internally stable and the $H_{\infty}$-norm of the closed-loop transfer matrix is equal to $\gamma^{*}$.

Definition 1.3.6. (Full Information Feedback Case). Consider the given system $\Sigma$ of (1.3.1). Then, the $H_{\infty}$ optimization problem for $\Sigma$ is called a full information feedback case if

$$
\begin{equation*}
y=\binom{x}{w} \Longrightarrow C_{1}=\binom{I}{0}, D_{1}=\binom{0}{I} \tag{1.3.15}
\end{equation*}
$$

We will also call such a system $\Sigma$ a full information feedback system.
Definition 1.3.7. (Full State Feedback Case). Consider the given system $\Sigma$ of (1.3.1). Then, the $H_{\infty}$ optimization problem for $\Sigma$ is called a full state feedback case if

$$
\begin{equation*}
y=x \quad \Longrightarrow \quad C_{1}=I, D_{1}=0 \tag{1.3.16}
\end{equation*}
$$

We will also call such a system $\Sigma$ a full state feedback system.

Definition 1.3.8. (Regular Case). Consider the given system $\Sigma$ of (1.3.1). Then, the $H_{\infty}$ optimization problem for $\Sigma$ is said to be a regular case or a regular problem provided that:

1. The following conditions are satisfied if $\Sigma$ is a continuous-time system,
(a) $D_{2}$ is of full column rank and $\Sigma_{\mathrm{P}}$ is free of imaginary invariant zeros;
(b) $D_{1}$ is of full row rank and $\Sigma_{Q}$ is free of imaginary invariant zeros.
2. The following conditions are satisfied if $\Sigma$ is a discrete-time system,
(a) $\Sigma_{P}$ is left invertible and is free of unit circle invariant zeros;
(b) $\Sigma_{Q}$ is right invertible and is free of unit circle invariant zeros.

Also, we will call such a system $\Sigma$ a regular system. We note that the characterizations of the regular case for discrete-time systems precisely correspond to those for continuous-time systems under a bilinear mapping. This will be seen clearly later in Chapter 4.

Definition 1.3.9. (Singular Case). Consider the given system $\Sigma$ of (1.3.1). Then, the $H_{\infty}$ optimization problem for $\Sigma$ is said to be a singular case or a singular problem if it is not a regular one. We will occasionally call such a system $\Sigma$ a singular system.

### 1.4. Preview of Chapters

A preview of each chapter is given next. The book can naturally be divided into three parts. The first part covers from Chapters 1 to 4 and contains some preliminary results and background materials. Chapter 2 recalls some linear system tools such as the Jordan and Brunovsky canonical forms and the special coordinate basis. The latter has the distinct feature of explicitly displaying the finite and infinite zero structures of a given system. It plays a dominant role in the development of the whole book. Chapter 3 recalls results on the existence conditions of $H_{\infty}$ suboptimal controllers for both continuous- and discrete-time systems, which are to be used in the proofs of results developed in the second part of the book. Chapter 4 presents two preliminary results, namely, a comprehensive study of the structural mapping of bilinear and inverse bilinear transformations, and solutions to general discrete-time Riccati equations. Both are instrumental in the development of main results in discrete-time $H_{\infty}$ optimization problems.

The second part of the book covers from Chapters 5 to 10 and is also the heart of the book. Chapter 5 deals with the computation of infimum in continuous-time $H_{\infty}$ optimization problem. For a fairly large class of singular problem in which the given system satisfies certain geometric conditions, we present a non-iterative procedure that computes its infimum exactly. For the case when the geometric conditions are not satisfied, we modify our algorithm and give an iterative scheme for approximating this infimum based on an auxiliary reduced order regular system, which generally has a much smaller dynamical order than that of the original system. Chapter 6 deals with finding $H_{\infty} \gamma$-suboptimal controllers for the state feedback case, and the full order and reduced order measurement feedback cases. We provide closed-form solutions to the $H_{\infty}$ suboptimal control problem for the class of singular systems which satisfy the above mentioned geometric conditions. Here by closed-form solutions we mean solutions which are explicitly parameterized in terms of $\gamma$ and are obtained without explicitly requiring a value of $\gamma$. Hence, one can easily tune the parameter $\gamma$ in order to obtain the desired level of disturbance attenuation. This method will be adapted to find $\gamma$-suboptimal control laws for general systems when the geometric conditions are not satisfied. Chapter 7 gives solutions to the general $H_{\infty}$ almost disturbance decoupling problem with either state feedback or measurement feedback and with internal stability for plants whose subsystems have invariant zeros on the imaginary axis of the complex plane. Similarly, Chapters 8 to 10 focus on the discrete-time counterparts of Chapters 5 to 7 , respectively.

The last part of the book consists of some real-life applications of the $H_{\infty}$ theory. Chapter 11 deals with a case study on a piezoelectric actuator control system design using the $H_{\infty}$ almost disturbance decoupling approach. Chapter 12 presents another case study on a gyro-stabilized mirror targeting system design using the $H_{\infty}$ control approach. Both designs are carried out with a clear understanding of the theories and the properties of the given systems. Simulation and/or real implementation results show that these applications turn out to be very satisfactory. Finally, an open problem associated with the computation of the infimum in $H_{\infty}$ optimization is posed in Chapter 13. That concludes the whole book.

## Chapter 2

## Linear System Tools

As WILL BE evident in the coming chapters, the finite and infinite zero structures as well as the invertibility structure of the given system play dominant roles in the computation of the infima and the solutions to both continuoustime and discrete-time $H_{\infty}$ optimization problems. Thus a good non-ambiguous understanding of linear system structure is essential for our study. In our opinion, the best way to display all the structural properties of linear systems is to transform them into a so-called special coordinate basis (SCB) developed by Sannuti and Saberi [93] and Saberi and Sannuti [89]. However, quite often it happens that the original special coordinate basis of Sannuti and Saberi is not fine enough to characterize all the details of the properties of linear systems. In order to see all the fine points of a given system, we would have to further decompose certain subsystems of its SCB using some well-known canonical forms such as the Jordan canonical form and the Brunovsky canonical form. Keeping this in mind, we recall in this chapter the following results: 1) the Jordan and real Jordan canonical forms for a square constant matrix; 2) the Brunovsky canonical form and the block diagonal controllability canonical form for a constant matrix pair; and 3) the special coordinate basis of a linear time invariant system characterized by either a matrix triple or a matrix quadruple. These canonical forms and the special coordinate basis will form a transformer for linear systems. Once a system is touched by this transformer, all its structural properties become clear and transparent. As such, we call it an X-transformer. We should note that the original work of $[93,89]$ dealt only with the continuous time systems. In this chapter, we will unify the special coordinate basis for both continuous-time and discrete-time systems under a single framework. More importantly, we will provide rigorous proofs to all the properties of the special coordinate basis for the first time in the literature.

### 2.1. Jordan and Real Jordan Canonical Forms

We recall in this section the Jordan canonical form and the real Jordan canonical form of a square constant matrix. We first have the following theorem.

Theorem 2.1.1. Consider a constant matrix $A \in \mathbb{R}^{n \times n}$. There exists a nonsingular transformation $T \in \mathbb{C}^{n \times n}$ and an integer $k$ such that

$$
\begin{equation*}
T^{-1} A T=\operatorname{blkdiag}\left\{J_{1}, J_{2}, \cdots, J_{k}\right\} \tag{2.1.1}
\end{equation*}
$$

where $J_{i}, i=1,2, \cdots, k$, are some $n_{i} \times n_{i}$ Jordan blocks, i.e.,

$$
J_{i}=\left[\begin{array}{cccc}
\lambda_{i} & 1 & &  \tag{2.1.2}\\
& \ddots & \ddots & \\
& & \lambda_{i} & 1 \\
& & & \lambda_{i}
\end{array}\right]
$$

Obviously, $\lambda_{i} \in \lambda(A), i=1,2, \cdots, k$, and $\sum_{i=1}^{k} n_{i}=n$.
The result of the above theorem is very well-known. The realization of this Jordan canonical form in MATLAB can be found in Chen [12]. The following theorem is to find a real Jordan canonical form.

Theorem 2.1.2. Consider a constant matrix $A \in \mathbb{R}^{n \times n}$. There exists a nonsingular transformation $P \in \mathbb{R}^{n \times n}$ and an integer $k$ such that

$$
\begin{equation*}
P^{-1} A P=\operatorname{blkdiag}\left\{J_{1}, J_{2}, \cdots, J_{k}\right\} \tag{2.1.3}
\end{equation*}
$$

where each block $J_{i}, i=1,2, \cdots, k$, has the following form: if $\lambda_{i} \in \lambda(A)$ is real,

$$
J_{i}=\left[\begin{array}{cccc}
\lambda_{i} & 1 & &  \tag{2.1.4}\\
& \ddots & \ddots & \\
& & \lambda_{i} & 1 \\
& & & \lambda_{i}
\end{array}\right]
$$

or if $\lambda_{i}=\mu_{i}+j \omega_{i} \in \lambda(A)$ and $\bar{\lambda}_{i}=\mu_{i}-j \omega_{i} \in \lambda(A)$ with $\omega_{i} \neq 0$,

$$
J_{i}=\left[\begin{array}{cccc}
\Lambda_{i} & I_{2} & &  \tag{2.1.5}\\
& \ddots & \ddots & \\
& & \Lambda_{i} & I_{2} \\
& & & \Lambda_{i}
\end{array}\right], \quad \Lambda_{i}=\left[\begin{array}{cc}
\mu_{i} & \omega_{i} \\
-\omega_{i} & \mu_{i}
\end{array}\right]
$$

The above structure of $P^{-1} A P$ is called the real Jordan canonical form. 团

The proof of the above theorem can be found in many texts (see e.g., Wonham [113]). The following is a constructive algorithm for obtaining the transformation $P$ that will transform the given matrix $A$ into the real Jordan canonical form. First, we compute a non-singular transformation $T \in \mathbf{R}^{n \times n}$ such that

$$
\begin{equation*}
T^{-1} A T=\operatorname{blkdiag}\left\{A_{1}, A_{2}, \cdots, A_{\ell}\right\} \tag{2.1.6}
\end{equation*}
$$

where sub-matrices $A_{i} \in \mathbb{R}^{n_{i} \times n_{i}}, i=1,2, \cdots, \ell$, have either a single or one repeated (if $n_{i}>1$ ) eigenvalue $\lambda_{i}$, if $\lambda_{i}$ is real, or two or two repeated (if $\left.n_{i}>2\right)$ eigenvalues $\lambda_{i}$ and $\bar{\lambda}_{i}$, if $\lambda_{i}$ is not real. Also, we have $\lambda_{i} \neq \lambda_{j}$, if $i \neq j$. Note that such a transformation $T$ can easily be obtained using some numerically very stable algorithms such as the real Schur decomposition.

For each $A_{i}$ with its corresponding $\lambda_{i}$ being a real number, we use the result of Theorem 2.1.1 to obtain a non-singular transformation $\tilde{S}_{i}=S_{i} \in \mathbb{R}^{n_{i} \times n_{i}}$ such that $A_{i}$ can be transformed into the Jordan canonical form. For each $A_{i}$ which has eigenvalues $\lambda_{i}=\mu_{i}+j \omega_{i}$ and $\bar{\lambda}_{i}=\mu_{i}-j \omega_{i}$ with $\omega_{i}>0$, we follow the result of Fama and Matthews [40] to define a new ( $2 n_{i}$ ) $\times\left(2 n_{i}\right)$ matrix,

$$
Z_{i}:=\left[\begin{array}{cc}
A_{i}-\mu_{i} I_{n_{i}} & \omega_{i} I_{n_{i}}  \tag{2.1.7}\\
-\omega_{i} I_{n_{i}} & A_{i}-\mu_{i} I_{n_{i}}
\end{array}\right]
$$

It is simple to show that $Z_{i}$ has $n_{i}$ real eigenvalues at 0 and $n_{i}$ purely imaginary eigenvalues at $\pm j 2 \omega_{i}$. Then, we use the real Schur decomposition technique to find a non-singular transformation $S_{i}^{0} \in \mathbb{R}^{\left(2 n_{i}\right) \times\left(2 n_{i}\right)}$ such that

$$
\left(S_{i}^{0}\right)^{-1} Z_{i} S_{i}^{0}=\left[\begin{array}{cc}
Z_{i 0} & 0  \tag{2.1.8}\\
0 & Z_{i x}
\end{array}\right]
$$

where $Z_{i 0}$ has all its eigenvalues at 0 while $Z_{i x}$ has no eigenvalue at 0 . Next, we utilize the result of Theorem 2.1.1 to obtain a non-singular transformation $S_{i}^{1} \in \mathbb{R}^{n_{i} \times n_{i}}$ such that

$$
\begin{equation*}
\left(S_{i}^{1}\right)^{-1} Z_{i 0} S_{i}^{1}=\operatorname{blkdiag}\left\{J_{0}^{1}, J_{0}^{1}, J_{0}^{2}, J_{0}^{2}, \cdots, J_{0}^{\sigma_{i}}, J_{0}^{\sigma_{i}}\right\} \tag{2.1.9}
\end{equation*}
$$

where $J_{0}^{m}, m=1,2, \cdots, \sigma_{i}$, have the form,

$$
J_{0}^{m}=\left[\begin{array}{cc}
0 & I_{n_{i m}-1}  \tag{2.1.10}\\
0 & 0
\end{array}\right]
$$

Let us partition

$$
\begin{align*}
S_{i}:=S_{i}^{0}\left[\begin{array}{cc}
S_{i}^{1} & 0 \\
0 & I_{n_{i}}
\end{array}\right]= & {\left[\begin{array}{lllllll}
S_{i, 1}^{1,1} & \cdots & S_{i, 1}^{1, n_{i 1}} & X_{i, 1}^{1,1} & \cdots & X_{i, 1}^{1, n_{i 1}} & \cdots \cdots \\
S_{i, 1}^{2,1} & \cdots & S_{i, 1}^{2, n_{i 1}} & X_{i, 1}^{2,1} & \cdots & X_{i, 1}^{2, n_{i 1}} & \cdots \cdots
\end{array}\right.} \\
& S_{i, \sigma_{i}}^{1,1} \tag{2.1.11}
\end{align*} \cdots
$$

where $S_{i, m}^{1, k}, S_{i, m}^{2, k}, X_{i, m}^{1, k}$ and $X_{i, m}^{2, k}, m=1,2, \cdots, \sigma_{i}$ and $k=1,2, \cdots, n_{i m}$, are $n_{i} \times 1$ column vectors. In fact, they are all real-valued. Next, define an $n_{i} \times n_{i}$ real-valued matrix,

$$
\tilde{S}_{i}=\left[\begin{array}{lllllllllll}
S_{i, 1}^{1,1} & S_{i, 1}^{2,1} & \cdots & S_{i, 1}^{1, n_{i 1}} & S_{i, 1}^{2, n_{i 1}} & \cdots & S_{i, \sigma_{i}}^{1,1} & S_{i, \sigma_{i}}^{2,1} & \cdots & S_{i, \sigma_{i}}^{1, n_{i}} & S_{i, \sigma_{i}}^{2, n_{i}}
\end{array}\right] .
$$

Finally, let

$$
\begin{equation*}
S=\operatorname{blkdiag}\left\{\tilde{S}_{1}, \cdots, \tilde{S}_{\ell}\right\} \tag{2.1.12}
\end{equation*}
$$

and $P=T S \in \mathbb{R}^{n \times n}$. It is now straightforward to show that $P^{-1} A P$ is in the real Jordan canonical form as described in Theorem 2.1.2. The algorithm has been implemented in Chen [12].

### 2.2. Brunovsky and Block Diagonal Controllability Forms

In this section, we first recall the well-known Brunovsky canonical form for a matrix pair, and then introduce a so-called canonical form for a controllable matrix pair, say $(A, B)$. Both will be the keys in the derivations of some important results later in the book. The derivation of the former is well-known in the literature and the software realization of the Brunovsky canonical form can be found in Chen [12]. We will give an explicit constructing algorithm for the latter to find non-singular transformations, say $T_{s}$ and $T_{i}$, such that $T_{s}^{-1} A T_{s}$ has a special block diagonal form and $T_{s}^{-1} B T_{i}$ has an upper block triangular form. Such special forms of $A$ and $B$ will play an important role in constructing solutions to the general $H_{\infty}$ almost disturbance decoupling problems later in this book. The existence of this block diagonal controllability canonical form was proved by Wonham [113].

We have the following theorems regarding the Brunovsky canonical form and the block diagonal controllability canonical form for a given matrix pair.

Theorem 2.2.1. Consider a constant matrix pair $(A, B)$ with $A \in \mathbb{R}^{n \times n}$ and $B \in \mathbb{R}^{n \times m}$ with $B$ being of full rank. There exist nonsingular state and input transformations $T_{s}$ and $T_{i}$ such that

$$
\tilde{A}:=T_{s}^{-1} A T_{s}=\left[\begin{array}{cccccc}
A_{o} & 0 & 0 & \cdots & 0 & 0  \tag{2.2.1}\\
0 & 0 & I_{k_{1}-1} & \cdots & 0 & 0 \\
\star & \star & \star & \cdots & \star & \star \\
\vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & 0 & \cdots & 0 & I_{k_{m}-1} \\
\star & \star & \star & \cdots & \star & \star
\end{array}\right]
$$

and

$$
\tilde{B}:=T_{s}^{-1} B T_{i}=\left[\begin{array}{ccc}
0 & \cdots & 0  \tag{2.2.2}\\
0 & \cdots & 0 \\
1 & \cdots & 0 \\
\vdots & \ddots & \vdots \\
0 & \cdots & 0 \\
0 & \cdots & 1
\end{array}\right]
$$

where $k_{i}>0, i=1, \cdots, m, A_{o}$ is of dimension $n_{o}:=n-\sum_{i=1}^{m} k_{i}$ and its eigenvalues are the uncontrollable modes of $(A, B)$. Moreover, the set of integers, $\mathcal{C}:=\left\{n_{o}, k_{1}, \cdots, k_{m}\right\}$, is called the controllability index of $(A, B)$.

Proof. It is well-known. The software realization of such a canonical form can be found in Chen [12].

Theorem 2.2.2. Consider a constant matrix pair $(A, B)$ with $A \in \mathbb{R}^{n \times n}$ and $B \in \mathbb{R}^{n \times m}$ and with $(A, B)$ being completely controllable. Then there exist an integer $k \leq m$, a set of $k$ integers $k_{1}, k_{2}, \cdots, k_{k}$, and non-singular transformations $T_{s}$ and $T_{i}$ such that

$$
T_{s}^{-1} A T_{s}=\left[\begin{array}{ccccc}
A_{1} & 0 & 0 & \cdots & 0  \tag{2.2.3}\\
0 & A_{2} & 0 & \cdots & 0 \\
0 & 0 & A_{3} & \cdots & 0 \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
0 & 0 & 0 & \cdots & A_{k}
\end{array}\right]
$$

and

$$
T_{s}^{-1} B T_{i}=\left[\begin{array}{cccccc}
B_{1} & \star & \star & \cdots & \star & \star  \tag{2.2.4}\\
0 & B_{2} & \star & \cdots & \star & \star \\
0 & 0 & B_{3} & \cdots & \star & \star \\
\vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & 0 & \cdots & B_{k} & \star
\end{array}\right]
$$

where $\star$ 's represent some matrices of less interest, and $A_{i}$ and $B_{i}, i=1,2, \cdots, k$, have the following controllability canonical form,

$$
A_{i}=\left[\begin{array}{ccccc}
0 & 1 & 0 & \cdots & 0  \tag{2.2.5}\\
0 & 0 & 1 & \cdots & 0 \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
0 & 0 & 0 & \cdots & 1 \\
-a_{k_{i}}^{i} & -a_{k_{i}-1}^{i} & -a_{k_{i}-2}^{i} & \cdots & -a_{1}^{i}
\end{array}\right], \quad B_{i}=\left[\begin{array}{c}
0 \\
0 \\
\vdots \\
0 \\
1
\end{array}\right]
$$

for some scalars $a_{1}^{i}, a_{2}^{i}, \cdots, a_{k_{i}}^{i}$. Obviously, $\sum_{i=1}^{k} k_{i}=n$. We call the above structure of $A$ and $B$ a block diagonal controllability canonical form.

Proof. The existence of the block diagonal controllability canonical form was shown in [113]. In what follows, we will give an explicit constructing algorithm for realizing realizing such a canonical form. First, we follow Theorem 2.1.2 to find a non-singular transformation $Q \in \mathbb{R}^{n \times n}$ such that matrix $A$ is transformed into a real Jordan canonical form, i.e.,

$$
\begin{equation*}
\tilde{A}=Q^{-1} A Q=\operatorname{blkdiag}\left\{J_{\lambda_{1}}^{1}, \cdots, J_{\lambda_{1}}^{\sigma_{1}}, J_{\lambda_{2}}^{1}, \cdots, J_{\lambda_{2}}^{\sigma_{2}}, \cdots \cdots, J_{\lambda_{\ell}}^{1}, \cdots, J_{\lambda_{l}}^{\sigma_{l}}\right\} \tag{2.2.6}
\end{equation*}
$$

where $\lambda_{i}=\mu_{i}+j \omega_{i} \in \lambda(A)$ with $\omega_{i} \geq 0$, and also $\lambda_{i_{1}} \neq \lambda_{i_{2}}$, if $i_{1} \neq i_{2}$. Moreover, for each $i \in\{1,2, \cdots, \ell\}$ and $s=1,2, \cdots, \sigma_{i}, J_{\lambda_{i}}^{s} \in \mathbb{R}^{n_{i s} \times n_{i s}}$ has the following real Jordan form,

$$
J_{\lambda_{i}}^{s}=\left[\begin{array}{cccc}
\mu_{i} & 1 & &  \tag{2.2.7}\\
& \ddots & \ddots & \\
& & \mu_{i} & 1 \\
& & & \mu_{i}
\end{array}\right]
$$

if $\omega_{i}=0$, or

$$
J_{\lambda_{i}}^{s}=\left[\begin{array}{cccc}
\Lambda_{i} & I_{2} & &  \tag{2.2.8}\\
& \ddots & \ddots & \\
& & \Lambda_{i} & I_{2} \\
& & & \Lambda_{i}
\end{array}\right], \quad \Lambda_{i}=\left[\begin{array}{cc}
\mu_{i} & \omega_{i} \\
-\omega_{i} & \mu_{i}
\end{array}\right]
$$

if $\omega_{i}>0$. For the sake of easy presentation later, we arrange the Jordan blocks in the way that $n_{i 1} \geq n_{i 2} \geq \cdots \geq n_{i \sigma_{i}}$. Next, compute

$$
\tilde{B}=Q^{-1} B=\left[\begin{array}{cccc}
B_{11}^{1} & B_{11}^{2} & \cdots & B_{11}^{m}  \tag{2.2.9}\\
\vdots & \vdots & \ddots & \vdots \\
B_{1 \sigma_{1}}^{1} & B_{1 \sigma_{1}}^{2} & \cdots & B_{1 \sigma_{1}}^{m} \\
B_{21}^{1} & B_{21}^{2} & \cdots & B_{21}^{m} \\
\vdots & \vdots & \ddots & \vdots \\
B_{2 \sigma_{2}}^{1} & B_{2 \sigma_{2}}^{2} & \cdots & B_{2 \sigma_{2}}^{m} \\
\vdots & \vdots & \ddots & \vdots \\
\vdots & \vdots & \ddots & \vdots \\
B_{\ell 1}^{1} & B_{\ell 1}^{2} & \cdots & B_{\ell 1}^{m} \\
\vdots & \vdots & \ddots & \vdots \\
B_{\ell \sigma_{\ell}}^{1} & B_{\ell \sigma_{\ell}}^{2} & \cdots & B_{\ell \sigma_{\ell}}^{m}
\end{array}\right] .
$$

It is straightforward to verify that the controllability of $(A, B)$ implies: there exists a $B_{i 1}^{\nu}$ with $\nu \in\{1,2, \cdots, m\}$ such that $\left(J_{\lambda_{i}}^{1}, B_{i 1}^{\nu}\right)$ is completely controllable, which is equivalent to the last row of $B_{i 1}^{\nu}$ being non-zero if $\lambda_{i}$ is real, or
at least one of the last two rows of $B_{i 1}^{\nu}$ being non-zero if $\lambda_{i}$ is not real. Thus, it is simple to find a vector

$$
T_{1}=\left[\begin{array}{c}
t_{11}  \tag{2.2.10}\\
t_{21} \\
\vdots \\
t_{m 1}
\end{array}\right], \quad t_{11} \neq 0
$$

and partition

$$
\tilde{B}_{1}=\tilde{B} T_{1}=\left[\begin{array}{c}
\tilde{B}_{11}^{1}  \tag{2.2.11}\\
\vdots \\
\tilde{B}_{1 \sigma_{1}}^{1} \\
\tilde{B}_{21}^{1} \\
\vdots \\
\tilde{B}_{2 \sigma_{2}}^{1} \\
\vdots \\
\vdots \\
\tilde{B}_{\ell 1}^{1} \\
\vdots \\
\tilde{B}_{\ell \sigma_{\ell}}^{1}
\end{array}\right]
$$

such that $\left(J_{\lambda_{i}}^{1}, \tilde{B}_{i 1}^{1}\right)$ is completely controllable. Because of the special structure of the real Jordan form and the fact that $n_{i 1} \geq n_{i 2} \geq \cdots \geq n_{i \sigma_{i}}$, the eigenstructures associated with $J_{\lambda_{i}}^{s}$ with $s>1$ are totally uncontrollable by $\tilde{B}_{1}$. Thus, it is straightforward to show that there exist non-singular transformations $T_{s 1}^{i}$, $i=1,2, \cdots, \ell$, such that

$$
\left(T_{s 1}^{i}\right)^{-1}\left[\begin{array}{cccc}
J_{\lambda_{i}}^{1} & & &  \tag{2.2.12}\\
& J_{\lambda_{i}}^{2} & & \\
& & \ddots & \\
& & & J_{\lambda_{i}}^{\sigma_{i}}
\end{array}\right] T_{s 1}^{i}=\left[\begin{array}{llll}
J_{\lambda_{i}}^{1} & & & \\
& J_{\lambda_{i}}^{2} & & \\
& & \ddots & \\
& & & J_{\lambda_{i}}^{\sigma_{i}}
\end{array}\right]
$$

and

$$
\left(T_{s 1}^{i}\right)^{-1}\left[\begin{array}{c}
\tilde{B}_{i 1}^{1}  \tag{2.2.13}\\
\tilde{B}_{i 2}^{1} \\
\vdots \\
\tilde{B}_{i \sigma_{i}}^{1}
\end{array}\right]=\left[\begin{array}{c}
\check{B}_{i 1}^{1} \\
0 \\
\vdots \\
0
\end{array}\right]
$$

with ( $J_{\lambda_{i}}^{1}, \check{B}_{i 1}^{1}$ ) being completely controllable. This can be done by utilizing the special structure of the well-known Brunovsky canonical form (see Theorem 2.2.1). Next, perform a permutation transformation $P_{s 1}$ such that

$$
\begin{align*}
&\left(P_{s 1}\right)^{-1}\left[\begin{array}{cccc}
T_{s 1}^{1} & & & \\
& T_{s 1}^{2} & & \\
& & \ddots & \\
& & & T_{s 1}^{\ell}
\end{array}\right]^{-1} \tilde{A}\left[\begin{array}{llll}
T_{s 1}^{1} & & & \\
& T_{s 1}^{2} & & \\
& & \ddots & \\
& & & T_{s 1}^{\ell}
\end{array}\right] P_{s 1} \\
&=\operatorname{blkdiag}\left\{J_{\lambda_{1}}^{1}, \cdots, J_{\lambda_{\ell}}^{1}, J_{\lambda_{1}}^{2}, \cdots, J_{\lambda_{1}}^{\sigma_{1}}, \cdots \cdots, J_{\lambda_{\ell}}^{2}, \cdots, J_{\lambda_{\ell}}^{\sigma_{\ell}}\right\} \tag{2.2.14}
\end{align*}
$$

and

$$
\begin{align*}
&\left(P_{s 1}\right)^{-1}\left[\begin{array}{cccc}
T_{s 1}^{1} & & & \\
& T_{s 1}^{2} & & \\
& & \ddots & \\
& & & T_{s 1}^{\ell}
\end{array}\right]^{-1}\left[\begin{array}{cccc}
t_{11} & 0 & \cdots & 0 \\
t_{21} & 1 & \cdots & 0 \\
\vdots & 0 & \ddots & 0 \\
t_{m 1} & 0 & \cdots & 1
\end{array}\right] \\
&=\left[\begin{array}{ccccc}
\check{B}_{11}^{1} & \check{B}_{11}^{2} & \cdots & \check{B}_{11}^{m} \\
\vdots & \vdots & \ddots & \vdots \\
\check{B}_{\ell 1}^{1} & \check{B}_{\ell 1}^{2} & \cdots & \check{B}_{\ell 1}^{m} \\
0 & \check{B}_{12}^{2} & \cdots & \check{B}_{12}^{m} \\
\vdots & \vdots & \ddots & \vdots \\
0 & \check{B}_{1 \sigma_{1}}^{2} & \cdots & \check{B}_{1 \sigma_{1}}^{m} \\
\vdots & \vdots & \ddots & \vdots \\
\vdots & \vdots & \ddots & \vdots \\
0 & \check{B}_{\ell 2}^{2} & \cdots & \check{B}_{\ell 2}^{m} \\
\vdots & \vdots & \ddots & \vdots \\
0 & \check{B}_{\ell \sigma_{\ell}}^{2} & \cdots & \check{B}_{\ell \sigma_{\ell}}^{m}
\end{array}\right] . \tag{2.2.15}
\end{align*}
$$

Because $\lambda_{i}, i=1,2, \cdots, \ell$, are distinct, the controllability of $\left(J_{\lambda_{i}}^{1}, \check{B}_{i 1}^{1}\right)$ implies that the pair

$$
\left(\check{A}_{1}, \check{B}_{1}\right):=\left(\left[\begin{array}{llll}
J_{\lambda_{1}}^{1} & & &  \tag{2.2.16}\\
& J_{\lambda_{2}}^{1} & & \\
& & \ddots & \\
& & & J_{\lambda_{l}}^{1}
\end{array}\right],\left[\begin{array}{c}
\check{B}_{11}^{1} \\
\check{B}_{21}^{1} \\
\vdots \\
\check{B}_{i 1}^{1}
\end{array}\right]\right)
$$

is completely controllable. Hence, there exists a non-singular transformation $X_{1} \in \mathbb{R}^{k_{1} \times k_{1}}$, where $k_{1}=\sum_{i=1}^{\ell} n_{i 1}$, such that

$$
X_{1}^{-1} \check{A}_{1} X_{1}=A_{1}=\left[\begin{array}{ccccc}
0 & 1 & 0 & \cdots & 0  \tag{2.2.17}\\
0 & 0 & 1 & \cdots & 0 \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
0 & 0 & 0 & \cdots & 1 \\
-a_{k_{1}}^{1} & -a_{k_{1}-1}^{1} & -a_{k_{1}-2}^{1} & \cdots & -a_{1}^{1}
\end{array}\right]
$$

and

$$
X_{1}^{-1} \check{B}_{1}=B_{1}=\left[\begin{array}{c}
0  \tag{2.2.18}\\
0 \\
\vdots \\
0 \\
1
\end{array}\right]
$$

Next, repeating the above procedure for the following pair

$$
\left(\operatorname{blkdiag}\left\{J_{\lambda_{1}}^{2}, \cdots, J_{\lambda_{1}}^{\sigma_{1}}, \cdots \cdots, J_{\lambda_{\ell}}^{2}, \cdots, J_{\lambda_{\ell}}^{\sigma_{\ell}}\right\},\left[\begin{array}{ccc}
\check{B}_{12}^{2} & \cdots & \check{B}_{12}^{m}  \tag{2.2.19}\\
\vdots & \ddots & \vdots \\
\check{B}_{1 \sigma_{1}}^{2} & \cdots & \check{B}_{1 \sigma_{1}}^{m} \\
\vdots & \ddots & \vdots \\
\vdots & \ddots & \vdots \\
\check{B}_{\ell 2}^{2} & \cdots & \check{B}_{\ell 2}^{m} \\
\vdots & \ddots & \vdots \\
\check{B}_{\ell \sigma_{\ell}}^{2} & \cdots & \check{B}_{\ell \sigma_{\ell}}^{m}
\end{array}\right]\right) \text {, }
$$

one is able to separate $\left(A_{2}, B_{2}\right)$. Keep repeating the same procedure for $k-2$ more steps, where $k=\max \left\{\sigma_{1}, \sigma_{2}, \cdots, \sigma_{\ell}\right\}$, one is able to obtain the block diagonal controllability canonical form as in Theorem 2.2.2. This completes the proof of the theorem. The result has been implemented in Chen [12]. 区

We illustrate the above results in the following example.
Example 2.2.1. Consider a matrix pair $(A, B)$ characterized by

$$
A=\left[\begin{array}{rrrrrrrr}
1 & 1 & 0 & 0 & 0 & 0 & 0 & 0  \tag{2.2.20}\\
0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 \\
0 & 0 & 0 & 0 & -1 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0 & 0 & -1 & 0
\end{array}\right], \quad B=\left[\begin{array}{ll}
1 & 8 \\
2 & 7 \\
3 & 6 \\
4 & 5 \\
5 & 4 \\
6 & 3 \\
7 & 2 \\
8 & 1
\end{array}\right],
$$

where matrix $A$ is already in the form of the real Jordan canonical form with $\lambda_{1}=1, \sigma_{1}=2$ and $\lambda_{2}=j, \sigma_{2}=1$. Following the proof of Theorem 2.2.2, we obtain

$$
\begin{gathered}
T_{s}=\left[\begin{array}{rrrrrrrr}
0.1508 & 0.1508 & 0.3015 & 0.3015 & 0.1508 & 0.1508 & -0.4002 & 1.8189 \\
-0.3015 & 0.3015 & -0.6030 & 0.6030 & -0.3015 & 0.3015 & -1.4188 & 1.4188 \\
0.1508 & 0.4523 & 0.3015 & 0.9045 & 0.1508 & 0.4523 & 0.3274 & -1.1641 \\
-0.6030 & 0.6030 & -1.2061 & 1.2061 & -0.6030 & 0.6030 & 0.8367 & -0.8367 \\
-0.1508 & 3.4674 & -4.5227 & 0 & 0.4523 & 0.7538 & 0 & 0 \\
-1.9598 & 2.7136 & 0.9045 & -1.2061 & -1.3568 & 0.9045 & 0 & 0 \\
1.2061 & -1.3568 & 0.3015 & -0.3015 & -0.9045 & 1.0553 & 0 & 0 \\
-1.0553 & 3.3166 & -4.5227 & 4.5227 & -3.4674 & 1.2061 & 0 & 0
\end{array}\right], \\
T_{i}=\left[\begin{array}{rrr}
0.1508 & 0 \\
0 & 0.4083
\end{array}\right],
\end{gathered}
$$

and

$$
T_{s}^{-1} A T_{s}=\left[\begin{array}{rrrrrrrr}
0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\
-1 & 2 & -3 & 4 & -3 & 2 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0 & 0 & -1 & 2
\end{array}\right], \quad T_{s}^{-1} B T_{i}=\left[\begin{array}{rr}
0 & -1.4368 \\
0 & -0.2982 \\
0 & 0.5207 \\
0 & 1.3969 \\
0 & 2.8085 \\
1 & 4.6900 \\
0 & 0 \\
0 & 1
\end{array}\right] .
$$

This verifies the results of Theorem 2.2.2.

### 2.3. Special Coordinate Basis

Let us consider a linear time-invariant (LTI) system $\Sigma_{*}$, which could be of either continuous-time or discrete-time, characterized by a matrix quadruple $\left(A_{*}, B_{*}, C_{*}, D_{*}\right)$ or in the state space form,

$$
\left\{\begin{align*}
\delta(x) & =A_{*} x+B_{*} u  \tag{2.3.1}\\
y & =C_{*} x+D_{*} u
\end{align*}\right.
$$

where $\delta(x)=\dot{x}(t)$, if $\Sigma_{*}$ is a continuous-time system, or $\delta(x)=x(k+1)$, if $\Sigma_{*}$ is a discrete-time system. Similarly, $x \in \mathbb{R}^{n}, u \in \mathbb{R}^{m}$ and $y \in \mathbb{R}^{p}$ are the state, the input and the output of $\Sigma_{*}$. They represent $x(t), u(t)$ and $y(t)$, respectively, if the given system is of continuous-time, or represent $x(k), u(k)$ and $y(k)$, respectively, if $\Sigma_{*}$ is of discrete-time. Without loss of any generality, we assume that both $\left[\begin{array}{ll}B_{*}^{\prime} & D_{*}^{\prime}\end{array}\right]$ and $\left[\begin{array}{ll}C_{*} & D_{*}\end{array}\right]$ are of full rank. The transfer function of $\Sigma_{*}$ is then given by

$$
\begin{equation*}
H_{*}(\varsigma)=C_{*}\left(\varsigma I-A_{*}\right)^{-1} B_{*}+D_{*} \tag{2.3.2}
\end{equation*}
$$

where $\varsigma=s$, the Laplace transform operator, if $\Sigma_{*}$ is of continuous-time, or $\varsigma=z$, the $z$-transform operator, if $\Sigma_{*}$ is of discrete-time. It is simple to verify that there exist non-singular transformations $U$ and $V$ such that

$$
U D_{*} V=\left[\begin{array}{cc}
I_{m_{0}} & 0  \tag{2.3.3}\\
0 & 0
\end{array}\right]
$$

where $m_{0}$ is the rank of matrix $D_{*}$. In fact, $U$ can be chosen as an orthogonal matrix. This fact will be used later in the computation of $\gamma^{*}$ throughout this book. Hence hereafter, without loss of generality, it is assumed that the matrix $D_{*}$ has the form given on the right hand side of (2.3.3). One can now rewrite system $\Sigma_{*}$ of (2.3.1) as,

$$
\left\{\begin{array}{l}
\delta(x)=A_{*} \quad x+\left[\begin{array}{ll}
B_{*, 0} & B_{*, 1}
\end{array}\right]\binom{u_{0}}{u_{1}}  \tag{2.3.4}\\
\binom{y_{0}}{y_{1}}=\left[\begin{array}{l}
C_{*, 0} \\
C_{*, 1}
\end{array}\right] x+\left[\begin{array}{cc}
I_{m_{0}} & 0 \\
0 & 0
\end{array}\right]\binom{u_{0}}{u_{1}}
\end{array}\right.
$$

where the matrices $B_{*, 0}, B_{*, 1}, C_{*, 0}$ and $C_{*, 1}$ have appropriate dimensions. We have the following theorem.

Theorem 2.3.1 (SCB). Given the linear system $\Sigma_{*}$ of (2.3.1), there exist

1. Coordinate free non-negative integers $n_{a}^{-}, n_{a}^{0}, n_{a}^{+}, n_{b}, n_{c}, n_{d}, m_{d} \leq m-m_{0}$ and $q_{i}, i=1, \cdots, m_{d}$, and
2. Non-singular state, output and input transformations $\Gamma_{s}, \Gamma_{o}$ and $\Gamma_{i}$ which take the given $\Sigma_{*}$ into a special coordinate basis that displays explicitly both the finite and infinite zero structures of $\Sigma_{*}$.

The special coordinate basis is described by the following set of equations:

$$
\begin{gather*}
x=\Gamma_{s} \tilde{x}, \quad y=\Gamma_{o} \tilde{y}, \quad u=\Gamma_{i} \tilde{u}  \tag{2.3.5}\\
\tilde{x}=\left(\begin{array}{c}
x_{a} \\
x_{b} \\
x_{c} \\
x_{d}
\end{array}\right), \quad x_{a}=\left(\begin{array}{c}
x_{a}^{-} \\
x_{a}^{0} \\
x_{a}^{+}
\end{array}\right), \quad x_{d}=\left(\begin{array}{c}
x_{1} \\
x_{2} \\
\vdots \\
x_{m_{d}}
\end{array}\right)  \tag{2.3.6}\\
\tilde{y}=\left(\begin{array}{c}
y_{0} \\
y_{d} \\
y_{b}
\end{array}\right), y_{d}=\left(\begin{array}{c}
y_{1} \\
y_{2} \\
\vdots \\
y_{m_{d}}
\end{array}\right), \tilde{u}=\left(\begin{array}{c}
u_{0} \\
u_{d} \\
u_{c}
\end{array}\right), u_{d}=\left(\begin{array}{c}
u_{1} \\
u_{2} \\
\vdots \\
u_{m_{d}}
\end{array}\right) \tag{2.3.7}
\end{gather*}
$$

and

$$
\begin{equation*}
\delta\left(x_{a}^{-}\right)=A_{a a}^{-} x_{a}^{-}+B_{0 a}^{-} y_{0}+L_{a d}^{-} y_{d}+L_{a b}^{-} y_{b} \tag{2.3.8}
\end{equation*}
$$

$$
\begin{gather*}
\delta\left(x_{a}^{0}\right)=A_{a a}^{0} x_{a}^{0}+B_{0 a}^{0} y_{0}+L_{a d}^{0} y_{d}+L_{a b}^{0} y_{b},  \tag{2.3.9}\\
\delta\left(x_{a}^{+}\right)=A_{a a}^{+} x_{a}^{+}+B_{0 a}^{+} y_{0}+L_{a d}^{+} y_{d}+L_{a b}^{+} y_{b},  \tag{2.3.10}\\
\delta\left(x_{b}\right)=A_{b b} x_{b}+B_{0 b} y_{0}+L_{b d} y_{d}, \quad y_{b}=C_{b} x_{b},  \tag{2.3.11}\\
\delta\left(x_{c}\right)=A_{c c} x_{c}+B_{0 c} y_{0}+L_{c b} y_{b}+L_{c d} y_{d}+B_{c}\left[E_{c a}^{-} x_{a}^{-}+E_{c a}^{0}+E_{c a}^{+} x_{a}^{+}\right]+B_{c} u_{c},  \tag{2.3.13}\\
y_{0}=C_{0 c} x_{c}+C_{0 a}^{-} x_{a}^{-}+C_{0 a}^{+} x_{a}^{0}+C_{0 a}^{+} x_{a}^{+}+C_{0 d} x_{d}+C_{0 b} x_{b}+u_{0}, \tag{2.3.12}
\end{gather*}
$$

and for each $i=1, \cdots, m_{d}$,

$$
\begin{equation*}
\delta\left(x_{i}\right)=A_{q_{i}} x_{i}+L_{i 0} y_{0}+L_{i d} y_{d}+B_{q_{i}}\left[u_{i}+E_{i a} x_{a}+E_{i b} x_{b}+E_{i c} x_{c}+\sum_{j=1}^{m_{d}} E_{i j} x_{j}\right] \tag{2.3.14}
\end{equation*}
$$

$$
\begin{equation*}
y_{i}=C_{q_{i}} x_{i}, \quad y_{d}=C_{d} x_{d} \tag{2.3.15}
\end{equation*}
$$

Here the states $x_{a}^{-}, x_{a}^{0}, x_{a}^{+}, x_{b}, x_{c}$ and $x_{d}$ are respectively of dimensions $n_{a}^{-}, n_{a}^{0}$, $n_{a}^{+}, n_{b}, n_{c}$ and $n_{d}=\sum_{i=1}^{m_{d}} q_{i}$, while $x_{i}$ is of dimension $q_{i}$ for each $i=1, \cdots, m_{d}$. The control vectors $u_{0}, u_{d}$ and $u_{c}$ are respectively of dimensions $m_{0}, m_{d}$ and $m_{c}=m-m_{0}-m_{d}$ while the output vectors $y_{0}, y_{d}$ and $y_{b}$ are respectively of dimensions $p_{0}=m_{0}, p_{d}=m_{d}$ and $p_{b}=p-p_{0}-p_{d}$. The matrices $A_{q_{i}}, B_{q_{i}}$ and $C_{q_{i}}$ have the following form:

$$
A_{q_{i}}=\left[\begin{array}{cc}
0 & I_{q_{i}-1}  \tag{2.3.16}\\
0 & 0
\end{array}\right], \quad B_{q_{i}}=\left[\begin{array}{l}
0 \\
1
\end{array}\right], \quad C_{q_{i}}=[1,0, \cdots, 0] .
$$

Assuming that $x_{i}, i=1,2, \cdots, m_{d}$, are arranged such that $q_{i} \leq q_{i+1}$, the matrix $L_{i d}$ has the particular form

$$
L_{i d}=\left[\begin{array}{lllllll}
L_{i 1} & L_{i 2} & \cdots & L_{i i-1} & 0 & \cdots & 0 \tag{2.3.17}
\end{array}\right]
$$

Also, the last row of each $L_{i d}$ is identically zero. Moreover,

1. If $\Sigma_{*}$ is a continuous-time system, then

$$
\begin{equation*}
\lambda\left(A_{a a}^{-}\right) \subset \mathbb{C}^{-}, \quad \lambda\left(A_{a a}^{0}\right) \subset \mathbb{C}^{0}, \quad \lambda\left(A_{a a}^{+}\right) \subset \mathbb{C}^{+} \tag{2.3.18}
\end{equation*}
$$

2. If $\Sigma_{*}$ is a discrete-time system, then

$$
\begin{equation*}
\lambda\left(A_{a a}^{-}\right) \subset \mathbb{C}^{\odot}, \quad \lambda\left(A_{a a}^{0}\right) \subset \mathbb{C}^{\circ}, \quad \lambda\left(A_{a a}^{+}\right) \subset \mathbb{C}^{\otimes} . \tag{2.3.19}
\end{equation*}
$$

Also, the pair $\left(A_{c c}, B_{c}\right)$ is controllable and the pair $\left(A_{b b}, C_{b}\right)$ is observable. T

Proof. For strictly proper systems, using a modified structural algorithm of Silverman [98], an explicit procedure of constructing the above special coordinate basis is given in [93]. The required modifications for non-strictly proper systems are given in [89].

Here in Theorem 2.3.1 by another change of basis, the variable $x_{a}$ is further decomposed into $x_{a}^{-}, x_{a}^{0}$ and $x_{a}^{+}$. For continuous-time systems, one can use the real Schur algorithm to obtain such a decomposition. For discrete-time systems, the algorithm of Chen [11] can be used.

The software toolboxes that realize the continuous-time SCB can be found in LAS by Chen [9] or in MATLAB by Lin [60]. The realization of this unified SCB can be found in Chen [12]. A numerical example will be given at the end of this section to illustrate the procedure of constructing the SCB and all its associated properties.

We can rewrite the special coordinate basis of the quadruple $\left(A_{*}, B_{*}, C_{*}, D_{*}\right)$ given by Theorem 2.3.1 in a more compact form,

$$
\begin{align*}
& \tilde{A}_{*}=\Gamma_{s}^{-1}\left(A_{*}-B_{*, 0} C_{*, 0}\right) \Gamma_{s} \\
& =\left[\begin{array}{cccccc}
A_{a a}^{-} & 0 & 0 & L_{a b}^{-} C_{b} & 0 & L_{a d}^{-} C_{d} \\
0 & A_{a a}^{0} & 0 & L_{a b}^{0} C_{b} & 0 & L_{a d}^{0} C_{d} \\
0 & 0 & A_{a a}^{+} & L_{a b}^{+} C_{b} & 0 & L_{a d}^{+} C_{d} \\
0 & 0 & 0 & A_{b b} & 0 & L_{b d} C_{d} \\
B_{c} E_{c a}^{-} & B_{c} E_{c a}^{0} & B_{c} E_{c a}^{+} & L_{c b} C_{b} & A_{c c} & L_{c d} C_{d} \\
B_{d} E_{d a}^{-} & B_{d} E_{d a}^{0} & B_{d} E_{d a}^{+} & B_{d} E_{d b} & B_{d} E_{d c} & A_{d d}
\end{array}\right],  \tag{2.3.20}\\
& \tilde{B}_{*}=\Gamma_{s}^{-1}\left[\begin{array}{ll}
B_{*, 0} & B_{*, 1}
\end{array}\right] \Gamma_{i}=\left[\begin{array}{ccc}
B_{0 a}^{-} & 0 & 0 \\
B_{0 a}^{0} & 0 & 0 \\
B_{0 a}^{+} & 0 & 0 \\
B_{0 b} & 0 & 0 \\
B_{0 c} & 0 & B_{c} \\
B_{0 d} & B_{d} & 0
\end{array}\right],  \tag{2.3.21}\\
& \tilde{C}_{*}=\Gamma_{o}^{-1}\left[\begin{array}{l}
C_{*, 0} \\
C_{*, 1}
\end{array}\right] \Gamma_{s}=\left[\begin{array}{cccccc}
C_{0 a}^{-} & C_{0 a}^{0} & C_{0 a}^{+} & C_{0 b} & C_{0 c} & C_{0 d} \\
0 & 0 & 0 & 0 & 0 & C_{d} \\
0 & 0 & 0 & C_{b} & 0 & 0
\end{array}\right],  \tag{2.3.22}\\
& \tilde{D}_{*}=\Gamma_{o}^{-1} D_{*} \Gamma_{i}=\left[\begin{array}{ccc}
I_{m_{0}} & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right] . \tag{2.3.23}
\end{align*}
$$

A block diagram of the special coordinate basis of Theorem 2.3.1 is given in Figure 2.3.1. In this figure, a signal given by a double-edged arrow is some

$B_{0 b} y_{0}+L_{b d} y_{d}$


Note that a signal given by a double-edged arrow with a solid dot is some linear combination of all the states, whereas a signal given by a simple double-edged arrow is some linear combination of only output $y_{d}$. Also, matrices $B_{0 a}, L_{a b}$, $L_{a d}$ and $E_{c a}$ are to be defined in Property 2.3.1.

Figure 2.3.1: A block diagram representation of the special coordinate basis.
linear combination of outputs $y_{i}, i=0$ to $m_{d}$, where as a signal given by the double-edged arrow with a solid dot is some linear combination of all the states. Also, the block $D$ is either an integrator if $\Sigma_{*}$ is of continuous-time or a backward shifting operator if $\Sigma_{*}$ is of discrete-time.

We note the following intuitive points regarding the special coordinate basis.

1. The variable $u_{i}$ controls the output $y_{i}$ through a stack of $q_{i}$ integrators (or backward shifting operators), while $x_{i}$ is the state associated with those integrators (or backward shifting operators) between $u_{i}$ and $y_{i}$. Moreover, $\left(A_{q_{i}}, B_{q_{i}}\right)$ and ( $A_{q_{i}}, C_{q_{i}}$ ) respectively form controllable and observable pairs. This implies that all the states $x_{i}$ are both controllable and observable.
2. The output $y_{b}$ and the state $x_{b}$ are not directly influenced by any inputs, however they could be indirectly controlled through the output $y_{d}$. Moreover, $\left(A_{b b}, C_{b}\right)$ forms an observable pair. This implies that the state $x_{b}$ is observable.
3. The state $x_{c}$ is directly controlled by the input $u_{c}$, but it does not directly affect any output. Moreover, $\left(A_{c c}, B_{c}\right)$ forms a controllable pair. This implies that the state $x_{c}$ is controllable.
4. The state $x_{a}$ is neither directly controlled by any input nor does it directly affect any output.

In what follows, we state some important properties of the above special coordinate basis which are pertinent to our present work and will be used throughout this book. The proofs of these properties will be given in the next section.

Property 2.3.1. The given system $\Sigma_{*}$ is observable (detectable) if and only if the pair ( $A_{\text {obs }}, C_{\text {obs }}$ ) is observable (detectable), where

$$
A_{\mathrm{obs}}:=\left[\begin{array}{cc}
A_{a a} & 0  \tag{2.3.24}\\
B_{c} E_{c a} & A_{c c}
\end{array}\right], \quad C_{\mathrm{obs}}:=\left[\begin{array}{cc}
C_{0 a} & C_{0 c} \\
E_{d a} & E_{d c}
\end{array}\right]
$$

and where

$$
\begin{align*}
A_{a a} & :=\left[\begin{array}{ccc}
A_{a a}^{-} & 0 & 0 \\
0 & A_{a a}^{0} & 0 \\
0 & 0 & A_{a a}^{+}
\end{array}\right], \quad C_{0 a}:=\left[\begin{array}{lll}
C_{0 a}^{-} & C_{0 a}^{0} & C_{0 a}^{+}
\end{array}\right],  \tag{2.3.25}\\
E_{d a} & :=\left[\begin{array}{lll}
E_{d a}^{-} & E_{d a}^{0} & E_{d a}^{+}
\end{array}\right], \quad E_{c a}:=\left[\begin{array}{lll}
E_{c a}^{-} & E_{c a}^{0} & E_{c a}^{+}
\end{array}\right] . \tag{2.3.26}
\end{align*}
$$

Also, define

$$
\begin{align*}
& A_{\mathrm{con}}:=\left[\begin{array}{cc}
A_{a a} & L_{a b} C_{b} \\
0 & A_{b b}
\end{array}\right], \quad B_{\mathrm{con}}:=\left[\begin{array}{ll}
B_{0 a} & L_{a d} \\
B_{0 b} & L_{b d}
\end{array}\right],  \tag{2.3.27}\\
& B_{0 a}:=\left[\begin{array}{c}
B_{0 a}^{-} \\
B_{0 a}^{0} \\
B_{0 a}^{+}
\end{array}\right], \quad L_{a b}:=\left[\begin{array}{c}
L_{a b}^{-} \\
L_{a b}^{0} \\
L_{a b}^{+}
\end{array}\right], \quad L_{a d}:=\left[\begin{array}{c}
L_{a d}^{-} \\
L_{a d}^{0} \\
L_{a d}^{+}
\end{array}\right] . \tag{2.3.28}
\end{align*}
$$

Similarly, $\Sigma_{*}$ is controllable (stabilizable) if and only if the pair $\left(A_{\text {con }}, B_{\text {con }}\right)$ is controllable (stabilizable).

The invariant zeros of a system $\Sigma_{*}$ characterized by $\left(A_{*}, B_{*}, C_{*}, D_{*}\right)$ can be defined via the Smith canonical form of the (Rosenbrock) system matrix [84] of $\Sigma_{*}$ defined as the polynomial matrix $P_{\Sigma_{*}}(\varsigma)$,

$$
P_{\Sigma_{*}}(\varsigma):=\left[\begin{array}{cc}
\varsigma I-A_{*} & -B_{*}  \tag{2.3.29}\\
C_{*} & D_{*}
\end{array}\right] .
$$

We have the following definition for the invariant zeros (see also [68]).
Definition 2.3.1. (Invariant Zeros). A complex scalar $\alpha \in \mathbb{C}$ is said to be an invariant zero of $\Sigma_{*}$ if

$$
\begin{equation*}
\operatorname{rank}\left\{P_{\Sigma_{*}}(\alpha)\right\}<n+\text { normrank }\left\{H_{*}(\varsigma)\right\} \tag{2.3.30}
\end{equation*}
$$

where normrank $\left\{H_{*}(\varsigma)\right\}$ denotes the normal rank of $H_{*}(\varsigma)$, which is defined as its rank over the field of rational functions of $\varsigma$ with real coefficients.

The special coordinate basis of Theorem 2.3.1 shows explicitly the invariant zeros and the normal rank of $\Sigma_{*}$. To be more specific, we have the following properties.

## Property 2.3.2.

1. The normal rank of $H_{*}(\varsigma)$ is equal to $m_{0}+m_{d}$.
2. Invariant zeros of $\Sigma_{*}$ are the eigenvalues of $A_{a a}$, which are the unions of the eigenvalues of $A_{a a}^{-}, A_{a a}^{0}$ and $A_{a a}^{+}$. Moreover, the given system $\Sigma_{*}$ is of minimum phase if and only if $A_{a a}$ has only stable eigenvalues, marginal minimum phase if and only if $A_{a a}$ has no unstable eigenvalue but has at least one marginally stable eigenvalue, and nonminimum phase if and only if $A_{a a}$ has at least one unstable eigenvalue.

In order to display various multiplicities of invariant zeros, let $X_{a}$ be a nonsingular transformation matrix such that $A_{a a}$ can be transformed into a Jordan canonical form (see Theorem 2.1.1), i.e.,

$$
\begin{equation*}
X_{a}^{-1} A_{a a} X_{a}=J=\operatorname{blkdiag}\left\{J_{1}, J_{2}, \cdots, J_{k}\right\} \tag{2.3.31}
\end{equation*}
$$

where $J_{i}, i=1,2, \cdots, k$, are some $n_{i} \times n_{i}$ Jordan blocks:

$$
J_{i}=\operatorname{diag}\left\{\alpha_{i}, \alpha_{i}, \cdots, \alpha_{i}\right\}+\left[\begin{array}{cc}
0 & I_{n_{i}-1}  \tag{2.3.32}\\
0 & 0
\end{array}\right] .
$$

For any given $\alpha \in \lambda\left(A_{a a}\right)$, let there be $\tau_{\alpha}$ Jordan blocks of $A_{a a}$ associated with $\alpha$. Let $n_{\alpha, 1}, n_{\alpha, 2}, \cdots, n_{\alpha, \tau_{\alpha}}$ be the dimensions of the corresponding Jordan blocks. Then we say $\alpha$ is an invariant zero of $\Sigma_{*}$ with multiplicity structure $S_{\alpha}^{\star}\left(\Sigma_{*}\right)$ (see also [85]),

$$
\begin{equation*}
S_{\alpha}^{\star}\left(\Sigma_{*}\right)=\left\{n_{\alpha, 1}, n_{\alpha, 2}, \cdots, n_{\alpha, \tau_{\alpha}}\right\} \tag{2.3.33}
\end{equation*}
$$

The geometric multiplicity of $\alpha$ is then simply given by $\tau_{\alpha}$, and the algebraic multiplicity of $\alpha$ is given by $\sum_{i=1}^{\tau_{\alpha}} n_{\alpha, i}$. Here we should note that the invariant zeros together with their structures of $\Sigma_{*}$ are related to the structural invariant indices list $\mathcal{I}_{1}\left(\Sigma_{*}\right)$ of Morse [70].

The special coordinate basis can also reveal the infinite zero structure of $\Sigma_{*}$. We note that the infinite zero structure of $\Sigma_{*}$ can be either defined in association with root-locus theory or as Smith-McMillan zeros of the transfer function at infinity. For the sake of simplicity, we only consider the infinite zeros from the point of view of Smith-McMillan theory here. To define the zero structure of $H_{*}(\varsigma)$ at infinity, one can use the familiar Smith-McMillan description of the zero structure at finite frequencies of a general not necessarily square but strictly proper transfer function matrix $H_{*}(\varsigma)$. Namely, a rational matrix $H_{*}(\varsigma)$ possesses an infinite zero of order $k$ when $H_{*}(1 / z)$ has a finite zero of precisely that order at $z=0$ (see [35], [81], [84] and [108]). The number of zeros at infinity together with their orders indeed defines an infinite zero structure. Owens [73] related the orders of the infinite zeros of the root-loci of a square system with a non-singular transfer function matrix to $\mathcal{C}^{*}$ structural invariant indices list $\mathcal{I}_{4}$ of Morse [70]. This connection reveals that even for general not necessarily strictly proper systems, the structure at infinity is in fact the topology of inherent integrations between the input and the output variables. The special coordinate basis of Theorem 2.3.1 explicitly shows this topology of inherent integrations. The following property pinpoints this.

Property 2.3.3. $\Sigma_{*}$ has $m_{0}=\operatorname{rank}\left(D_{*}\right)$ infinite zeros of order 0 . The infinite zero structure (of order greater than 0 ) of $\Sigma_{*}$ is given by

$$
\begin{equation*}
S_{\infty}^{\star}\left(\Sigma_{*}\right)=\left\{q_{1}, q_{2}, \cdots, q_{m_{d}}\right\} \tag{2.3.34}
\end{equation*}
$$

That is, each $q_{i}$ corresponds to an infinite zero of $\Sigma_{*}$ of order $q_{i}$. Note that for a single-input-single-output. system $\Sigma_{*}$, we have $S_{\infty}^{\star}\left(\Sigma_{*}\right)=\left\{q_{1}\right\}$, where $q_{1}$ is the relative degree of $\Sigma_{*}$.

The special coordinate basis can also exhibit the invertibility structure of a given system $\Sigma_{*}$. The formal definitions of right invertibility and left invertibility of a linear system can be found in [71]. Basically, for the usual case when $\left[\begin{array}{ll}B_{*}^{\prime} & D_{*}^{\prime}\end{array}\right]$ and $\left[C_{*} D_{*}\right.$ ] are of maximal rank, the system $\Sigma_{*}$ or equivalently $H_{*}(\varsigma)$ is said to be left invertible if there exists a rational matrix function, say $L_{*}(\varsigma)$, such that

$$
\begin{equation*}
L_{*}(\varsigma) H_{*}(\varsigma)=I_{m} . \tag{2.3.35}
\end{equation*}
$$

$\Sigma_{*}$ or $H_{*}(\varsigma)$ is said to be right invertible if there exists a rational matrix function, say $R_{*}(\varsigma)$, such that

$$
\begin{equation*}
H_{*}(\varsigma) R_{*}(\varsigma)=I_{p} \tag{2.3.36}
\end{equation*}
$$

$\Sigma_{*}$ is invertible if it is both left and right invertible, and $\Sigma_{*}$ is degenerate if it is neither left nor right invertible.

Property 2.3.4. The given system $\Sigma_{*}$ is right invertible if and only if $x_{b}$ (and hence $y_{b}$ ) are non-existent, left invertible if and only if $x_{c}$ (and hence $u_{c}$ ) are non-existent, and invertible if and only if both $x_{b}$ and $x_{c}$ are non-existent. Moreover, $\Sigma_{*}$ is degenerate if and only if both $x_{b}$ and $x_{c}$ are present.

The special coordinate basis can also be modified to obtain the structural invariant indices lists $\mathcal{I}_{2}$ and $\mathcal{I}_{3}$ of Morse [70] of the given system $\Sigma_{*}$. In order to display $\mathcal{I}_{2}\left(\Sigma_{*}\right)$, we let $X_{c}$ and $X_{i}$ be non-singular matrices such that the controllable pair $\left(A_{c c}, B_{c}\right)$ is transformed into Brunovsky canonical form (see Theorem 2.2.1), i.e.,

$$
X_{c}^{-1} A_{c c} X_{c}=\left[\begin{array}{ccccc}
0 & I_{\ell_{1}-1} & \cdots & 0 & 0  \tag{2.3.37}\\
\star & \star & \cdots & \star & \star \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & 0 & I_{\ell_{\ell_{c}}-1} \\
\star & \star & \cdots & \star & \star
\end{array}\right], \quad X_{c}^{-1} B_{c} X_{i}=\left[\begin{array}{ccc}
0 & \cdots & 0 \\
1 & \cdots & 0 \\
\vdots & \ddots & \vdots \\
0 & \cdots & 0 \\
0 & \cdots & 1
\end{array}\right]
$$

where $\star$ 's denote constant scalars or row vectors. Then we have

$$
\begin{equation*}
\mathcal{I}_{2}\left(\Sigma_{*}\right)=\left\{\ell_{1}, \cdots, \ell_{m_{c}}\right\} \tag{2.3.38}
\end{equation*}
$$

which is also called the controllability index of $\left(A_{c c}, B_{c}\right)$. Similarly, we have

$$
\begin{equation*}
\mathcal{I}_{3}\left(\Sigma_{*}\right)=\left\{\mu_{1}, \cdots, \mu_{p_{b}}\right\} \tag{2.3.39}
\end{equation*}
$$

where $\left\{\mu_{1}, \cdots, \mu_{p_{b}}\right\}$ is the controllability index of the controllable pair $\left(A_{b b}^{\prime}, C_{b}^{\prime}\right)$.

By now it is clear that the special coordinate basis decomposes the statespace into several distinct parts. In fact, the state-space $\mathcal{X}$ is decomposed as

$$
\begin{equation*}
\mathcal{X}=\mathcal{X}_{a}^{-} \oplus \mathcal{X}_{a}^{0} \oplus \mathcal{X}_{a}^{+} \oplus \mathcal{X}_{b} \oplus \mathcal{X}_{c} \oplus \mathcal{X}_{d} \tag{2.3.40}
\end{equation*}
$$

Here $\mathcal{X}_{a}^{-}$is related to the stable invariant zeros, i.e., the eigenvalues of $A_{a a}^{-}$are the stable invariant zeros of $\Sigma_{*}$. Similarly, $\mathcal{X}_{a}^{0}$ and $\mathcal{X}_{a}^{+}$are respectively related to the invariant zeros of $\Sigma_{*}$ located in the marginally stable and unstable regions. On the other hand, $\mathcal{X}_{b}$ is related to the right invertibility, i.e., the system is right invertible if and only if $\mathcal{X}_{b}=\{0\}$, while $\mathcal{X}_{c}$ is related to left invertibility, i.e., the system is left invertible if and only if $\mathcal{X}_{c}=\{0\}$. Finally, $\mathcal{X}_{d}$ is related to zeros of $\Sigma_{*}$ at infinity.

There are interconnections between the special coordinate basis and various invariant geometric subspaces. To show these interconnections, we introduce the following geometric subspaces:

Definition 2.3.2. (Geometric Subspaces $\mathcal{V}^{\mathrm{x}}$ and $\mathcal{S}^{\mathrm{x}}$ ). The weakly unobservable subspaces of $\Sigma_{*}, \mathcal{V}^{\mathrm{x}}$, and the strongly controllable subspaces of $\Sigma_{*}$, $\mathcal{S}^{\mathrm{x}}$, are defined as follows:

1. $\mathcal{V}^{\mathrm{x}}\left(\Sigma_{*}\right)$ is the maximal subspace of $\mathbb{R}^{n}$ which is $\left(A_{*}+B_{*} F_{*}\right)$-invariant and contained in $\operatorname{Ker}\left(C_{*}+D_{*} F_{*}\right)$ such that the eigenvalues of $\left(A_{*}+B_{*} F_{*}\right) \mid \mathcal{V}^{\mathrm{x}}$ are contained in $\mathbb{C}^{\mathbf{x}} \subseteq \mathbb{C}$ for some constant matrix $F_{*}$.
2. $\mathcal{S}^{\mathrm{x}}\left(\Sigma_{*}\right)$ is the minimal $\left(A_{*}+K_{*} C_{*}\right)$-invariant subspace of $\mathbb{R}^{n}$ containing $\operatorname{Im}\left(B_{*}+K_{*} D_{*}\right)$ such that the eigenvalues of the map which is induced by $\left(A_{*}+K_{*} C_{*}\right)$ on the factor space $\mathbb{R}^{n} / \mathcal{S}^{\mathrm{x}}$ are contained in $\mathbb{C}^{\mathrm{x}} \subseteq \mathbb{C}$ for some constant matrix $K_{*}$.

Furthermore, we let $\mathcal{V}^{-}=\mathcal{V}^{\mathrm{x}}$ and $\mathcal{S}^{-}=\mathcal{S}^{\mathrm{x}}$, if $\mathbb{C}^{\mathrm{x}}=\mathbb{C}^{-} \cup \mathbb{C}^{0} ; \mathcal{V}^{+}=\mathcal{V}^{\mathrm{x}}$ and $\mathcal{S}^{+}=\mathcal{S}^{\mathrm{x}}$, if $\mathbb{C}^{\mathrm{x}}=\mathbb{C}^{+} ; \mathcal{V}^{\odot}=\mathcal{V}^{\mathrm{x}}$ and $\mathcal{S}^{\odot}=\mathcal{S}^{\mathrm{x}}$, if $\mathbb{C}^{\mathrm{x}}=\mathbb{C}^{\odot} \cup \mathbb{C}^{\mathrm{O}} ; \mathcal{V}^{\otimes}=\mathcal{V}^{\mathrm{x}}$ and $\mathcal{S}^{\otimes}=\mathcal{S}^{\mathrm{x}}$, if $\mathbb{C}^{\mathrm{x}}=\mathbb{C}^{\otimes}$; and finally $\mathcal{V}^{*}=\mathcal{V}^{\mathrm{x}}$ and $\mathcal{S}^{*}=\mathcal{S}^{\mathrm{x}}$, if $\mathbb{C}^{\mathrm{x}}=\mathbb{C}$.

Various components of the state vector of the special coordinate basis have the following geometrical interpretations.

## Property 2.3.5.

1. $\mathcal{X}_{a}^{-} \oplus \mathcal{X}_{a}^{0} \oplus \mathcal{X}_{c}$ spans $\begin{cases}\mathcal{V}^{-}\left(\Sigma_{*}\right), & \text { if } \Sigma_{*} \text { is of continuous-time }, \\ \mathcal{V}^{\odot}\left(\Sigma_{*}\right), & \text { if } \Sigma_{*} \text { is of discrete-time. }\end{cases}$
2. $\mathcal{X}_{a}^{+} \oplus \mathcal{X}_{c}$ spans $\begin{cases}\mathcal{V}^{+}\left(\Sigma_{*}\right), & \text { if } \Sigma_{*} \text { is of continuous-time }, \\ \mathcal{V}^{\otimes}\left(\Sigma_{*}\right), & \text { if } \Sigma_{*} \text { is of discrete-time. }\end{cases}$
3. $\mathcal{X}_{a}^{-} \oplus \mathcal{X}_{a}^{0} \oplus \mathcal{X}_{a}^{+} \oplus \mathcal{X}_{c}$ spans $\mathcal{V}^{*}\left(\Sigma_{*}\right)$.
4. $\mathcal{X}_{a}^{+} \oplus \mathcal{X}_{c} \oplus \mathcal{X}_{d}$ spans $\begin{cases}\mathcal{S}^{-}\left(\Sigma_{*}\right), & \text { if } \Sigma_{*} \text { is of continuous-time, } \\ \mathcal{S}^{\odot}\left(\Sigma_{*}\right), & \text { if } \Sigma_{*} \text { is of discrete-time. }\end{cases}$
5. $\mathcal{X}_{a}^{-} \oplus \mathcal{X}_{a}^{0} \oplus \mathcal{X}_{c} \oplus \mathcal{X}_{d}$ spans $\begin{cases}\mathcal{S}^{+}\left(\Sigma_{*}\right), & \text { if } \Sigma_{*} \text { is of continuous-time, } \\ \mathcal{S}^{\otimes}\left(\Sigma_{*}\right), & \text { if } \Sigma_{*} \text { is of discrete-time. }\end{cases}$
6. $\mathcal{X}_{c} \oplus \mathcal{X}_{d}$ spans $\mathcal{S}^{*}\left(\Sigma_{*}\right)$.

Finally, for future development on deriving solvability conditions for $H_{\infty}$ almost disturbance decoupling problems, we introduce two more subspaces of $\Sigma_{*}$. The original definitions of these subspaces were given by Scherer $[95,96]$.

Definition 2.3.3. (Geometric Subspaces $\mathcal{V}_{\lambda}$ and $\mathcal{S}_{\lambda}$ ). For any $\lambda \in \mathbb{C}$, we define

$$
\mathcal{V}_{\lambda}\left(\Sigma_{*}\right):=\left\{\zeta \in \mathbb{C}^{n} \mid \exists \omega \in \mathbb{C}^{m}: 0=\left[\begin{array}{cc}
A_{*}-\lambda I & B_{*}  \tag{2.3.41}\\
C_{*} & D_{*}
\end{array}\right]\binom{\zeta}{\omega}\right\}
$$

and

$$
\mathcal{S}_{\lambda}\left(\Sigma_{*}\right):=\left\{\zeta \in \mathbb{C}^{n} \mid \exists \omega \in \mathbb{C}^{n+m}:\binom{\zeta}{0}=\left[\begin{array}{cc}
A_{*}-\lambda I & B_{*}  \tag{2.3.42}\\
C_{*} & D_{*}
\end{array}\right] \omega\right\}
$$

$\mathcal{V}_{\lambda}\left(\Sigma_{*}\right)$ and $\mathcal{S}_{\lambda}\left(\Sigma_{*}\right)$ are associated with the so-called state zero directions of $\Sigma_{*}$ if $\lambda$ is an invariant zero of $\Sigma_{*}$.

These subspaces $\mathcal{S}_{\lambda}\left(\Sigma_{*}\right)$ and $\mathcal{V}_{\lambda}\left(\Sigma_{*}\right)$ can also be easily obtained using the special coordinate basis. We have the following new property of the special coordinate basis.

## Property 2.3.6.

$$
\mathcal{S}_{\lambda}\left(\Sigma_{*}\right)=\operatorname{Im}\left\{\Gamma_{s}\left[\begin{array}{cccc}
\lambda I-A_{a a} & 0 & 0 & 0  \tag{2.3.43}\\
0 & Y_{b \lambda} & 0 & 0 \\
0 & 0 & I_{n_{c}} & 0 \\
0 & 0 & 0 & I_{n_{d}}
\end{array}\right]\right\}
$$

where

$$
\begin{equation*}
\operatorname{Im}\left\{Y_{b \lambda}\right\}=\operatorname{Ker}\left[C_{b}\left(A_{b b}+K_{b} C_{b}-\lambda I\right)^{-1}\right], \tag{2.3.44}
\end{equation*}
$$

and where $K_{b}$ is any appropriately dimensional matrix subject to the constraint that $A_{b b}+K_{b} C_{b}$ has no eigenvalue at $\lambda$. We note that such a $K_{b}$ always exists as $\left(A_{b b}, C_{b}\right)$ is completely observable.

$$
\mathcal{V}_{\lambda}\left(\Sigma_{*}\right)=\operatorname{Im}\left\{\Gamma_{s}\left[\begin{array}{cc}
X_{a \lambda} & 0  \tag{2.3.45}\\
0 & 0 \\
0 & X_{c \lambda} \\
0 & 0
\end{array}\right]\right\}
$$

where $X_{a \lambda}$ is a matrix whose columns form a basis for the subspace,

$$
\begin{equation*}
\left\{\zeta_{a} \in \mathbb{C}^{n_{a}} \mid\left(\lambda I-A_{a a}\right) \zeta_{a}=0\right\} \tag{2.3.46}
\end{equation*}
$$

and

$$
\begin{equation*}
X_{c \lambda}:=\left(A_{c c}+B_{c} F_{c}-\lambda I\right)^{-1} B_{c} \tag{2.3.47}
\end{equation*}
$$

with $F_{c}$ being any appropriately dimensional matrix subject to the constraint that $A_{c c}+B_{c} F_{c}$ has no eigenvalue at $\lambda$. Again, we note that the existence of such an $F_{c}$ is guaranteed by the controllability of $\left(A_{c c}, B_{c}\right)$.

Clearly, if $\lambda \notin \lambda\left(A_{a a}\right)$, then we have

$$
\begin{equation*}
\mathcal{V}_{\lambda}\left(\Sigma_{*}\right) \subseteq \mathcal{V}^{\mathrm{x}}\left(\Sigma_{*}\right) \tag{2.3.48}
\end{equation*}
$$

and

$$
\begin{equation*}
\mathcal{S}_{\lambda}\left(\Sigma_{*}\right) \supseteq \mathcal{S}^{\mathbf{x}}\left(\Sigma_{*}\right) \tag{2.3.49}
\end{equation*}
$$

Next, we would like to note that the subspaces $\mathcal{V}^{\mathrm{x}}\left(\Sigma_{*}\right)$ and $\mathcal{S}^{\mathrm{x}}\left(\Sigma_{*}\right)$ are dual in the sense that $\mathcal{V}^{\mathrm{x}}\left(\Sigma_{*}^{\star}\right)=\mathcal{S}^{\mathrm{x}}\left(\Sigma_{*}\right)^{\perp}$, where $\Sigma_{*}^{\star}$ is characterized by the quadruple $\left(A_{*}^{\prime}, C_{*}^{\prime}, B_{*}^{\prime}, D_{*}^{\prime}\right)$. Also, $\mathcal{S}_{\lambda}\left(\Sigma_{*}\right)=\mathcal{V}_{\bar{\lambda}}\left(\Sigma_{*}^{\star}\right)^{\perp}$.

We illustrate the procedure for constructing the special coordinate basis and all its associated properties in the following numerical example.

Example 2.3.1. Consider a linear time-invariant system $\Sigma_{*}$ characterized by

$$
\left\{\begin{align*}
\delta(x) & =A_{*} x+B_{*} u  \tag{2.3.50}\\
y & =C_{*} x+D_{*} u
\end{align*}\right.
$$

where

$$
A_{*}=\left[\begin{array}{llll}
1 & 2 & 3 & 1  \tag{2.3.51}\\
2 & 3 & 4 & 5 \\
4 & 5 & 6 & 7 \\
5 & 6 & 7 & 8
\end{array}\right], \quad B_{*}=\left[\begin{array}{l}
1 \\
2 \\
3 \\
4
\end{array}\right]
$$

and

$$
C_{*}=\left[\begin{array}{llll}
0 & 3 & -2 & 0 \tag{2.3.52}
\end{array}\right], \quad D_{*}=0
$$

The procedure for constructing the special coordinate basis of $\Sigma_{*}$ proceeds as follows:

Step 1. Differentiating (shifting) the output of the given system. It involves the following sub-steps.

1. Since $D_{*}=0$, we have

$$
\delta(y)=C_{*} \delta(x)=C_{*} A_{*} x+C_{*} B_{*} u=\left[\begin{array}{cccc}
-2 & -1 & 0 & 1
\end{array}\right] x+0 \cdot u .
$$

2. Since $C_{*} B_{*}=0$, we compute

$$
\delta^{2}(y)=C_{*} A_{*}^{2} x+C_{*} A_{*} B_{*} u=\left[\begin{array}{llll}
1 & -1 & -3 & 1
\end{array}\right] x+0 \cdot u
$$

where $\delta^{2}(\cdot)=\delta(\delta(\cdot))$.
3. Since $C_{*} A_{*} B_{*}=0$, we continue on computing

$$
\delta^{3}(y)=C_{*} A_{*}^{3} x+C_{*} A_{*}^{2} B_{*}=-\left[\begin{array}{llll}
8 & 10 & 12 & 17
\end{array}\right] x-6 \cdot u
$$

where $\delta^{3}(\cdot)=\delta(\delta(\delta(\cdot)))$. Step 1 stops here as $C_{*} A_{*}^{2} B_{*} \neq 0$.
Step 2. Constructing a preliminary state transformation. Let $X_{0}$ be an appropriately dimensional matrix such that

$$
T=\left[\begin{array}{c}
X_{0}  \tag{2.3.53}\\
C_{*} \\
C_{*} A_{*} \\
C_{*} A_{*}^{2}
\end{array}\right],
$$

is non-singular. Then, define a new set of state variables $\check{x}$,

$$
\check{x}=\left(\begin{array}{c}
\check{x}_{1}  \tag{2.3.54}\\
\check{x}_{2} \\
\check{x}_{3} \\
\check{x}_{4}
\end{array}\right):=T x=\left[\begin{array}{c}
X_{0} \\
C_{*} \\
C_{*} A_{*} \\
C_{*} A_{*}^{2}
\end{array}\right] x=\left(\begin{array}{c}
X_{0} x \\
y \\
\delta(y) \\
\delta^{2}(y)
\end{array}\right) .
$$

It is simple to verify that $T$ with $X_{0}=\left[\begin{array}{llll}1 & 0 & 0 & 0\end{array}\right]$ is a non-singular matrix. Furthermore,

$$
\begin{align*}
& \delta\left(\check{x}_{1}\right)=8 \check{x}_{1}+\check{x}_{2}+\frac{8}{3} \check{x}_{3}-\frac{5}{3} \check{x}_{4}+u  \tag{2.3.55}\\
& \delta\left(\check{x}_{2}\right)=\check{x}_{3}  \tag{2.3.56}\\
& \delta\left(\check{x}_{3}\right)=\check{x}_{4}  \tag{2.3.57}\\
& \delta\left(\check{x}_{4}\right)=-72 \check{x}_{1}-9 \check{x}_{2}-27 \check{x}_{3}+10 \check{x}_{4}-6 u . \tag{2.3.58}
\end{align*}
$$

Step 3. Eliminating $u$ in $\delta\left(\check{x}_{1}\right)$. Equation (2.3.58) implies that

$$
\begin{equation*}
u=-12 \check{x}_{1}-\frac{3}{2} \check{x}_{2}-\frac{9}{2} \check{x}_{3}+\frac{5}{3} \check{x}_{4}-\frac{1}{6} \delta\left(\check{x}_{4}\right) . \tag{2.3.59}
\end{equation*}
$$

Substituting this into (2.3.55), we obtain

$$
\begin{equation*}
\delta\left(\check{x}_{1}\right)=-4 \check{x}_{1}-\frac{1}{2} \check{x}_{2}-\frac{11}{6} \check{x}_{3}-\frac{1}{6} \delta\left(\check{x}_{4}\right) \tag{2.3.60}
\end{equation*}
$$

We have got rid of $u$ in $\delta\left(\check{x}_{1}\right)$. Unfortunately, we have also introduced an additional $\delta\left(\check{x}_{4}\right)$ in (2.3.60).

Step 4. Eliminating $\delta\left(\check{x}_{4}\right)$ in $\delta\left(\check{x}_{1}\right)$. Define a new variable $\bar{x}_{1}$ as follows,

$$
\begin{equation*}
\bar{x}_{1}:=\check{x}_{1}+\frac{1}{6} \check{x}_{4} . \tag{2.3.61}
\end{equation*}
$$

We have

$$
\begin{equation*}
\delta\left(\bar{x}_{1}\right)=-4 \bar{x}_{1}-\frac{1}{2} \check{x}_{2}-\frac{11}{6} \check{x}_{3}+\frac{2}{3} \check{x}_{4}, \tag{2.3.62}
\end{equation*}
$$

and

$$
\begin{equation*}
\delta\left(\check{x}_{4}\right)=-72 \bar{x}_{1}-9 \check{x}_{2}-27 \check{x}_{3}+22 \check{x}_{4}-6 u . \tag{2.3.63}
\end{equation*}
$$

Step 5. Eliminating $\check{x}_{3}$ and $\check{x}_{4}$ in $\delta\left(\bar{x}_{1}\right)$. This step involves two sub-steps.

1. Letting

$$
\begin{equation*}
\hat{x}_{1}:=\bar{x}_{1}-\frac{2}{3} \check{x}_{3} \tag{2.3.64}
\end{equation*}
$$

we have

$$
\begin{equation*}
\delta\left(\hat{x}_{1}\right)=-4 \hat{x}_{1}-\frac{1}{2} \check{x}_{2}-\frac{9}{2} \check{x}_{3} \tag{2.3.65}
\end{equation*}
$$

and

$$
\begin{equation*}
\delta\left(\check{x}_{4}\right)=-72 \hat{x}_{1}-9 \check{x}_{2}-75 \check{x}_{3}+22 \check{x}_{4}-6 u \tag{2.3.66}
\end{equation*}
$$

2. Letting

$$
\begin{equation*}
\tilde{x}_{1}:=\hat{x}_{1}+\frac{9}{2} \check{x}_{2} \tag{2.3.67}
\end{equation*}
$$

we have

$$
\begin{equation*}
\delta\left(\tilde{x}_{1}\right)=-4 \tilde{x}_{1}+\frac{35}{2} \check{x}_{2} \tag{2.3.68}
\end{equation*}
$$

and

$$
\begin{equation*}
\delta\left(\check{x}_{4}\right)=-72 \tilde{x}_{1}+315 \check{x}_{2}-75 \check{x}_{3}+22 \check{x}_{4}-6 u \tag{2.3.69}
\end{equation*}
$$

Step 6. Forming the non-singular state, output and input transformations. Let

$$
\begin{equation*}
\tilde{x}_{2}=\check{x}_{2}, \quad \tilde{x}_{3}=\check{x}_{3}, \quad \tilde{x}_{3}=\check{x}_{3} \tag{2.3.70}
\end{equation*}
$$

or equivalently let

$$
x=\Gamma_{s} \tilde{x}=\Gamma_{s}\left(\begin{array}{c}
\tilde{x}_{1}  \tag{2.3.71}\\
\tilde{x}_{2} \\
\tilde{x}_{3} \\
\tilde{x}_{4}
\end{array}\right),
$$

with

$$
\Gamma_{s}=\left\{\left[\begin{array}{cccc}
1 & 9 / 2 & -2 / 3 & 1 / 6  \tag{2.3.72}\\
0 & 1 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 0 & 0 & 1
\end{array}\right]\left[\begin{array}{rrrr}
1 & 0 & 0 & 0 \\
0 & 3 & -2 & 0 \\
-2 & -1 & 0 & 1 \\
1 & -1 & -3 & 1
\end{array}\right]\right\}^{-1}
$$

Also, let

$$
\begin{equation*}
u=\Gamma_{i} \tilde{u}=-\frac{1}{6} \tilde{u}, \quad y=\Gamma_{o} \tilde{y}=1 \cdot y \tag{2.3.73}
\end{equation*}
$$

Finally, we obtain the dynamic equations of the transformed system,

$$
\begin{align*}
& \delta\left(\tilde{x}_{1}\right)=-4 \tilde{x}_{1}+\frac{35}{2} \tilde{x}_{2}  \tag{2.3.74}\\
& \delta\left(\tilde{x}_{2}\right)=\tilde{x}_{3}, \quad \tilde{y}=\tilde{x}_{2}  \tag{2.3.75}\\
& \delta\left(\tilde{x}_{3}\right)=\tilde{x}_{4}  \tag{2.3.76}\\
& \delta\left(\tilde{x}_{4}\right)=-72 \tilde{x}_{1}+315 \tilde{x}_{2}-75 \tilde{x}_{3}+22 \tilde{x}_{4}+\tilde{u} . \tag{2.3.77}
\end{align*}
$$

The above structure is now in the standard form of the special coordinate basis. $\tilde{x}_{1}$ is associated with $\mathcal{X}_{a}$ and $\tilde{x}_{2}, \tilde{x}_{3}$ and $\tilde{x}_{4}$ are associated with $\mathcal{X}_{d}$. Both $\mathcal{X}_{b}$ and $\mathcal{X}_{c}$ are non-existent for the given $\Sigma_{*}$.

Let us now examine the properties of $\Sigma_{*}$. Following Properties 2.3.1 to 2.3.6 of the special coordinate basis, it is simple to verify that $\Sigma_{*}$ is controllable and observable, and has an invariant zero at -4 as well as an infinite zero (relative degree) of order 3. It is obvious that the given system is invertible as both $x_{c}$ and $x_{b}$ are non-existent.

The geometric subspaces $\mathcal{V}_{\lambda}\left(\Sigma_{*}\right)$ and $\mathcal{S}_{\lambda}\left(\Sigma_{*}\right)$ can be obtained as follows: for $\lambda=-4$,

$$
\begin{gather*}
\mathcal{V}_{\lambda}\left(\Sigma_{*}\right)=\operatorname{Im}\left\{\left[\begin{array}{l}
3 \\
2 \\
3 \\
8
\end{array}\right]\right\},  \tag{2.3.78}\\
\mathcal{S}_{\lambda}\left(\Sigma_{*}\right)=\operatorname{Im}\left\{\left[\begin{array}{lll}
1 & 2 & 27 \\
2 & 2 & 16 \\
3 & 3 & 27 \\
4 & 9 & 70
\end{array}\right]\right\}, \tag{2.3.79}
\end{gather*}
$$

and for $\lambda \neq-4$,

$$
\begin{equation*}
\mathcal{V}_{\lambda}\left(\Sigma_{*}\right)=\{0\}, \quad \mathcal{S}_{\lambda}\left(\Sigma_{*}\right)=\mathbb{R}^{4} \tag{2.3.80}
\end{equation*}
$$

The geometric subspaces $\mathcal{V}^{\mathrm{x}}\left(\Sigma_{*}\right)$ and $\mathcal{S}^{\mathrm{x}}\left(\Sigma_{*}\right)$ of $\Sigma_{*}$ can also be easily computed:

1. If $\Sigma_{*}$ is a continuous-time system, then

$$
\mathcal{V}^{-}\left(\Sigma_{*}\right)=\mathcal{V}^{*}\left(\Sigma_{*}\right)=\operatorname{Im}\left\{\left[\begin{array}{l}
3  \tag{2.3.81}\\
2 \\
3 \\
8
\end{array}\right]\right\}, \quad \mathcal{V}^{+}\left(\Sigma_{*}\right)=\{0\}
$$

and

$$
\mathcal{S}^{-}\left(\Sigma_{*}\right)=\mathcal{S}^{*}\left(\Sigma_{*}\right)=\operatorname{Im}\left\{\left[\begin{array}{lll}
1 & 2 & 27  \tag{2.3.82}\\
2 & 2 & 16 \\
3 & 3 & 27 \\
4 & 9 & 70
\end{array}\right]\right\}, \quad \mathcal{S}^{+}\left(\Sigma_{*}\right)=\mathbb{R}^{4}
$$

2. If $\Sigma_{*}$ is a discrete-time system, then

$$
\mathcal{V}^{\otimes}\left(\Sigma_{*}\right)=\mathcal{V}^{*}\left(\Sigma_{*}\right)=\operatorname{Im}\left\{\left[\begin{array}{l}
3  \tag{2.3.83}\\
2 \\
3 \\
8
\end{array}\right]\right\}, \quad \mathcal{V}^{\odot}\left(\Sigma_{*}\right)=\{0\}
$$

and

$$
\mathcal{S}^{\otimes}\left(\Sigma_{*}\right)=\mathcal{S}^{*}\left(\Sigma_{*}\right)=\operatorname{Im}\left\{\left[\begin{array}{ccc}
1 & 2 & 27  \tag{2.3.84}\\
2 & 2 & 16 \\
3 & 3 & 27 \\
4 & 9 & 70
\end{array}\right]\right\}, \quad \mathcal{S}^{\odot}\left(\Sigma_{*}\right)=\mathbb{R}^{4}
$$

Here we would like to note that the computation of the special coordinate basis for a multiple-input-multiple-output system is of course much more complicated than that for a single-input-single-output system, but the idea is basically the same.

Finally, we conclude this section by summarizing in graphical form in Figure 2.3.2 some major properties of the X-transformer of linear systems, which combines the mechanisms of the special coordinate basis, the Jordan canonical form and the Brunovsky canonical form. Such a transformer has been used in the literature to solve many system and control problems such as the squaring down and decoupling of linear systems (see e.g., Sannuti and Saberi [93]), linear system factorizations (see e.g., Chen et al [27], and Lin et al [64]), blocking zeros and strong stabilizability (see e.g., Chen et al [28]), zero placements (see e.g. Chen and Zheng [33]), loop transfer recovery (see e.g., Chen [10], Chen and Chen [16], and Saberi et al [87]), $H_{2}$ optimal control (see e.g., Chen et al [29,31], and Saberi et al [88]), disturbance decoupling (see e.g., Chen [15], and Ozcetin et al [74,75]), and control with saturations (see e.g., Lin [61]), to name a few. This X-transformer will be used intensively throughout this book to solve problems related to $H_{\infty}$ control.


Figure 2.3.2: An X-transformer of linear time-invariant systems.

### 2.4. Proofs of Properties of Special Coordinate Basis

In this section, we provide detailed proofs for all the properties of the special coordinate basis listed in the previous section. Somehow, these proofs were missing in the original work of Sannuti and Saberi [93]. And unfortunately, somehow, they are still missing in the literature. We would like to note that although some of the properties of the special coordinate basis, e.g., the controllability and observability, are quite obvious, some of them, e.g., the interconnections between the geometric subspaces and the subsystems of the special coordinate basis, are not transparent at all to general readers. The goal of this section is to give rigorous proofs to all these properties once for all.

We recall the following two lemmas whose results are quite well-known in the literature. The first lemma is about the effects of state feedback laws.

Lemma 2.4.1. Consider a given system $\Sigma_{*}$ characterized by a constant matrix quadruple ( $A_{*}, B_{*}, C_{*}, D_{*}$ ) or in the state space form of (2.3.1). Also, consider a constant state feedback gain matrix $F_{*} \in \mathbb{R}^{m \times n}$. Then, $\Sigma_{* \mathrm{~F}}$ characterized by the quadruple $\left(A_{*}+B_{*} F_{*}, B_{*}, C_{*}+D_{*} F_{*}, D_{*}\right)$ has the following properties:

1. $\Sigma_{* \mathrm{~F}}$ is a controllable (stabilizable) system if and only if $\Sigma_{*}$ is a controllable (stabilizable) system;
2. The normal rank of $\Sigma_{* F}$ is equal to that of $\Sigma_{*}$;
3. The invariant zero structure of $\Sigma_{* F}$ is the same as that of $\Sigma_{*}$;
4. The infinite zero structure of $\Sigma_{* F}$ is the same as that of $\Sigma_{*}$;
5. $\Sigma_{* F}$ is (left- or right- or non-) invertible if and only if $\Sigma_{*}$ is (left- or rightor non-) invertible;
6. $\mathcal{V}^{\mathrm{x}}\left(\Sigma_{* \mathrm{~F}}\right)=\mathcal{V}^{\mathrm{x}}\left(\Sigma_{*}\right)$ and $\mathcal{S}^{\mathrm{x}}\left(\Sigma_{* \mathrm{~F}}\right)=\mathcal{S}^{\mathrm{x}}\left(\Sigma_{*}\right)$; and
7. $\mathcal{V}_{\lambda}\left(\Sigma_{* \mathrm{~F}}\right)=\mathcal{V}_{\lambda}\left(\Sigma_{*}\right)$ and $\mathcal{S}_{\lambda}\left(\Sigma_{* \mathrm{~F}}\right)=\mathcal{S}_{\lambda}\left(\Sigma_{*}\right)$.

Proof. Items 1 is obvious. Items 3, 4 and 5 are well-known as all the lists of Morse, i.e., $\mathcal{I}_{1}$ to $\mathcal{I}_{4}$, are invariant under any state feedback laws. Furthermore, items 2 and 5 can be seen trivially from the following simple manipulations:

$$
\begin{align*}
H_{* \mathrm{~F}}(\varsigma) & :=\left(C_{*}+D_{*} F_{*}\right)\left(\varsigma I-A_{*}-B_{*} F_{*}\right)^{-1} B_{*}+D_{*} \\
& =\left(C_{*}+D_{*} F_{*}\right)\left(\varsigma I-A_{*}\right)^{-1}\left[I-B_{*} F_{*}\left(\varsigma I-A_{*}\right)^{-1}\right]^{-1} B_{*}+D_{*} \\
& =\left(C_{*}+D_{*} F_{*}\right)\left(\varsigma I-A_{*}\right)^{-1} B_{*}\left[I-F_{*}\left(\varsigma I-A_{*}\right)^{-1} B_{*}\right]^{-1}+D_{*} \\
& =\left[C_{*}\left(\varsigma I-A_{*}\right)^{-1} B_{*}+D_{*}\right]\left[I-F_{*}\left(\varsigma I-A_{*}\right)^{-1} B_{*}\right]^{-1} \\
& =H_{*}(\varsigma)\left[I-F_{*}\left(\varsigma I-A_{*}\right)^{-1} B_{*}\right]^{-1} . \tag{2.4.1}
\end{align*}
$$

Since $\left[I-F_{*}\left(\varsigma I-A_{*}\right)^{-1} B_{*}\right]^{-1}$ is well-defined almost everywhere on the complex plane, the results of items 2 and 5 follow.

For item 6, it is obvious from the definition of $\mathcal{V}^{\mathrm{x}}$, it is invariant under any state feedback laws. Next, for any subspace $\mathcal{S}$ that satisfies the following conditions:

$$
\begin{array}{r}
\left(A_{*}+K_{*} C_{*}\right) \mathcal{S} \subseteq \mathcal{S} \\
\operatorname{Im}\left(B_{*}+K_{*} D_{*}\right) \subseteq \mathcal{S} \tag{2.4.3}
\end{array}
$$

we have

$$
\left(A_{*}+K_{*} C_{*}+B_{*} F_{*}+K_{*} D_{*} F_{*}\right) \mathcal{S}=\left(A_{*}+K_{*} C_{*}\right) \mathcal{S}+\left(B_{*}+K_{*} D_{*}\right) F_{*} \mathcal{S} \subseteq \mathcal{S}
$$

Thus, $\mathcal{S}^{\mathbf{x}}$ is invariant under any state feedback laws as well.
Let us now prove item 7. Recalling the definition of $\mathcal{V}_{\lambda}$, we have

$$
\mathcal{V}_{\lambda}\left(\Sigma_{* \mathbf{F}}\right)=\left\{\zeta \in \mathbb{C}^{n} \mid \exists \omega \in \mathbb{C}^{m}: 0=\left[\begin{array}{cc}
A_{*}+B_{*} F_{*}-\lambda I & B_{*} \\
C_{*}+D_{*} F_{*} & D_{*}
\end{array}\right]\binom{\zeta}{\omega}\right\}
$$

Then, for any $\zeta \in \mathcal{V}_{\lambda}\left(\Sigma_{* F}\right)$, there exist an $\omega \in \mathbb{C}^{m}$ such that

$$
0=\left[\begin{array}{cc}
A_{*}+B_{*} F_{*}-\lambda I & B_{*} \\
C_{*}+D_{*} F_{*} & D_{*}
\end{array}\right]\binom{\zeta}{\omega}=\left[\begin{array}{cc}
A_{*}-\lambda I & B_{*} \\
C_{*} & D_{*}
\end{array}\right]\left[\begin{array}{cc}
I & 0 \\
F_{*} & I
\end{array}\right]\binom{\zeta}{\omega},
$$

or

$$
0=\left[\begin{array}{cc}
A_{*}-\lambda I & B_{*} \\
C_{*} & D_{*}
\end{array}\right]\binom{\zeta}{\tilde{\omega}}
$$

where $\tilde{\omega}=F_{*} \zeta+\omega$. Thus, $\zeta \in \mathcal{V}_{\lambda}\left(\Sigma_{*}\right)$ and hence $\mathcal{V}_{\lambda}\left(\Sigma_{* \mathrm{~F}}\right) \subseteq \mathcal{V}_{\lambda}\left(\Sigma_{*}\right)$. Similarly, one can show that $\mathcal{V}_{\lambda}\left(\Sigma_{*}\right) \subseteq \mathcal{V}_{\lambda}\left(\Sigma_{* F}\right)$, and hence $\mathcal{V}_{\lambda}\left(\Sigma_{*}\right)=\mathcal{V}_{\lambda}\left(\Sigma_{* F}\right)$. The result that $\mathcal{S}_{\lambda}\left(\Sigma_{* \mathrm{~F}}\right)=\mathcal{S}_{\lambda}\left(\Sigma_{*}\right)$ can be shown using the similar arguments.

The following lemma is about the effects of output injection laws.
Lemma 2.4.2. Consider a given system $\Sigma_{*}$ characterized by a constant matrix quadruple ( $A_{*}, B_{*}, C_{*}, D_{*}$ ) or in the state space form of (2.3.1). Also, consider a constant output injection gain matrix $K_{*} \in \mathbb{R}^{n \times p}$. Then, $\Sigma_{* \mathrm{~K}}$ characterized by the quadruple $\left(A_{*}+K_{*} C_{*}, B_{*}+K_{*} D_{*}, C_{*}, D_{*}\right)$ has the following properties:

1. $\Sigma_{* \mathrm{~K}}$ is an observable (detectable) system if and only if $\Sigma_{*}$ is an observable (detectable) system;
2. The normal rank of $\Sigma_{* \mathrm{~K}}$ is equal to that of $\Sigma_{*}$;
3. The invariant zero structure of $\Sigma_{* \mathrm{~K}}$ is the same as that of $\Sigma_{*}$;
4. The infinite zero structure of $\Sigma_{* K}$ is the same as that of $\Sigma_{*}$;
5. $\Sigma_{* \mathrm{~K}}$ is (left- or right- or non-) invertible if and only if $\Sigma_{*}$ is (left- or rightor non-) invertible;
6. $\mathcal{V}^{\mathrm{x}}\left(\Sigma_{* \mathrm{~K}}\right)=\mathcal{V}^{\mathrm{x}}\left(\Sigma_{*}\right)$ and $\mathcal{S}^{\mathrm{x}}\left(\Sigma_{* \mathrm{~K}}\right)=\mathcal{S}^{\mathrm{x}}\left(\Sigma_{*}\right)$; and
7. $\mathcal{V}_{\lambda}\left(\Sigma_{* \mathrm{~K}}\right)=\mathcal{V}_{\lambda}\left(\Sigma_{*}\right)$ and $\mathcal{S}_{\lambda}\left(\Sigma_{* \mathrm{~K}}\right)=\mathcal{S}_{\lambda}\left(\Sigma_{*}\right)$.

Proof. It is the dual version of Lemma 2.4.1.

Now, we are ready to prove the properties of the special coordinate basis. Without loss of any generality but for simplicity of presentation, we assume throughout the rest of this section that the given system $\Sigma_{*}$ has already been transformed into the special coordinate basis of Theorem 2.3.1 or into the compact form of (2.3.20) to (2.3.23), i.e.,

$$
\begin{gather*}
A_{*}=\left[\begin{array}{cccc}
A_{a a} & L_{a b} C_{b} & 0 & L_{a d} C_{d} \\
0 & A_{b b} & 0 & L_{b d} C_{d} \\
B_{c} E_{c a} & L_{c b} C_{b} & A_{c c} & L_{c d} C_{d} \\
B_{d} E_{d a} & B_{d} E_{d b} & B_{d} E_{d c} & A_{d d}^{*}+B_{d} E_{d d}+L_{d d} C_{d}
\end{array}\right]+B_{*, 0} C_{*, 0}  \tag{2.4.4}\\
 \tag{2.4.5}\\
B_{*}=\left[\begin{array}{ll}
B_{*, 0} & B_{*, 1}
\end{array}\right]=\left[\begin{array}{ccc}
B_{0 a} & 0 & 0 \\
B_{0 b} & 0 & 0 \\
B_{0 c} & 0 & B_{c} \\
B_{0 d} & B_{d} & 0
\end{array}\right]
\end{gather*}
$$

and

$$
C_{*}=\left[\begin{array}{l}
C_{*, 0}  \tag{2.4.6}\\
C_{*, 1}
\end{array}\right]=\left[\begin{array}{cccc}
C_{0 a} & C_{0 b} & C_{0 c} & C_{0 d} \\
0 & 0 & 0 & C_{d} \\
0 & C_{b} & 0 & 0
\end{array}\right], \quad D_{*}=\left[\begin{array}{ccc}
I_{m_{0}} & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right] .
$$

We further note that $A_{d d}^{*}, B_{d}$ and $C_{d}$ have the following forms,

$$
\begin{equation*}
A_{d d}^{*}=\operatorname{blkdiag}\left\{A_{q_{1}}, \cdots, A_{q_{m_{d}}}\right\} \tag{2.4.7}
\end{equation*}
$$

and

$$
\begin{equation*}
B_{d}=\operatorname{blkdiag}\left\{B_{q_{1}}, \cdots, B_{q_{m_{d}}}\right\}, \quad C_{d}=\operatorname{blkdiag}\left\{C_{q_{1}}, \cdots, C_{q_{m_{d}}}\right\} \tag{2.4.8}
\end{equation*}
$$

where $A_{q_{i}}, B_{q_{i}}$ and $C_{q_{i}}, i=1,2, \cdots, m_{d}$, are defined as in (2.3.16).
Proof of Property 2.3.1. Let us define a state feedback gain matrix $F_{*}$ as follows:

$$
F_{*}=-\left[\begin{array}{cccc}
C_{0 a} & C_{0 b} & C_{0 c} & C_{0 d}  \tag{2.4.9}\\
E_{d a} & E_{d b} & E_{d c} & E_{d d} \\
E_{c a} & 0 & 0 & 0
\end{array}\right] .
$$

Then, we have

$$
A_{*}+B_{*} F_{*}=\left[\begin{array}{cccc}
A_{a a} & L_{a b} C_{b} & 0 & L_{a d} C_{d}  \tag{2.4.10}\\
0 & A_{b b} & 0 & L_{b d} C_{d} \\
0 & L_{c b} C_{b} & A_{c c} & L_{c d} C_{d} \\
0 & 0 & 0 & A_{d d}^{*}+L_{d d} C_{d}
\end{array}\right]
$$

Noting that $\left(A_{c c}, B_{c}\right)$ is completely controllable, we have for any $\lambda \in \mathbb{C}$,

$$
\begin{align*}
& \operatorname{rank}\left[A_{*}+B_{*} F_{*}-\lambda I \quad B_{*}\right] \\
& =\operatorname{rank}\left[\begin{array}{ccccccc}
A_{a a}-\lambda I & L_{a b} C_{b} & 0 & L_{a d} C_{d} & B_{0 a} & 0 & 0 \\
0 & A_{b b}-\lambda I & 0 & L_{b d} C_{d} & B_{0 b} & 0 & 0 \\
0 & L_{c b} C_{b} & A_{c c}-\lambda I & L_{c d} C_{d} & B_{0 c} & 0 & B_{c} \\
0 & 0 & 0 & A_{d d}^{*}+L_{d d} C_{d}-\lambda I & B_{0 d} & B_{d} & 0
\end{array}\right] \\
& =\operatorname{rank}\left[\begin{array}{cccccc}
A_{a a}-\lambda I & L_{a b} C_{b} & 0 & L_{a d} C_{d} & B_{0 a} & 0 \\
0 & A_{b b}-\lambda I & 0 & L_{b d} C_{d} & B_{0 b} & 0 \\
0 \\
0 & 0 & A_{c c}-\lambda I & 0 & 0 & 0 \\
B_{c} \\
0 & 0 & 0 & A_{d d}^{*}+L_{d d} C_{d}-\lambda I & B_{0 d} & B_{d} \\
0
\end{array}\right] \\
& =\operatorname{rank}\left[\begin{array}{cccccc}
A_{\text {con }}-\lambda I & 0 & B_{\text {con1 }} C_{d} & B_{\text {con } 0} & 0 & 0 \\
0 & A_{c c}-\lambda I & 0 & 0 & 0 & B_{c} \\
0 & 0 & A_{d d}^{*}+L_{d d} C_{d}-\lambda I & B_{0 d} & B_{d} & 0
\end{array}\right],(2.4 .11) \tag{2.4.11}
\end{align*}
$$

where

$$
A_{\mathrm{con}}=\left[\begin{array}{cc}
A_{a a} & L_{a b} C_{b}  \tag{2.4.12}\\
0 & A_{b b}
\end{array}\right], \quad B_{\mathrm{con}}=\left[\begin{array}{ll}
B_{\mathrm{con} 0} & B_{\mathrm{con} 1}
\end{array}\right]=\left[\begin{array}{ll}
B_{0 a} & L_{a d} \\
B_{0 b} & L_{b d}
\end{array}\right]
$$

Also, noting the special structure of $\left(A_{d d}^{*}, B_{d}, C_{d}\right)$, it is simple to verify that $\left[A_{*}+B_{*} F_{*}-\lambda I \quad B_{*}\right]$ is of maximal rank if and only if $\left[\begin{array}{lll}A_{\text {con }}-\lambda I & B_{\text {con }}\end{array}\right]$ is of maximal rank. By Lemma 2.4.1, we have that $(A, B)$ is controllable (stabilizable) if and only if ( $A_{\text {con }}, B_{\text {con }}$ ) is controllable (stabilizable).

Similarly, one can show that ( $A, C$ ) is observable (detectable) if and only if ( $A_{\text {obs }}, C_{\text {obs }}$ ) is observable (detectable).

Proof of Property 2.3.2. Let us define a state feedback gain matrix $F_{*}$ as in (2.4.9) and an output injection gain matrix $K_{*}$ as follows:

$$
K_{*}=-\left[\begin{array}{ccc}
B_{0 a} & L_{a d} & L_{a b}  \tag{2.4.13}\\
B_{0 b} & L_{b d} & 0 \\
B_{0 c} & L_{c d} & L_{c b} \\
B_{0 d} & L_{d d} & 0
\end{array}\right]
$$

We have

$$
\begin{gather*}
\check{A}_{*}=A_{*}+B_{*} F_{*}+K_{*} C_{*}+K_{*} D_{*} F_{*}=\left[\begin{array}{cccc}
A_{a a} & 0 & 0 & 0 \\
0 & A_{b b} & 0 & 0 \\
0 & 0 & A_{c c} & 0 \\
0 & 0 & 0 & A_{d d}^{*}
\end{array}\right],  \tag{2.4.14}\\
\check{B}_{*}=B_{*}+K_{*} D_{*}=\left[\begin{array}{ccc}
0 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & B_{c} \\
0 & B_{d} & 0
\end{array}\right],  \tag{2.4.15}\\
\check{C}_{*}=C_{*}+D_{*} F_{*}=\left[\begin{array}{cccc}
0 & 0 & 0 & 0 \\
0 & 0 & 0 & C_{d} \\
0 & C_{b} & 0 & 0
\end{array}\right] \tag{2.4.16}
\end{gather*}
$$

and

$$
\check{D}_{*}=D_{*}=\left[\begin{array}{ccc}
I_{m_{0}} & 0 & 0  \tag{2.4.17}\\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right]
$$

Let $\check{\Sigma}_{*}$ be characterized by the quadruple ( $\check{A}_{*}, \check{B}_{*}, \check{C}_{*}, \check{D}_{*}$ ). It is simple to verify that the transfer function of $\check{\Sigma}_{*}$ is given by

$$
\check{H}_{*}(\varsigma)=\check{C}_{*}\left(\varsigma I-\check{A}_{*}\right)^{-1} \check{B}_{*}+\check{D}_{*}=\left[\begin{array}{ccc}
I_{m_{0}} & 0 & 0  \tag{2.4.18}\\
0 & C_{d}\left(\varsigma I-A_{d d}^{*}\right)^{-1} B_{d} & 0 \\
0 & 0 & 0
\end{array}\right]
$$

Furthermore, we can show that

$$
C_{d}\left(\varsigma I-A_{d d}^{*}\right)^{-1} B_{d}=\left[\begin{array}{lll}
\frac{1}{\varsigma^{q_{1}}} & &  \tag{2.4.19}\\
& \ddots & \\
& & \frac{1}{\varsigma^{q_{m_{d}}}}
\end{array}\right] .
$$

By Lemmas 2.4.1 and 2.4.2, we have

$$
\begin{equation*}
\operatorname{normrank}\left\{H_{*}(\varsigma)\right\}=\operatorname{normrank}\left\{\check{H}_{*}(\varsigma)\right\}=m_{0}+m_{d} \tag{2.4.20}
\end{equation*}
$$

Next, it follows from Lemmas 2.4.1 and 2.4.2 that the invariant zeros of $\Sigma_{*}$ and $\check{\Sigma}_{*}$ are equivalent. By the definition of the invariant zeros of a linear system, i.e., a complex scalar $\alpha$ is an invariant zero of $\check{\Sigma}_{*}$ if

$$
\operatorname{rank}\left[\begin{array}{cc}
\check{A}_{*}-\alpha I & \check{B}_{*}  \tag{2.4.21}\\
\check{C}_{*} & \check{D}_{*}
\end{array}\right]<n+\operatorname{normrank}\left\{\check{H}_{*}(\varsigma)\right\}=n+m_{0}+m_{d}
$$

and also noting the special structure of $\left(A_{d d}^{*}, B_{d}, C_{d}\right)$ and the facts that $\left(A_{b b}, C_{b}\right)$ is observable，and $\left(A_{c c}, B_{c}\right)$ is controllable，we have

$$
\begin{align*}
\operatorname{rank}\left\{P_{\check{\Sigma}_{*}}(\alpha)\right\} & =\operatorname{rank}\left[\begin{array}{ccccccc}
\check{A}_{*}-\alpha I & \check{B}_{*} \\
\check{C}_{*} & \check{D}_{*}
\end{array}\right] \\
& =\operatorname{rank}\left[\begin{array}{ccccccc}
A_{a a}-\alpha I & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & A_{b b}-\alpha I & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & A_{c c}-\alpha I & 0 & 0 & 0 & B_{c} \\
0 & 0 & 0 & A_{d d}^{*}-\alpha I & 0 & B_{d} & 0 \\
0 & 0 & 0 & 0 & I_{m_{0}} & 0 & 0 \\
0 & 0 & 0 & C_{d} & 0 & 0 & 0 \\
0 & C_{b} & 0 & 0 & 0 & 0 & 0
\end{array}\right] \\
& =n_{b}+n_{c}+n_{d}+m_{0}+m_{d}+\operatorname{rank}\left\{A_{a a}-\alpha I\right\} . \tag{2.4.22}
\end{align*}
$$

Clearly，the rank of $P_{\Sigma^{*}}(\alpha)$ drops below $n+m_{0}+m_{d}$ if and only if $\alpha \in \lambda\left(A_{a a}\right)$ ． Hence，the invariant zeros of $\check{\Sigma}_{*}$ ，or equivalently the invariant zeros of $\Sigma_{*}$ ，are given by the eigenvalues of $A_{a a}$ ，which are the union of $\lambda\left(A_{a a}^{-}\right), \lambda\left(A_{a a}^{0}\right)$ ，and $\lambda\left(A_{a a}^{+}\right)$．This completes the proof of Property 2．3．2．

Proof of Property 2．3．3．It follows from Lemmas 2.4 .1 and 2.4 .2 that the infinite zeros of $\Sigma_{*}$ and $\check{\Sigma}_{*}$ are equivalent．It is clear to see from（2．4．18）and （2．4．19）that the infinite zeros of $\check{\Sigma}_{*}$ ，or equivalently the infinite zeros of $\Sigma_{*}$ ，of order higher than 0 ，are given by

$$
\begin{equation*}
S_{\infty}^{\star}\left(\Sigma_{*}\right)=S_{\infty}^{\star}\left(\check{\Sigma}_{*}\right)=\left\{q_{1}, q_{2}, \cdots, \boldsymbol{q}_{m_{d}}\right\} \tag{2.4.23}
\end{equation*}
$$

Furthermore，$\check{\Sigma}_{*}$ or $\Sigma_{*}$ has $m_{0}$ infinite zeros of order 0 ．

Proof of Property 2．3．4．Again，it follows from Lemmas 2．4．1 and 2．4．2 that $\Sigma_{*}$ or $H_{*}(\varsigma)$ is（left－or right－or non－）invertible if and only if $\check{\Sigma}_{*}$ or $\breve{H}_{*}(\varsigma)$ is （left－or right－or non－）invertible．The results of Property 2．3．4 can be seen trivially from the transfer function $\check{H}_{*}(\varsigma)$ in（2．4．18）．

因

Proof of Property 2．3．5．We will only prove the geometric subspace $\mathcal{V}^{*}\left(\Sigma_{*}\right)$ ， i．e．，

$$
\mathcal{V}^{*}\left(\Sigma_{*}\right)=\mathcal{X}_{a} \oplus \mathcal{X}_{c}=\operatorname{Im}\left\{\Gamma_{s}\left[\begin{array}{cc}
I_{n_{a}} & 0  \tag{2.4.24}\\
0 & 0 \\
0 & I_{n_{c}} \\
0 & 0
\end{array}\right]\right\}
$$

Here $\Gamma_{s}=I_{n}$ as the given system $\Sigma_{*}$ is assumed to be in the form of the special coordinate basis already．It follows from Lemma 2．4．2 that $\mathcal{V}^{*}$ is invariant under
any output injection laws. Let us choose an output injection gain matrix $K_{*}$ as in (2.4.13). Then, we have

$$
\hat{A}_{*}=A_{*}+K_{*} C_{*}=\left[\begin{array}{cccc}
A_{a a} & 0 & 0 & 0  \tag{2.4.25}\\
0 & A_{b b} & 0 & 0 \\
B_{c} E_{c a} & 0 & A_{c c} & 0 \\
B_{d} E_{d a} & B_{d} E_{d b} & B_{d} E_{d c} & A_{d d}^{*}+B_{d} E_{d d}
\end{array}\right]
$$

and

$$
\hat{B}_{*}=B_{*}+K_{*} D_{*}=\check{B}_{*}=\left[\begin{array}{ccc}
0 & 0 & 0  \tag{2.4.26}\\
0 & 0 & 0 \\
0 & 0 & B_{c} \\
0 & B_{d} & 0
\end{array}\right]
$$

Let $\hat{\Sigma}_{*}$ be a system characterized by $\left(\hat{A}_{*}, \hat{B}_{*}, C_{*}, D_{*}\right)$. Then it is sufficient to show the property of $\mathcal{V}^{*}\left(\Sigma_{*}\right)$ by showing that

$$
\mathcal{V}^{*}\left(\hat{\Sigma}_{*}\right)=\operatorname{Im}\left\{\left[\begin{array}{cc}
I_{n_{a}} & 0  \tag{2.4.27}\\
0 & 0 \\
0 & I_{n_{c}} \\
0 & 0
\end{array}\right]\right\}
$$

First, let us choose a matrix $F_{*}$ as given in (2.4.9). Then, we have

$$
\hat{A}_{*}+\hat{B}_{*} F_{*}=\left[\begin{array}{cccc}
A_{a a} & 0 & 0 & 0  \tag{2.4.28}\\
0 & A_{b b} & 0 & 0 \\
0 & 0 & A_{c c} & 0 \\
0 & 0 & 0 & A_{d d}^{*}
\end{array}\right],
$$

and

$$
C_{*}+D_{*} F_{*}=\left[\begin{array}{cccc}
0 & 0 & 0 & 0  \tag{2.4.29}\\
0 & 0 & 0 & C_{d} \\
0 & C_{b} & 0 & 0
\end{array}\right]
$$

It is simple to see now that for any

$$
\zeta \in \mathcal{X}_{a} \oplus \mathcal{X}_{c}=\operatorname{Im}\left\{\left[\begin{array}{cc}
I_{n_{a}} & 0  \tag{2.4.30}\\
0 & 0 \\
0 & I_{n_{c}} \\
0 & 0
\end{array}\right]\right\}
$$

we have

$$
\zeta=\left(\begin{array}{c}
\zeta_{a}  \tag{2.4.31}\\
0 \\
\zeta_{c} \\
0
\end{array}\right)
$$

and

$$
\left(\hat{A}_{*}+\hat{B}_{*} F_{*}\right) \zeta=\left(\begin{array}{c}
A_{a a} \zeta_{a}  \tag{2.4.32}\\
0 \\
A_{c c} \zeta_{c} \\
0
\end{array}\right) \in \operatorname{Im}\left\{\left[\begin{array}{cc}
I_{n_{a}} & 0 \\
0 & 0 \\
0 & I_{n_{c}} \\
0 & 0
\end{array}\right]\right\}=\mathcal{X}_{a} \oplus \mathcal{X}_{c}
$$

and

$$
\begin{equation*}
\left(C_{*}+D_{*} F_{*}\right) \zeta=0 \tag{2.4.33}
\end{equation*}
$$

Clearly, $\mathcal{X}_{a} \oplus \mathcal{X}_{c}$ is a $\left(\hat{A}_{*}+\hat{B}_{*} F_{*}\right)$-invariant subspace of $\mathbb{R}^{n}$ and is contained in $\operatorname{Ker}\left(C_{*}+D_{*} F_{*}\right)$. By the definition of $\mathcal{V}^{*}$, we have

$$
\begin{equation*}
\mathcal{X}_{a} \oplus \mathcal{X}_{c} \subseteq \mathcal{V}^{*}\left(\hat{\Sigma}_{*}\right) . \tag{2.4.34}
\end{equation*}
$$

Conversely, for any $\zeta \in \mathcal{V}^{*}\left(\hat{\Sigma}_{*}\right)$, by Definition 2.3.2, there exists a gain matrix $\hat{F}_{*} \in \mathbb{R}^{m \times n}$ such that

$$
\begin{equation*}
\left(\hat{A}_{*}+\hat{B}_{*} \hat{F}_{*}\right) \zeta \in \mathcal{V}^{*}\left(\hat{\Sigma}_{*}\right), \tag{2.4.35}
\end{equation*}
$$

and

$$
\begin{equation*}
\left(C_{*}+D_{*} \hat{F}_{*}\right) \zeta=0 \tag{2.4.36}
\end{equation*}
$$

(2.4.35) and (2.4.36) imply that for any $\zeta \in \mathcal{V}^{*}\left(\hat{\Sigma}_{*}\right)$,

$$
\begin{equation*}
\left(C_{*}+D_{*} \hat{F}_{*}\right)\left(\hat{A}_{*}+\hat{B}_{*} \hat{F}_{*}\right)^{k} \zeta=0, \quad k=0,1, \cdots, n-1 \tag{2.4.37}
\end{equation*}
$$

Thus, (2.4.34) and (2.4.37) imply that

$$
\left(C_{*}+D_{*} \hat{F}_{*}\right)\left(\hat{A}_{*}+\hat{B}_{*} \hat{F}_{*}\right)^{k}\left[\begin{array}{cc}
I_{n_{a}} & 0  \tag{2.4.38}\\
0 & 0 \\
0 & I_{n_{c}} \\
0 & 0
\end{array}\right]=0, k=0,1, \cdots, n-1
$$

Next, let us partition this $\hat{F}_{*}$ as follows:

$$
\hat{F}_{*}=\left[\begin{array}{cccc}
F_{a 0}-C_{0 a} & F_{b 0}-C_{0 b} & F_{c 0}-C_{0 c} & F_{d 0}-C_{0 d}  \tag{2.4.39}\\
F_{a d}-E_{d a} & F_{b d}-E_{d b} & F_{c d}-E_{d c} & F_{d d}-E_{d d} \\
F_{a c}-E_{c a} & F_{b c} & F_{c c} & F_{d c}
\end{array}\right] .
$$

We have

$$
C_{*}+D_{*} \hat{F}_{*}=\left[\begin{array}{cccc}
F_{a 0} & F_{b 0} & F_{c 0} & F_{d 0}  \tag{2.4.40}\\
0 & 0 & 0 & C_{d} \\
0 & C_{b} & 0 & 0
\end{array}\right]
$$

and

$$
\hat{A}_{*}+\hat{B}_{*} \hat{F}_{*}=\left[\begin{array}{cccc}
A_{a a} & 0 & 0 & 0  \tag{2.4.41}\\
0 & A_{b b} & 0 & 0 \\
B_{c} F_{a c} & B_{c} F_{b c} & A_{c c}+B_{c} F_{c c} & B_{c} F_{d c} \\
B_{d} F_{a d} & B_{d} F_{b d} & B_{d} F_{c d} & A_{d d}^{* *}
\end{array}\right]
$$

where $A_{d d}^{* *}=A_{d d}^{*}+B_{d} F_{d d}$. Then, using (2.4.38) with $k=0$, we have

$$
\left(C_{*}+D_{*} \hat{F}_{*}\right)\left[\begin{array}{cc}
I_{n_{a}} & 0  \tag{2.4.42}\\
0 & 0 \\
0 & I_{n_{c}} \\
0 & 0
\end{array}\right]=0
$$

which implies

$$
\begin{equation*}
F_{a 0}=0, \quad F_{c 0}=0, \tag{2.4.43}
\end{equation*}
$$

and

$$
C_{*}+D_{*} \hat{F}_{*}=\left[\begin{array}{cccc}
0 & \star & 0 & \star  \tag{2.4.44}\\
0 & 0 & 0 & C_{d} \\
0 & C_{b} & 0 & 0
\end{array}\right]
$$

where $\star$ 's are some matrices of not much interest. Using (2.4.38) with $k=1$ together with (2.4.44), we have

$$
\begin{equation*}
C_{d} B_{d} F_{a d}=0, \quad C_{d} B_{d} F_{c d}=0 \tag{2.4.45}
\end{equation*}
$$

and

$$
\left(C_{*}+D_{*} \hat{F}_{*}\right)\left(\hat{A}_{*}+\hat{B}_{*} \hat{F}_{*}\right)=\left[\begin{array}{cccc}
0 & \star & 0 & \star  \tag{2.4.46}\\
0 & C_{d} B_{d} F_{b d} & 0 & C_{d} A_{d d}^{* *} \\
0 & C_{b} A_{b b} & 0 & 0
\end{array}\right]
$$

In general, one can show that for any positive integer $k$,

$$
\begin{equation*}
C_{d}\left(A_{d d}^{* *}\right)^{k-1} B_{d} F_{a d}=0, \quad C_{d}\left(A_{d d}^{* *}\right)^{k-1} B_{d} F_{c d}=0 \tag{2.4.47}
\end{equation*}
$$

and

$$
\left(C_{*}+D_{*} \hat{F}_{*}\right)\left(\hat{A}_{*}+\hat{B}_{*} \hat{F}_{*}\right)^{k}=\left[\begin{array}{cccc}
0 & \star & 0 & \star  \tag{2.4.48}\\
0 & \star & 0 & C_{d}\left(A_{d d}^{* *}\right)^{k} \\
0 & C_{b}\left(A_{b b}\right)^{k} & 0 & 0
\end{array}\right]
$$

As a by product, we can easily show that $F_{a d}=0$ and $F_{c d}=0$, because of the fact that $\left(A_{d d}^{* *}, B_{d}, C_{d}\right)$ is controllable, observable, invertible and is free of invariant zeros. Now, for any

$$
\zeta=\left(\begin{array}{l}
\zeta_{a}  \tag{2.4.49}\\
\zeta_{b} \\
\zeta_{c} \\
\zeta_{d}
\end{array}\right) \in \mathcal{V}^{*}\left(\hat{\Sigma}_{*}\right)
$$

it follows from (2.4.37) and (2.4.48) that

$$
\begin{equation*}
C_{b}\left(A_{b b}\right)^{k} \zeta_{b}=0, \quad k=0,1, \cdots, n-1 \tag{2.4.50}
\end{equation*}
$$

which implies $\zeta_{b}=0$ because $\left(A_{b b}, C_{b}\right)$ is completely observable, and

$$
\begin{equation*}
C_{d}\left(A_{d d}^{* *}\right)^{k} \zeta_{d}+\star \cdot \zeta_{b}=C_{d}\left(A_{d d}^{* *}\right)^{k} \zeta_{d}=0, \quad k=0,1, \cdots, n-1 \tag{2.4.51}
\end{equation*}
$$

which implies $\zeta_{d}=0$ because $\left(A_{d d}^{* *}, C_{d}\right)$ is also completely observable. Hence,

$$
\zeta=\left(\begin{array}{c}
\zeta_{a}  \tag{2.4.52}\\
0 \\
\zeta_{c} \\
0
\end{array}\right) \in \operatorname{Im}\left\{\left[\begin{array}{cc}
I_{n_{a}} & 0 \\
0 & 0 \\
0 & I_{n_{c}} \\
0 & 0
\end{array}\right]\right\}=\mathcal{X}_{a} \oplus \mathcal{X}_{c}
$$

and

$$
\begin{equation*}
\mathcal{V}^{*}\left(\hat{\Sigma}_{*}\right) \subseteq \mathcal{X}_{a} \oplus \mathcal{X}_{c} \tag{2.4.53}
\end{equation*}
$$

Obviously, (2.4.34) and (2.4.53) imply the result.
Similarly, one can follow the same procedure as in the above to show the properties of the other subspaces in Property 2.3.5.

Proof of Property 2.3.6. Let us prove the property of $\mathcal{V}_{\lambda}\left(\Sigma_{*}\right)$. It follows from Lemmas 2.4.1 and 2.4.2 that $\mathcal{V}_{\lambda}$ is invariant under any state feedback and output injection laws. Thus, it is sufficient to prove the property of $\mathcal{V}_{\lambda}\left(\Sigma_{*}\right)$ by showing that

$$
\mathcal{V}_{\lambda}\left(\check{\Sigma}_{*}\right)=\operatorname{Im}\left\{\left[\begin{array}{cc}
X_{a \lambda} & 0  \tag{2.4.54}\\
0 & 0 \\
0 & X_{c \lambda} \\
0 & 0
\end{array}\right]\right\}
$$

where $\check{\Sigma}_{*}$ is as defined in the proof of Property 2.3.2, $X_{a \lambda}$ is a matrix whose columns form a basis for the subspace,

$$
\begin{equation*}
\left\{\zeta_{a} \in \mathbb{C}^{n_{a}} \mid\left(\lambda I-A_{a a}\right) \zeta_{a}=0\right\} \tag{2.4.55}
\end{equation*}
$$

and

$$
\begin{equation*}
X_{c \lambda}=\left(A_{c c}+B_{c} F_{c}-\lambda I\right)^{-1} B_{c} \tag{2.4.56}
\end{equation*}
$$

with $F_{c}$ being an appropriately dimensional matrix such that $A_{c c}+B_{c} F_{c}-\lambda I$ is invertible.

For any $\zeta \in \mathcal{V}_{\lambda}\left(\check{\Sigma}_{*}\right)$, by Definition 2.3.3, there exists a vector $\omega \in \mathbb{C}^{m}$ such that

$$
\left[\begin{array}{cc}
\check{A}_{*}-\lambda I & \check{B}_{*}  \tag{2.4.57}\\
\check{C}_{*} & \check{D}_{*}
\end{array}\right]\binom{\zeta}{\omega}=0
$$

or equivalently,

$$
\left[\begin{array}{ccccccc}
A_{a a}-\lambda I & 0 & 0 & 0 & 0 & 0 & 0  \tag{2.4.58}\\
0 & A_{b b}-\lambda I & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & A_{c c}-\lambda I & 0 & 0 & 0 & B_{c} \\
0 & 0 & 0 & A_{d d}^{*}-\lambda I & 0 & B_{d} & 0 \\
0 & 0 & 0 & 0 & I_{m_{0}} & 0 & 0 \\
0 & 0 & 0 & C_{d} & 0 & 0 & 0 \\
0 & C_{b} & 0 & 0 & 0 & 0 & 0
\end{array}\right]\left(\begin{array}{c}
\zeta_{a} \\
\zeta_{b} \\
\zeta_{c} \\
\zeta_{d} \\
\omega_{0} \\
\omega_{d} \\
\omega_{c}
\end{array}\right)=0
$$

Hence, we have

$$
\begin{equation*}
\left(A_{a a}-\lambda I\right) \zeta_{a}=0 \tag{2.4.59}
\end{equation*}
$$

which implies that $\zeta_{a} \in \operatorname{Im}\left\{X_{a \lambda}\right\}$,

$$
\left[\begin{array}{c}
A_{b b}-\lambda I  \tag{2.4.60}\\
C_{b}
\end{array}\right] \zeta_{b}=0
$$

which implies that $\zeta_{b}=0$ as $\left(A_{b b}, C_{b}\right)$ is completely observable, and

$$
\left[\begin{array}{cc}
A_{d d}^{*}-\lambda I & B_{d}  \tag{2.4.61}\\
C_{d} & 0
\end{array}\right]\binom{\zeta_{d}}{\omega_{d}}=0
$$

which implies that $\zeta_{d}=0$ and $\omega_{d}=0$ as $\left(A_{d d}^{*}, B_{d}, C_{d}\right)$ is square invertible and is free of invariant zeros. We also have

$$
\begin{equation*}
\left(A_{c c}-\lambda I\right) \zeta_{c}+B_{c} \omega_{c}=0 \tag{2.4.62}
\end{equation*}
$$

which implies that

$$
\begin{equation*}
\left(A_{c c}+B_{c} F_{c}-\lambda I\right) \zeta_{c}+B_{c}\left(\omega_{c}-F_{c} \zeta_{c}\right)=0 \tag{2.4.63}
\end{equation*}
$$

or

$$
\begin{equation*}
\zeta_{c}=\left(A_{c c}+B_{c} F_{c}-\lambda I\right)^{-1} B_{c}\left(F_{c} \zeta_{c}-\omega_{c}\right)=X_{c \lambda}\left(F_{c} \zeta_{c}-\omega_{c}\right) . \tag{2.4.64}
\end{equation*}
$$

Hence $\zeta_{c} \in \operatorname{Im}\left\{X_{c \lambda}\right\}$. Clearly,

$$
\zeta \in \operatorname{Im}\left\{\left[\begin{array}{cc}
X_{a \lambda} & 0  \tag{2.4.65}\\
0 & 0 \\
0 & X_{c \lambda} \\
0 & 0
\end{array}\right]\right\} \Rightarrow \mathcal{V}_{\lambda}\left(\check{\Sigma}_{*}\right) \subseteq \operatorname{Im}\left\{\left[\begin{array}{cc}
X_{a \lambda} & 0 \\
0 & 0 \\
0 & X_{c \lambda} \\
0 & 0
\end{array}\right]\right\}
$$

Conversely, for any

$$
\zeta=\left(\begin{array}{l}
\zeta_{a}  \tag{2.4.66}\\
\zeta_{b} \\
\zeta_{c} \\
\zeta_{d}
\end{array}\right) \in \operatorname{Im}\left\{\left[\begin{array}{cc}
X_{a \lambda} & 0 \\
0 & 0 \\
0 & X_{c \lambda} \\
0 & 0
\end{array}\right]\right\}
$$

we have $\zeta_{b}=0, \zeta_{d}=0, \zeta_{a} \in \operatorname{Im}\left\{X_{a \lambda}\right\}$, which implies that $\left(\lambda I-A_{a a}\right) \zeta_{a}=0$, and $\zeta_{c} \in \operatorname{Im}\left\{X_{c \lambda}\right\}$, which implies that there exists a vector $\tilde{\omega}_{c}$ such that

$$
\begin{equation*}
\zeta_{c}=X_{c \lambda} \tilde{\omega}_{c}=\left(A_{c c}+B_{c} F_{c}-\lambda I\right)^{-1} B_{c} \tilde{\omega}_{c} \tag{2.4.67}
\end{equation*}
$$

Thus, we have

$$
\begin{equation*}
\left(A_{c c}+B_{c} F_{c}-\lambda I\right) \zeta_{c}=B_{c} \tilde{\omega}_{c} \tag{2.4.68}
\end{equation*}
$$

or

$$
\begin{equation*}
\left(A_{c c}-\lambda I\right) \zeta_{c}+B_{c}\left(F_{c} \zeta_{c}-\tilde{\omega}_{c}\right)=0 \tag{2.4.69}
\end{equation*}
$$

Let

$$
\omega=\left(\begin{array}{c}
\omega_{0}  \tag{2.4.70}\\
\omega_{d} \\
\omega_{c}
\end{array}\right)=\left(\begin{array}{c}
0 \\
0 \\
F_{c} \zeta_{c}-\tilde{\omega}_{c}
\end{array}\right) .
$$

It is now straightforward to verify using (2.4.58) that

$$
\left[\begin{array}{cc}
\check{A}_{*}-\lambda I & \check{B}_{*}  \tag{2.4.71}\\
\check{C}_{*} & \check{D}_{*}
\end{array}\right]\binom{\zeta}{\omega}=0 .
$$

By Definition 2.3.3, we have

$$
\zeta \in \mathcal{V}_{\lambda}\left(\check{\Sigma}_{*}\right) \Longrightarrow \quad \operatorname{Im}\left\{\left[\begin{array}{cc}
X_{a \lambda} & 0  \tag{2.4.72}\\
0 & 0 \\
0 & X_{c \lambda} \\
0 & 0
\end{array}\right]\right\} \subseteq \mathcal{V}_{\lambda}\left(\check{\Sigma}_{*}\right)
$$

Finally, (2.4.65) and (2.4.72) imply the result.
The proof of $\mathcal{S}_{\lambda}\left(\Sigma_{*}\right)$ follows from the same lines of reasoning.

## Chapter 3

## Existence Conditions of $H_{\infty}$ Suboptimal Controllers

In an $H_{\infty}$ optimization problem, the first fundamental issue one faces is when, or under what conditions does a $\gamma$ suboptimal controller exist. Fortunately, the problem regarding the existence conditions of $\gamma$-suboptimal controllers for either the regular or singular type of continuous-time or discrete-time systems has almost been completely solved in the literature. For the regular continuous-time systems, the problem was solved by Doyle et al [39] and Tadmor [105]. For general singular continuous-time systems with no invariant zero on the imaginary axis, the problem was solved by Stoorvogel and Trentelman [104] and Stoorvogel [100]. In the situation when systems have invariant zeros on the imaginary axis, the result was derived by Scherer [94-96]. The existence conditions of $\gamma$-suboptimal controllers for discrete-time systems were reported in Stoorvogel [100] and Stoorvogel, Saberi and Chen [102]. In this chapter, we will recall the above mentioned results as they will form a base for the results reported in the second part of this book.

### 3.1. Continuous-time Systems

We consider in this section a general continuous-time linear time-invariant (LTI) system $\Sigma$ with a state-space description,

$$
\Sigma:\left\{\begin{array}{l}
\dot{x}=A x+B u+E w  \tag{3.1.1}\\
y=C_{1} x+D_{1} w \\
h=C_{2} x+D_{2} u+D_{22} w
\end{array}\right.
$$

where $x \in \mathbb{R}^{n}$ is the state, $u \in \mathbb{R}^{m}$ is the control input, $w \in \mathbb{R}^{q}$ is the external disturbance input, $y \in \mathbb{R}^{p}$ is the measurement output, and $h \in \mathbb{R}^{\ell}$ is the
controlled output of $\Sigma$. We also consider the following proper measurement feedback control law,

$$
\Sigma_{\mathrm{cmp}}:\left\{\begin{array}{l}
\dot{v}=A_{\mathrm{cmp}} v+B_{\mathrm{cmp}} y  \tag{3.1.2}\\
u=C_{\mathrm{cmp}} v+D_{\mathrm{cmp}} y
\end{array}\right.
$$

For simplicity of presentation, we will first set the direct feedthrough term from the disturbance $w$ to controlled output $h$ in (3.1.1) to be equal to zero, i.e., $D_{22}=0$. For easy reference, we define $\Sigma_{\mathrm{P}}$ to be the subsystem characterized by the matrix quadruple ( $A, B, C_{2}, D_{2}$ ), and $\Sigma_{\mathrm{Q}}$ to be the subsystem characterized by the matrix quadruple $\left(A, E, C_{1}, D_{1}\right)$, which respectively have transfer functions:

$$
\begin{equation*}
G_{\mathrm{P}}(s)=C_{2}(s I-A)^{-1} B+D_{2} \tag{3.1.3}
\end{equation*}
$$

and

$$
\begin{equation*}
G_{\mathrm{Q}}(s)=C_{1}(s I-A)^{-1} E+D_{1} \tag{3.1.4}
\end{equation*}
$$

We recall in this section some important results in the literature regarding the existence conditions of $\gamma$-suboptimal control laws for the continuous-time $H_{\infty}$ optimization problem.

The first result given below is due to Stoorvogel [100]. Before we introduce the theorem, let us define the following quadratic matrices,

$$
F_{\gamma}(P):=\left[\begin{array}{cc}
A^{\prime} P+P A+C_{2}^{\prime} C_{2}+\gamma^{-2} P E E^{\prime} P & P B+C_{2}^{\prime} D_{2}  \tag{3.1.5}\\
B^{\prime} P+D_{2}^{\prime} C_{2} & D_{2}^{\prime} D_{2}
\end{array}\right]
$$

and

$$
G_{\gamma}(Q):=\left[\begin{array}{cc}
A Q+Q A^{\prime}+E E^{\prime}+\gamma^{-2} Q C_{2}^{\prime} C_{2} Q & Q C_{1}^{\prime}+E D_{1}^{\prime}  \tag{3.1.6}\\
C_{1} Q+D_{1} E^{\prime} & D_{1} D_{1}^{\prime}
\end{array}\right] .
$$

It should be noted that the above matrices are dual of each other. In addition to these two matrices, we define two polynomial matrices whose roles are again completely dual:

$$
\begin{equation*}
L_{\gamma}(P, s):=\left[s I-A-\gamma^{-2} E E^{\prime} P \quad-B\right] \tag{3.1.7}
\end{equation*}
$$

and

$$
M_{\gamma}(Q, s):=\left[\begin{array}{c}
s I-A-\gamma^{-2} Q C_{2}^{\prime} C_{2}  \tag{3.1.8}\\
-C_{1}
\end{array}\right]
$$

Now we are ready to introduce the following theorem which gives a set of necessary and sufficient conditions for the existence of a $\gamma$-suboptimal controller for the continuous-time system (3.1.1) with $D_{22}=0$ and with both subsystems $\Sigma_{\mathrm{P}}$ and $\Sigma_{\mathrm{Q}}$ having no invariant zero on the imaginary axis.

Theorem 3.1.1. Consider the continuous-time linear time-invariant system of (3.1.1) with $D_{22}=0$. Assume that $\Sigma_{P}$ and $\Sigma_{Q}$ have no invariant zero on the imaginary axis. Then the following statements are equivalent:

1. There exists a linear time-invariant and proper dynamic compensator $\Sigma_{\mathrm{cmp}}$ of (3.1.2) such that by applying it to (3.1.1) the resulting closed-loop system is internally stable. Moreover, the $H_{\infty}$-norm of the closed-loop transfer function from the disturbance input $w$ to the controlled output $h$ is less than $\gamma$.
2. There exist positive semi-definite matrices $P$ and $Q$ such that the following conditions are satisfied:
(a) $F_{\gamma}(P) \geq 0$.
(b) $\operatorname{rank}\left\{F_{\gamma}(P)\right\}=\operatorname{normrank}\left\{G_{\mathrm{P}}(s)\right\}$.
(c) $\operatorname{rank}\left[\begin{array}{c}L_{\gamma}(P, s) \\ F_{\gamma}(P)\end{array}\right]=n+\operatorname{normrank}\left\{G_{\mathrm{P}}(s)\right\}, \forall s \in \mathbb{C}^{0} \cup \mathbb{C}^{+}$.
(d) $G_{\gamma}(Q) \geq 0$.
(e) $\operatorname{rank}\left\{G_{\gamma}(Q)\right\}=\operatorname{normrank}\left\{G_{\mathrm{Q}}(s)\right\}$.
(f) $\operatorname{rank}\left[M_{\gamma}(Q, s), G_{\gamma}(Q)\right]=n+\operatorname{normrank}\left\{G_{Q}(s)\right\}, \forall s \in \mathbb{C}^{0} \cup \mathbb{C}^{+}$.
(g) $\rho(P Q)<\gamma^{2}$.

Here $G_{\mathrm{P}}(s)$ and $G_{\mathrm{Q}}(s)$ are respectively the transfer function of $\Sigma_{\mathrm{P}}$ and $\Sigma_{\mathrm{Q}}$, and "normrank" denotes the rank of a matrix with entries in the field of rational functions.

The following remark concerns the full information feedback and full state feedback cases. It turns out that for the system with $D_{22}=0$, the existence conditions of $\gamma$-suboptimal controllers for the full information feedback case and for the full state feedback case are identical.

Remark 3.1.1. For the special cases of full information and full state feedback, the solution to the linear matrix inequality (LMI), i.e., condition 2.(d) of Theorem 3.1.1, which satisfies conditions 2.(e) and 2.(f), is identically zero. This implies that condition $2 .(\mathrm{g})$ is automatically satisfied. Hence, the existence conditions of $\gamma$-suboptimal controllers for both the full information and the full state feedback cases are reduced to conditions 2.(a)-2.(c). Moreover, it can be shown that a $\gamma$-suboptimal static control law exists.

The following corollary deals with the regular systems or regular case. It was first reported in Doyle et al [39] and Tadmor [105].

Corollary 3.1.1. Consider the continuous-time linear time-invariant system of (3.1.1) with $D_{22}=0$. Assume that $\Sigma_{\mathrm{P}}$ and $\Sigma_{\mathrm{Q}}$ have no invariant zero on the imaginary axis, $D_{2}$ is of full column rank and $D_{1}$ is of full row rank. Then the following statements are equivalent:

1. There exists a linear time-invariant and proper dynamic compensator $\Sigma_{\mathrm{cmp}}$ of (3.1.2) such that by applying it to (3.1.1) the resulting closed-loop system is internally stable. Moreover, the $H_{\infty}$-norm of the closed-loop transfer function from the disturbance input $w$ to the controlled output $h$ is less than $\gamma$.
2. There exist positive semi-definite matrices $P$ and $Q$ such that the following conditions are satisfied:
(a) $P$ is the solution of the Riccati equation:

$$
\begin{align*}
A^{\prime} P+P A+ & C_{2}^{\prime} C_{2}+\gamma^{2} P E E^{\prime} P \\
& \quad-\left(P B+C_{2}^{\prime} D_{2}\right)\left(D_{2}^{\prime} D_{2}\right)^{-1}\left(B^{\prime} P+D_{2}^{\prime} C_{2}\right)=0 \tag{3.1.9}
\end{align*}
$$

(b) $A_{\text {clp }}$ is asymptotically stable, where

$$
\begin{equation*}
A_{\mathrm{clP}}:=A+\gamma^{-2} E E^{\prime} P-B\left(D_{2}^{\prime} D_{2}\right)^{-1}\left(B^{\prime} P+D_{2}^{\prime} C_{2}\right) \tag{3.1.10}
\end{equation*}
$$

(c) $Q$ is the solution of the Riccati equation:

$$
\begin{align*}
A Q+Q A^{\prime} & +E E^{\prime}+\gamma^{2} Q C_{2}^{\prime} C_{2} Q \\
& \quad-\left(Q C_{1}^{\prime}+E D_{1}^{\prime}\right)\left(D_{1} D_{1}^{\prime}\right)^{-1}\left(C_{1} Q+D_{1} E^{\prime}\right)=0 \tag{3.1.11}
\end{align*}
$$

(d) $A_{\text {clQ }}$ is asymptotically stable, where

$$
\begin{equation*}
A_{\mathrm{clQ}}:=A+\gamma^{-2} Q C_{2}^{\prime} C_{2}-\left(Q C_{1}^{\prime}+E D_{1}^{\prime}\right)\left(D_{1} D_{1}^{\prime}\right)^{-1} C_{1} \tag{3.1.12}
\end{equation*}
$$

(e) $\rho(P Q)<\gamma^{2}$.

If the given system (3.1.1) with nonzero $D_{22}$ term, then the general conditions for the existence of $\gamma$-suboptimal controllers are rather complicated. We will derive these conditions later in Chapter 5. In what follows, we recall a corollary that deals with a special full information feedback case when $D_{2}$ is of full column rank and $\Sigma_{\mathrm{P}}$ has no invariant zero on the imaginary axis.

Corollary 3.1.2. Consider the continuous-time linear time-invariant system of (3.1.1) with $y=\left(x^{\prime} w^{\prime}\right)^{\prime}$ and $D_{2}$ being of full column rank. Assume that $\Sigma_{\mathrm{P}}$ has no invariant zero on the imaginary axis. Then the following statements are equivalent:

1. There exist constant gain matrices $F_{1}$ and $F_{2}$ such that by applying $u=$ $F_{1} x+F_{2} w$ to (3.1.1) the resulting closed-loop system is internally stable. Moreover, the $H_{\infty}$-norm of the closed-loop transfer function from the disturbance input $w$ to the controlled output $h$ is less than $\gamma$.
2. The following conditions are satisfied:
(a) $D_{22}^{\prime}\left(I-D_{2}\left(D_{2}^{\prime} D_{2}\right)^{-1} D_{2}^{\prime}\right) D_{22}<\gamma^{2} I$.
(b) There exists a positive semi-definite solution $P$ to the Riccati equation:

$$
0=P A+A^{\prime} P+C_{2}^{\prime} C_{2}-\left[\begin{array}{c}
B^{\prime} P+D_{2}^{\prime} C_{2} \\
E^{\prime} P+D_{22}^{\prime} C_{2}
\end{array}\right]^{\prime} G^{-1}\left[\begin{array}{c}
B^{\prime} P+D_{2}^{\prime} C \\
E^{\prime} P+D_{22}^{\prime} C
\end{array}\right]
$$

where

$$
G:=\left[\begin{array}{cc}
D_{2}^{\prime} D_{2} & D_{2}^{\prime} D_{22} \\
D_{22}^{\prime} D_{2} & D_{22}^{\prime} D_{22}-\gamma^{2} I
\end{array}\right],
$$

such that the matrix,

$$
A_{\mathrm{clP}}:=A-\left[\begin{array}{ll}
B & E
\end{array}\right] G^{-1}\left[\begin{array}{c}
B^{\prime} P+D_{1}^{\prime} C \\
E^{\prime} P+D_{2}^{\prime} C
\end{array}\right]
$$

is asymptotically stable.
Note that the existence conditions of a $\gamma$-suboptimal controller for the full state feedback case with $D_{2}$ being of full column rank and $\Sigma_{P}$ having no invariant zero on the imaginary axis, are similar to those in item 2 of Corollary 3.1.2 except one has to replace 2.(a) by $D_{22}^{\prime} D_{22}<\gamma^{2} I$.

Next, we will remove the restrictions on the invariant zeros of the subsystems $\Sigma_{\mathrm{P}}$ and $\Sigma_{\mathrm{Q}}$, i.e., we will allow both $\Sigma_{\mathrm{P}}$ and $\Sigma_{\mathrm{Q}}$ to have invariant zeros on the imaginary axis. The following theorem is due to Scherer [96].

Theorem 3.1.2. Consider the continuous-time linear time-invariant system of (3.1.1) with $D_{22}=0$. Then the following statements are equivalent:

1. There exists a linear time-invariant and proper dynamic compensator $\Sigma_{\text {cmp }}$ of (3.1.2) such that by applying it to (3.1.1) the resulting closed-loop system is internally stable. Moreover, the $H_{\infty}$-norm of the closed-loop transfer function from the disturbance input $w$ to the controlled output $h$ is less than $\gamma$.
2. There exist appropriate dimensional constant matrices $F$ and $K$, and positive definite matrices $P>0$ and $Q>0$ such that the following conditions are satisfied:
(a) $(A+B F)^{\prime} P+P(A+B F)+\gamma^{-2} P E E^{\prime} P+\left(C_{2}+D_{2} F\right)^{\prime}\left(C_{2}+D_{2} F\right)<0$.
(b) $\left(A+K C_{1}\right) Q+Q\left(A+K C_{1}\right)^{\prime}+\gamma^{-2} Q C_{2}^{\prime} C_{2} Q+\left(E+K D_{1}\right)\left(E+K D_{1}\right)^{\prime}<0$.
(c) $\rho(P Q)<\gamma^{2}$.

团

The above conditions 2.(a) and 2.(b) in Theorem 3.1.2 can be converted into conditions of the existences of positive definite solutions for some reduced order algebraic Riccati inequalities, which are independent of $F$ and $K$. This can be done by transforming the subsystems $\Sigma_{\mathrm{P}}$ and $\Sigma_{\mathrm{Q}}$ of the given system into the special coordinate basis as in Chapter 2.

### 3.2. Discrete-time Systems

We now consider in this section a general discrete-time linear time-invariant (LTI) system $\Sigma$ with a state-space description
where $x \in \mathbb{R}^{n}$ is the state, $u \in \mathbb{R}^{m}$ is the control input, $w \in \mathbb{R}^{q}$ is the disturbance input, $y \in \mathbb{R}^{p}$ is the measurement output, and $h \in \mathbb{R}^{\ell}$ is the controlled output of $\Sigma$. The following $\Sigma_{\mathrm{cmp}}$ is the controller considered:

$$
\Sigma_{\mathrm{cmp}}:\left\{\begin{align*}
v(k+1) & =A_{\mathrm{cmp}} v+B_{\mathrm{cmp}} y  \tag{3.2.2}\\
u & =C_{\mathrm{cmp}} v+D_{\mathrm{cmp}} y
\end{align*}\right.
$$

Again, as in the continuous-time case, we define $\Sigma_{\mathbf{P}}$ to be the subsystem characterized by the matrix quadruple $\left(A, B, C_{2}, D_{2}\right)$, and $\Sigma_{\mathbf{Q}}$ to be the subsystem characterized by the matrix quadruple $\left(A, E, C_{1}, D_{1}\right)$, which respectively have transfer functions:

$$
\begin{equation*}
G_{\mathrm{P}}(z)=C_{2}(z I-A)^{-1} B+D_{2} \tag{3.2.3}
\end{equation*}
$$

and

$$
\begin{equation*}
G_{\mathrm{Q}}(z)=C_{1}(z I-A)^{-1} E+D_{1} \tag{3.2.4}
\end{equation*}
$$

The following result is due to Stoorvogel, Saberi and Chen [102].
Theorem 3.2.1. Consider the system (3.2.1). Assume that the subsystems $\Sigma_{\mathrm{P}}$ and $\Sigma_{\mathrm{Q}}$ have no invariant zero on the unit circle. Then the following two statements are equivalent:

1. There exists a linear time-invariant and causal dynamic compensator $\Sigma_{\mathrm{cmp}}$ of (3.2.2) such that by applying it to (3.2.1) the resulting closed loop system is internally stable and the closed loop transfer matrix from the disturbance input $w$ to the controlled output $h$ is less than $\gamma$.
2. There exist symmetric matrices $P \geq 0$ and $Q \geq 0$ such that
(a) The following matrix $R$ is positive definite,

$$
\begin{align*}
R:=\gamma^{2} I & -D_{22}^{\prime} D_{22}-E^{\prime} P E \\
& +\left(E^{\prime} P B+D_{22}^{\prime} D_{2}\right) V^{\dagger}\left(B^{\prime} P E+D_{2}^{\prime} D_{22}\right)>0 \tag{3.2.5}
\end{align*}
$$

where

$$
\begin{equation*}
V:=B^{\prime} P B+D_{2}^{\prime} D_{2} \tag{3.2.6}
\end{equation*}
$$

(b) $P$ satisfies the discrete algebraic Riccati equation:

$$
P=A^{\prime} P A+C_{2}^{\prime} C_{2}-\left[\begin{array}{c}
B^{\prime} P A+D_{2}^{\prime} C_{2}  \tag{3.2.7}\\
E^{\prime} P A+D_{22}^{\prime} C_{2}
\end{array}\right]^{\prime} G^{\dagger}\left[\begin{array}{c}
B^{\prime} P A+D_{2}^{\prime} C_{2} \\
E^{\prime} P A+D_{22}^{\prime} C_{2}
\end{array}\right]
$$

where

$$
G:=\left[\begin{array}{cc}
D_{2}^{\prime} D_{2}+B^{\prime} P B & D_{2}^{\prime} D_{22}+B^{\prime} P E  \tag{3.2.8}\\
D_{22}^{\prime} D_{2}+E^{\prime} P B & E^{\prime} P E+D_{22}^{\prime} D_{22}-\gamma^{2} I
\end{array}\right]
$$

(c) For all $z \in \mathbb{C}$ with $|z| \geq 1$, we have

$$
\begin{array}{rcc}
\operatorname{rank}\left[\begin{array}{ccc}
z I-A & -B & -E \\
B^{\prime} P A+D_{2}^{\prime} C_{2} & B^{\prime} P B+D_{2}^{\prime} D_{2} & B^{\prime} P E+D_{2}^{\prime} D_{22} \\
E^{\prime} P A+D_{22}^{\prime} C_{2} & E^{\prime} P B+D_{22}^{\prime} D_{2} & E^{\prime} P E+D_{22}^{\prime} D_{22}-\gamma^{2} I
\end{array}\right] \\
& =n+q+\operatorname{normrank}\left\{G_{\mathrm{P}}(z)\right\} .
\end{array}
$$

(d) The following matrix $S$ is positive definite,

$$
\begin{align*}
S:=\gamma^{2} I & -D_{22} D_{22}^{\prime}-C_{2} Q C_{2}^{\prime} \\
& +\left(C_{2} Q C_{1}^{\prime}+D_{22} D_{1}^{\prime}\right) W^{\dagger}\left(C_{1} Q C_{2}^{\prime}+D_{1} D_{22}^{\prime}\right)>0 \tag{3.2.9}
\end{align*}
$$

where

$$
\begin{equation*}
W:=D_{1} D_{1}^{\prime}+C_{1} Q C_{1}^{\prime} . \tag{3.2.10}
\end{equation*}
$$

(e) $Q$ satisfies the following discrete algebraic Riccati equation:

$$
Q=A Q A^{\prime}+E E^{\prime}-\left[\begin{array}{c}
C_{1} Q A^{\prime}+D_{1} E^{\prime}  \tag{3.2.11}\\
C_{2} Q A^{\prime}+D_{22} E^{\prime}
\end{array}\right]^{\prime} H^{\dagger}\left[\begin{array}{c}
C_{1} Q A^{\prime}+D_{1} E^{\prime} \\
C_{2} Q A^{\prime}+D_{22} E^{\prime}
\end{array}\right]
$$

where

$$
H:=\left[\begin{array}{cc}
D_{1} D_{1}^{\prime}+C_{1} Q C_{1}^{\prime} & D_{1} D_{22}^{\prime}+C_{1} Q C_{2}^{\prime}  \tag{3.2.12}\\
D_{22} D_{1}^{\prime}+C_{2} Q C_{1}^{\prime} & C_{2} Q C_{2}^{\prime}+D_{22} D_{22}^{\prime}-\gamma^{2} I
\end{array}\right]
$$

(f) For all $z \in \mathbb{C}$ with $|z| \geq 1$, we have

$$
\begin{array}{r}
\operatorname{rank}\left[\begin{array}{ccc}
z I-A & A Q C_{1}^{\prime}+E D_{1}^{\prime} & A Q C_{2}^{\prime}+E D_{22}^{\prime} \\
-C_{1} & C_{1} Q C_{1}^{\prime}+D_{1} D_{1}^{\prime} & C_{1} Q C_{2}^{\prime}+D_{1} D_{22}^{\prime} \\
-C_{2} & C_{2} Q C_{1}^{\prime}+D_{22} D_{1}^{\prime} & C_{2} Q C_{2}^{\prime}+D_{22} D_{22}^{\prime}-\gamma^{2} I
\end{array}\right] \\
=n+\ell+\operatorname{normrank}\left\{G_{\mathrm{Q}}(z)\right\} \tag{团}
\end{array}
$$

(g) $\rho(P Q)<\gamma^{2}$.

Here we should note that condition 2.(b) is the standard Riccati equation used in discrete-time $H_{\infty}$ optimization except that the inverse is replaced by a generalized inverse. Condition 2.(c) is nothing other than the requirement that $P$ must be a stabilizing solution of the Riccati equation. Conditions 2.(b) and 2.(c) uniquely determine, if it exists, the matrix $P$. A similar comment can be made about conditions 2.(d)-2.(f). Condition 2.(g) is as usual the coupling condition. The solutions to the above mentioned $P$ and $Q$ can be obtained by transforming the subsystems $\Sigma_{\mathrm{P}}$ and $\Sigma_{\mathrm{Q}}$ into the special coordinate basis as in Chapter 2 and then solving two standard discrete-time Riccati equations without generalized inverses. These will be given later in Chapter 8.

The following remark concerns the full information feedback and full state feedback cases.

Remark 3.2.1. For the special cases of full information and full state feedback we can dispense with the second Riccati equation. More specifically:

1. Full information feedback case: In this case we know both the state and the disturbance of the system at time $k$. It is easy to check that $Q=0$ satisfies conditions 2.(d)-2.(f). Moreover this guarantees that the coupling condition 2 .(g) is automatically satisfied. Therefore there exists a stabilizing controller which yields a closed loop system with the $H_{\infty}$ norm strictly less than $\gamma$ if and only if there exists a positive semi-definite matrix $P$ satisfying conditions 2.(a)-2.(c).
2. Full state feedback case: In this case, it is easy to see that a necessary condition for the existence of a positive semi-definite matrix $Q$ satisfying conditions 2.(d)-2.(f) is that $\left\|D_{22}\right\|<\gamma$. It is also easy to check that for the full state feedback case,

$$
\begin{equation*}
Q=E\left(I-\gamma^{-2} D_{22} D_{22}^{\prime}\right)^{-1} E^{\prime} \tag{3.2.13}
\end{equation*}
$$

satisfies conditions 2.(d)-2.(f). Condition 2.(g) then reduces to

$$
\begin{equation*}
\gamma^{2} I-D_{22} D_{22}^{\prime}-E^{\prime} P E>0 \tag{3.2.14}
\end{equation*}
$$

Moreover, condition (3.2.14) implies that condition 2.(a) is automatically satisfied. Therefore there exists a stabilizing controller which yields a closed loop system with the $H_{\infty}$ norm strictly less than $\gamma$ if and only if there exists a positive semi-definite matrix $P$ satisfying conditions 2.(b), 2.(c) and additionally condition (3.2.14).

Furthermore, it can be shown that either in the full information case or in the full state feedback case, there always exists a $\gamma$-suboptimal static control law whenever the above-mentioned conditions are satisfied.

图

The following corollary deals with the regular case in discrete-time $H_{\infty}$ optimization and is due to Stoorvogel [100].

Corollary 3.2.1. Consider the system (3.2.1). Assume that the subsystem $\Sigma_{\text {P }}$ is left invertible and has no invariant zero on the unit circle, and the subsystem $\Sigma_{\mathrm{Q}}$ is right invertible and has no invariant zero on the unit circle. Then the following two statements are equivalent:

1. There exists a linear time-invariant and causal dynamic compensator $\Sigma_{\mathrm{cmp}}$ of (3.2.2) such that by applying it to (3.2.1) the resulting closed loop system is internally stable and the closed loop transfer matrix from the disturbance input $w$ to the controlled output $h$ is less than $\gamma$.
2. There exist symmetric matrices $P \geq 0$ and $Q \geq 0$ such that
(a) The following matrices $V$ and $R$ are positive definite,

$$
\begin{equation*}
V:=B^{\prime} P B+D_{2}^{\prime} D_{2}>0 \tag{3.2.15}
\end{equation*}
$$

and

$$
\begin{align*}
R:=\gamma^{2} I & -D_{22}^{\prime} D_{22}-E^{\prime} P E \\
& +\left(E^{\prime} P B+D_{22}^{\prime} D_{2}\right) V^{-1}\left(B^{\prime} P E+D_{2}^{\prime} D_{22}\right)>0 \tag{3.2.16}
\end{align*}
$$

(b) $P$ satisfies the discrete algebraic Riccati equation:

$$
P=A^{\prime} P A+C_{2}^{\prime} C_{2}-\left[\begin{array}{c}
B^{\prime} P A+D_{2}^{\prime} C_{2}  \tag{3.2.17}\\
E^{\prime} P A+D_{22}^{\prime} C_{2}
\end{array}\right]^{\prime} G(P)^{-1}\left[\begin{array}{c}
B^{\prime} P A+D_{2}^{\prime} C_{2} \\
E^{\prime} P A+D_{22}^{\prime} C_{2}
\end{array}\right]
$$

where

$$
G(P):=\left[\begin{array}{cc}
D_{2}^{\prime} D_{2}+B^{\prime} P B & D_{2}^{\prime} D_{22}+B^{\prime} P E  \tag{3.2.18}\\
D_{22}^{\prime} D_{2}+E^{\prime} P B & E^{\prime} P E+D_{22}^{\prime} D_{22}-\gamma^{2} I
\end{array}\right]
$$

(c) The following matrix $A_{\text {clp }}$ is asymptotically stable,

$$
A_{\mathrm{clP}}:=A-\left[\begin{array}{ll}
B & E
\end{array}\right] G(P)^{-1}\left[\begin{array}{c}
B^{\prime} P A+D_{2}^{\prime} C_{2}  \tag{3.2.19}\\
E^{\prime} P A+D_{22}^{\prime} C_{2}
\end{array}\right]
$$

(d) The following matrices $W$ and $S$ are positive definite,

$$
\begin{equation*}
W:=D_{1} D_{1}^{\prime}+C_{1} Q C_{1}^{\prime}>0 \tag{3.2.20}
\end{equation*}
$$

and

$$
\begin{align*}
S:=\gamma^{2} I & -D_{22} D_{22}^{\prime}-C_{2} Q C_{2}^{\prime} \\
& +\left(C_{2} Q C_{1}^{\prime}+D_{22} D_{1}^{\prime}\right) W^{-1}\left(C_{1} Q C_{2}^{\prime}+D_{1} D_{22}^{\prime}\right)>0 \tag{3.2.21}
\end{align*}
$$

(e) $Q$ satisfies the following discrete algebraic Riccati equation:

$$
Q=A Q A^{\prime}+E E^{\prime}-\left[\begin{array}{c}
C_{1} Q A^{\prime}+D_{1} E^{\prime}  \tag{3.2.22}\\
C_{2} Q A^{\prime}+D_{22} E^{\prime}
\end{array}\right]^{\prime} H(Q)^{-1}\left[\begin{array}{c}
C_{1} Q A^{\prime}+D_{1} E^{\prime} \\
C_{2} Q A^{\prime}+D_{22} E^{\prime}
\end{array}\right]
$$

where

$$
H(Q):=\left[\begin{array}{cc}
D_{1} D_{1}^{\prime}+C_{1} Q C_{1}^{\prime} & D_{1} D_{22}^{\prime}+C_{1} Q C_{2}^{\prime}  \tag{3.2.23}\\
D_{22} D_{1}^{\prime}+C_{2} Q C_{1}^{\prime} & C_{2} Q C_{2}^{\prime}+D_{22} D_{22}^{\prime}-\gamma^{2} I
\end{array}\right]
$$

(f) The following matrix $A_{\text {clQ }}$ is asymptotically stable,

$$
A_{\mathrm{clQ}}:=A-\left[\begin{array}{c}
C_{1} Q A^{\prime}+D_{1} E^{\prime}  \tag{3.2.24}\\
C_{2} Q A^{\prime}+D_{22} E^{\prime}
\end{array}\right]^{\prime} H(Q)^{-1}\left[\begin{array}{l}
C_{1} \\
C_{2}
\end{array}\right]
$$

(g) $\rho(P Q)<\gamma^{2}$.

It is interesting to note that all the conditions in Corollary 3.2.1 are related to those in Corollary 3.1 .1 by a properly defined bilinear transformation. This will be shown later in Chapter 4. Finally, we conclude this chapter by noting that if $\Sigma_{P}$ or $\Sigma_{Q}$ or both have invariant zeros on the unit circle, one could use the results of the bilinear and inverse bilinear transformations, which are to be presented in Section 4.1 of Chapter 4, and follow Theorem 3.1.2 to derive a similar result.

## Chapter 4

## Bilinear Transformations and Discrete Riccati Equations

IN THIS CHAPTER we will present several preliminary results which are instrumental to the main results dealing with the discrete-time $H_{\infty}$ optimization problems.

In Section 4.1, we will recall a recent result of Chen and Weller [32] on bilinear and inverse bilinear transformations of linear time-invariant systems. Their result presents a comprehensive picture of the mapping of structural properties associated with general linear multivariable systems under bilinear and inverse bilinear transformations. They have completely investigated the problem of how the finite and infinite zero structures, as well as invertibility structures of a general continuous-time (discrete-time) linear time-invariant multivariable system are mapped to those of its discrete-time (continuous-time) counterpart under the bilinear (inverse bilinear) transformation. It is worth noting that we have added in this chapter some new results on the mapping of geometric subspaces under the bilinear (inverse bilinear) transformation.

Section 4.2 recalls from Chen, Saberi and Shamash [30] non-recursive methods for solving the general discrete-time algebraic Riccati equation (DARE) and the discrete-time algebraic Riccati equation related to the $H_{\infty}$ control problem ( $H_{\infty}$-DARE). In particular, they have cast the problem of solving a given $H_{\infty^{-}}$ DARE to the problem of solving an auxiliary continuous-time algebraic Riccati equation associated with the continuous-time $H_{\infty}$ control problem ( $H_{\infty}$-CARE) for which the well known non-recursive solving methods are available. The
advantages of this approach are: it reduces the computation involved in the recursive algorithms while giving much more accurate solutions, and it readily provides the properties of the general $H_{\infty}$-DARE. More importantly, the results given in Section 4.2 build an interconnection between the discrete-time and continuous-time $H_{\infty}$ optimization problems.

The results of Sections 4.1 and 4.2 will be heavily utilized in the development of algorithms for computing infima and solutions to the discrete-time $H_{\infty}$-optimization problems.

### 4.1. Structural Mappings of Bilinear Transformations

The bilinear and inverse bilinear transformations have widespread use in digital control and signal processing. As will be seen shortly, the bilinear transformation is actually playing a crucial role in the computation of infima for discrete-time systems as well as in finding the solutions to discrete-time Riccati equations. The results presented in this section were first reported in Chen and Weller [32]. In fact, the need to perform continuous-time to discrete-time model conversions arises in a range of engineering contexts, including sampled-data control system design, and digital signal processing. As a consequence, numerous discretization procedures exist, including zero- and first-order hold input approximations, impulse invariant transformation, and bilinear transformation (see, for example [2] and [43]). Despite the widespread use of the bilinear transform, however, a comprehensive treatment detailing how key structural properties of continuous-time systems, such as the finite and infinite zero structures, and invertibility properties, are inherited by their discrete-time counterparts is lacking in the literature. Given the important role played by the infinite and finite zero structures in control system design, a clear understanding of the zero structures under bilinear transformation would be useful in the design of sampled-data control systems, and would complement existing results on the mapping of finite and infinite zero structures under zero-order hold sampling (see, for example, [1] and [46]).

In this section, we present a comprehensive study of how the structures, i.e., the finite and infinite zero structures, invertibility structures, as well as geometric subspaces of a general continuous-time (discrete-time) linear timeinvariant system are mapped to those of its discrete-time (continuous-time) counterpart under the well known bilinear (inverse bilinear) transformations

$$
\begin{equation*}
s=a\left(\frac{z-1}{z+1}\right) \quad \text { and } \quad z=\frac{a+s}{a-s} \tag{4.1.1}
\end{equation*}
$$

respectively.

### 4.1.1. Continuous-time to Discrete-time

In this subsection, we will consider a continuous-time linear time-invariant system $\Sigma_{c}$ characterized by

$$
\Sigma_{c}:\left\{\begin{array}{l}
\dot{x}=A x+B u  \tag{4.1.2}\\
y=C x+D u
\end{array}\right.
$$

where $x \in \mathbb{R}^{n}, y \in \mathbb{R}^{p}, u \in \mathbb{R}^{m}$ and $A, B, C$ and $D$ are matrices of appropriate dimensions. Without loss of any generality, we assume that both matrices $\left[\begin{array}{ll}C & D\end{array}\right]$ and $\left[\begin{array}{ll}B^{\prime} & D^{\prime}\end{array}\right]$ are of full rank. $\Sigma_{c}$ has a transfer function

$$
\begin{equation*}
G_{c}(s)=C(s I-A)^{-1} B+D \tag{4.1.3}
\end{equation*}
$$

Let us apply a bilinear transformation to the above continuous-time system, by replacing $s$ in (4.1.3) with

$$
\begin{equation*}
s=\frac{2}{T}\left(\frac{z-1}{z+1}\right)=a\left(\frac{z-1}{z+1}\right) \tag{4.1.4}
\end{equation*}
$$

where $T=2 / a$ is the sampling period. As presented in (4.1.4), the bilinear transformation is often called Tustin's approximation [2], while the choice

$$
\begin{equation*}
a=\frac{\omega_{1}}{\tan \left(\omega_{1} T / 2\right)} \tag{4.1.5}
\end{equation*}
$$

yields the pre-warped Tustin approximation, in which the frequency responses of the continuous-time system and its discrete-time counterpart are matched at frequency $\omega_{1}$. In this way, we obtain a discrete-time system

$$
\begin{equation*}
G_{d}(z)=C\left(a \frac{z-1}{z+1} I-A\right)^{-1} B+D \tag{4.1.6}
\end{equation*}
$$

The following lemma provides a direct state-space realization of $G_{d}(z)$. While this result is well known (see for example [43]), the proof is included as it is brief and self-contained.

Lemma 4.1.1. A state-space realization of $G_{d}(z)$, the discrete-time counterpart of the continuous-time system $\Sigma_{c}$ of (4.1.2) under the bilinear transformation (4.1.4), is given by

$$
\Sigma_{d}:\left\{\begin{align*}
x(k+1) & =\tilde{A} x(k)+\tilde{B} u(k),  \tag{4.1.7}\\
y(k) & =\tilde{C} x(k)+\tilde{D} u(k),
\end{align*}\right.
$$

where

$$
\left.\begin{array}{l}
\tilde{A}=(a I+A)(a I-A)^{-1}  \tag{4.1.8}\\
\tilde{B}=\sqrt{2 a}(a I-A)^{-1} B \\
\tilde{C}=\sqrt{2 a} C(a I-A)^{-1} \\
\tilde{D}=D+C(a I-A)^{-1} B
\end{array}\right\}
$$

or

$$
\left.\begin{array}{l}
\tilde{A}=(a I+A)(a I-A)^{-1}  \tag{4.1.9}\\
\tilde{B}=2 a(a I-A)^{-2} B \\
\tilde{C}=C \\
\tilde{D}=D+C(a I-A)^{-1} B
\end{array}\right\}
$$

or

$$
\left.\begin{array}{l}
\tilde{A}=(a I+A)(a I-A)^{-1}  \tag{4.1.10}\\
\tilde{B}=B \\
\tilde{C}=2 a C(a I-A)^{-2} \\
\tilde{D}=D+C(a I-A)^{-1} B
\end{array}\right\}
$$

Here we clearly assume that matrix $A$ has no eigenvalue at $a$.

Proof. First, it is straightforward to verify that

$$
\begin{align*}
G_{d}(z) & =C\left(a \frac{z-1}{z+1} I-A\right)^{-1} B+D \\
& =(z+1) C[a(z-1) I-(z+1) A]^{-1} B+D \\
& =(z+1) C(a I-A)^{-1}\left[z I-(a I+A)(a I-A)^{-1}\right]^{-1} B+D \\
& =z C(a I-A)^{-1}(z I-\tilde{A})^{-1} B+\left[C(a I-A)^{-1}(z I-\tilde{A})^{-1} B+D\right] \tag{4.1.11}
\end{align*}
$$

If we introduce $\tilde{G}_{d}(z)=z C(a I-A)^{-1}(z I-\tilde{A})^{-1} B$, it follows that

$$
\left\{\begin{align*}
\tilde{x}(k+1) & =\tilde{A}^{\prime} \tilde{x}(k)+\left(a I-A^{\prime}\right)^{-1} C^{\prime} \tilde{u}(k),  \tag{4.1.12}\\
\tilde{y}(k) & =B^{\prime} \tilde{x}(k+1)=B^{\prime} \tilde{A}^{\prime} \tilde{x}(k)+B^{\prime}\left(a I-A^{\prime}\right)^{-1} C^{\prime} \tilde{u}(k),
\end{align*}\right.
$$

is a state-space realization of $\tilde{G}_{d}^{\prime}(z)$, from which

$$
\begin{equation*}
\tilde{G}_{d}(z)=C(a I-A)^{-1}(z I-\tilde{A})^{-1} \tilde{A} B+C(a I-A)^{-1} B \tag{4.1.13}
\end{equation*}
$$

Substituting (4.1.13) into (4.1.11), we obtain

$$
\begin{aligned}
G_{d}(z) & =C(a I-A)^{-1}(z I-\tilde{A})^{-1}(\tilde{A}+I) B+\left[C(a I-A)^{-1} B+D\right] \\
& =\tilde{C}(z I-\tilde{A})^{-1} \tilde{B}+\tilde{D}
\end{aligned}
$$

and the rest of Lemma 4.1.1 follows.

The following theorem establishes the interconnection of the structural properties of $\Sigma_{c}$ and $\Sigma_{d}$, and forms the major contribution of this section.

Theorem 4.1.1. Consider the continuous-time system $\Sigma_{c}$ of (4.1.2) characterized by the quadruple $(A, B, C, D)$ with matrix $A$ having no eigenvalue at $a$, and its discrete-time counterpart under the bilinear transformation (4.1.4), i.e., $\Sigma_{d}$ of (4.1.7) characterized by the quadruple $(\tilde{A}, \tilde{B}, \tilde{C}, \tilde{D})$ of (4.1.8). We have the following properties:

1. Controllability (stabilizability) and observability (detectability) of $\Sigma_{d}$ :
(a) The pair ( $\tilde{A}, \tilde{B})$ is controllable (stabilizable) if and only if the pair $(A, B)$ is controllable (stabilizable).
(b) The pair $(\tilde{A}, \tilde{C})$ is observable (detectable) if and only if the pair ( $A, C$ ) is observable (detectable).
2. Effects of nonsingular state, output and input transformations, together with state feedback and output injection laws:
(a) For any given nonsingular state, output and input transformations $T_{s}, T_{o}$ and $T_{i}$, the quadruple

$$
\begin{equation*}
\left(T_{s}^{-1} \tilde{A} T_{s}, T_{s}^{-1} \tilde{B} T_{i}, T_{o}^{-1} \tilde{C} T_{s}, T_{o}^{-1} \tilde{D} T_{i}\right), \tag{4.1.14}
\end{equation*}
$$

is the discrete-time counterpart under the bilinear transformation (4.1.4), of the continuous time system

$$
\begin{equation*}
\left(T_{s}^{-1} A T_{s}, T_{s}^{-1} B T_{i}, T_{o}^{-1} C T_{s}, T_{o}^{-1} D T_{i}\right) \tag{4.1.15}
\end{equation*}
$$

(b) For any $F \in \mathbb{R}^{m \times n}$ with $A+B F$ having no eigenvalue at $a$, define a nonsingular matrix

$$
\begin{align*}
\tilde{T}_{i} & :=I+F(a I-A-B F)^{-1} B \\
& =\left[I-F(a I-A)^{-1} B\right]^{-1} \in \mathbb{R}^{m \times m} \tag{4.1.16}
\end{align*}
$$

and a constant matrix

$$
\begin{equation*}
\tilde{F}:=\sqrt{2 a} F(a I-A-B F)^{-1} \in \mathbb{R}^{m \times n} . \tag{4.1.17}
\end{equation*}
$$

Then a continuous-time system $\Sigma_{c F}$ characterized by

$$
\begin{equation*}
(A+B F, B, C+D F, D) \tag{4.1.18}
\end{equation*}
$$

is mapped to a discrete-time system $\Sigma_{d \mathbf{F}}$, characterized by

$$
\begin{equation*}
\left(\tilde{A}+\tilde{B} \tilde{F}, \tilde{B} \tilde{T}_{i}, \tilde{C}+\tilde{D} \tilde{F}, \tilde{D} \tilde{T}_{i}\right) \tag{4.1.19}
\end{equation*}
$$

under the bilinear transformation (4.1.4). Here we note that $\Sigma_{c F}$ is the closed-loop system comprising $\Sigma_{c}$ and a state feedback law with gain matrix $F$, and $\Sigma_{d \mathrm{~F}}$ is the closed-loop system comprising $\Sigma_{d}$ and a state feedback law with gain matrix $\tilde{F}$, together with a nonsingular input transformation $\tilde{T}_{i}$.
(c) For any $K \in \mathbb{R}^{n \times p}$ with $A+K C$ having no eigenvalue at $a$, define a nonsingular matrix

$$
\begin{equation*}
\tilde{T}_{o}:=\left[I+C(a I-A-K C)^{-1} K\right]^{-1} \in \mathbb{R}^{p \times p} \tag{4.1.20}
\end{equation*}
$$

and a constant matrix

$$
\begin{equation*}
\tilde{K}:=\sqrt{2 a}(a I-A-K C)^{-1} K \tag{4.1.21}
\end{equation*}
$$

Then a continuous-time system $\Sigma_{c \kappa}$ characterized by

$$
\begin{equation*}
(A+K C, B+K D, C, D) \tag{4.1.22}
\end{equation*}
$$

is mapped to a discrete-time system $\Sigma_{d \mathrm{~K}}$, characterized by

$$
\begin{equation*}
\left(\tilde{A}+\tilde{K} \tilde{C}, \tilde{B}+\tilde{K} \tilde{D}, \tilde{T}_{o}^{-1} \tilde{C}, \tilde{T}_{o}^{-1} \tilde{D}\right) \tag{4.1.23}
\end{equation*}
$$

under the bilinear transformation (4.1.4). We note that $\Sigma_{c K}$ is the closed-loop system comprising $\Sigma_{c}$ and an output injection law with gain matrix $K$, and $\Sigma_{d \kappa}$ is the closed-loop system comprising $\Sigma_{d}$ and an output injection law with gain matrix $\tilde{K}$, together with a nonsingular output transformation $\tilde{T}_{o}$.
3. Invertibility and structural invariant indices lists $\mathcal{I}_{2}$ and $\mathcal{I}_{3}$ of $\Sigma_{d}$ :
(a) $\mathcal{I}_{2}\left(\Sigma_{d}\right)=\mathcal{I}_{2}\left(\Sigma_{c}\right)$, and $\mathcal{I}_{3}\left(\Sigma_{d}\right)=\mathcal{I}_{3}\left(\Sigma_{c}\right)$.
(b) $\Sigma_{d}$ is left (right) invertible if and only if $\Sigma_{c}$ is left (right) invertible.
(c) $\Sigma_{d}$ is invertible (degenerate) if and only if $\Sigma_{c}$ is invertible (degenerate).
4. The invariant zeros of $\Sigma_{d}$ and their associated structures consist of the following two parts:
(a) Let the infinite zero structure (of order greater than 0 ) of $\Sigma_{c}$ be given by $S_{\infty}^{\star}\left(\Sigma_{c}\right)=\left\{q_{1}, q_{2}, \cdots, q_{m_{d}}\right\}$. Then $z=-1$ is an invariant zero of $\Sigma_{d}$ with the multiplicity structure $S_{-1}^{\star}\left(\Sigma_{d}\right)=\left\{q_{1}, q_{2}, \cdots, q_{m_{d}}\right\}$.
(b) Let $s=\alpha \neq a$ be an invariant zero of $\Sigma_{c}$ with the multiplicity structure $S_{\alpha}^{\star}\left(\Sigma_{c}\right)=\left\{n_{\alpha, 1}, n_{\alpha, 2}, \cdots, n_{\alpha, \tau_{\alpha}}\right\}$. Then $z=\beta=(a+\alpha) /(a-\alpha)$ is an invariant zero of $\Sigma_{d}$ with the multiplicity structure $S_{\beta}^{\star}\left(\Sigma_{d}\right)=$ $\left\{n_{\alpha, 1}, n_{\alpha, 2}, \cdots, n_{\alpha, \tau_{\alpha}}\right\}$.
5. The infinite zero structure of $\Sigma_{d}$ consists of the following two parts:
(a) Let $m_{0}$ be the number of infinite zeros of $\Sigma_{c}$ of order 0 , i.e., $m_{0}=$ rank ( $D$ ), and let $m_{d}$ be the total number of infinite zeros of $\Sigma_{c}$ of order greater than 0 . Also, let $\tau_{a}$ be the geometric multiplicity of the invariant zero of $\Sigma_{c}$ at $s=a$. Then the total number of infinite zeros of $\Sigma_{d}$ of order 0, i.e., $\quad \operatorname{rank}(\tilde{D})$, is equal to $m_{0}+m_{d}-\tau_{a}$.
(b) Let $s=a$ be an invariant zero of the given continuous-time system $\Sigma_{c}$ with a multiplicity structure $S_{a}^{\star}\left(\Sigma_{c}\right)=\left\{n_{a, 1}, n_{a, 2}, \cdots, n_{a, \tau_{a}}\right\}$. Then $\Sigma_{d}$ has an infinite zero (of order greater than 0) structure $S_{\infty}^{\star}\left(\Sigma_{d}\right)=$ $\left\{n_{a, 1}, n_{a, 2}, \cdots, n_{a, \tau_{a}}\right\}$.
6. The mappings of geometric subspaces:
(a) $\mathcal{V}^{+}\left(\Sigma_{c}\right)=\mathcal{S}^{\odot}\left(\Sigma_{d}\right)$.
(b) $\mathcal{S}^{+}\left(\Sigma_{c}\right)=\mathcal{V}^{\ominus}\left(\Sigma_{d}\right)$.

Proof. See Subsection 4.1.3.
We have the following two interesting observations. The first is with regard to the minimum phase and nonminimum phase properties of $\Sigma_{d}$, while the second concerns the asymptotic behavior of $\Sigma_{d}$ as the sampling period $T$ tends to zero (or, equivalently, as $a \rightarrow \infty$ ).

Observation 4.1.1. Consider a general continuous-time system $\Sigma_{c}$ and its discrete-time counterpart $\Sigma_{d}$ under the bilinear transformation (4.1.4). Then it follows from 4(a) and 4(b) of Theorem 4.1.1 that

1. $\Sigma_{d}$ has all its invariant zeros inside the unit circle if and only if $\Sigma_{c}$ has all its invariant zeros in the open left-half plane and has no infinite zero of order greater than 0 ;
2. $\Sigma_{d}$ has invariant zeros on the unit circle if and only if $\Sigma_{c}$ has invariant zeros on the imaginary axis, and/or $\Sigma_{c}$ has at least one infinite zero of order greater than 0 ;
3. $\Sigma_{d}$ has invariant zeros outside the unit circle if and only if $\Sigma_{c}$ has invariant zeros in the open right-half plane.

0
Observation 4.1.2. Consider a general continuous-time system $\Sigma_{c}$ and its discrete-time counterpart $\Sigma_{d}$ under the bilinear transformation (4.1.4). Then a consequence of Theorem 4.1.1, $\Sigma_{d}$ has the following asymptotic properties as the sampling period $T$ tends to zero (but not equal to zero):

1. $\Sigma_{d}$ has no infinite zero of order greater than 0 , i.e., no delays from the input to the output;
2. $\Sigma_{d}$ has one invariant zero at $z=-1$ with an appropriate multiplicity structure if $\Sigma_{c}$ has any infinite zero of order greater than 0 ; and
3. The remaining invariant zeros of $\Sigma_{d}$, if any, tend to the point $z=1$. More interestingly, the invariant zeros of $\Sigma_{d}$ corresponding to the stable invariant zeros of $\Sigma_{c}$ are always stable, and approach the point $z=1$ from inside the unit circle. Conversely, the invariant zeros of $\Sigma_{d}$ corresponding to the unstable invariant zeros of $\Sigma_{c}$ are always unstable, and approach the point $z=1$ from outside the unit circle. Finally, those associated with the imaginary axis invariant zeros of $\Sigma_{c}$ are always mapped onto the unit circle and move towards to the point $z=1$.

The following example illustrates the results in Theorem 4.1.1.
Example 4.1.1. Consider a continuous-time system $\Sigma_{c}$ characterized by the quadruple $(A, B, C, D)$ with

$$
A=\left[\begin{array}{llllll}
1 & 1 & 0 & 0 & 1 & 0  \tag{4.1.24}\\
0 & 1 & 1 & 0 & 1 & 0 \\
0 & 0 & 1 & 0 & 1 & 0 \\
0 & 0 & 0 & 3 & 1 & 0 \\
0 & 0 & 0 & 0 & 0 & 1 \\
1 & 1 & 1 & 1 & 1 & 1
\end{array}\right], \quad B=\left[\begin{array}{ll}
0 & 0 \\
0 & 0 \\
0 & 0 \\
1 & 0 \\
0 & 0 \\
0 & 1
\end{array}\right]
$$

and

$$
C=\left[\begin{array}{llllll}
0 & 0 & 0 & 1 & 0 & 0  \tag{4.1.25}\\
0 & 0 & 0 & 0 & 1 & 0
\end{array}\right], \quad D=\left[\begin{array}{ll}
1 & 0 \\
0 & 0
\end{array}\right]
$$

We note that the above system $\Sigma_{c}$ is already in the form of the special coordinate basis as in Theorem 2.3.1. Furthermore, $\Sigma_{c}$ is controllable, observable and invertible with one infinite zero of order 0 , and one infinite zero of order 2, i.e., $S_{\infty}^{\star}\left(\Sigma_{c}\right)=\{2\}$. The system $\Sigma_{c}$ also has two invariant zeros at $s=2$ and $s=1$, respectively, with structures $S_{2}^{\star}\left(\Sigma_{c}\right)=\{1\}$ and $S_{1}^{\star}\left(\Sigma_{c}\right)=\{3\}$.

1. If $a=1$, we obtain a discrete-time system $\Sigma_{d}$ characterized by the quadruple $(\tilde{A}, \tilde{B}, \tilde{C}, \tilde{D})$, with

$$
\begin{array}{cc}
\tilde{A}=\left[\begin{array}{rrrrrr}
1 & 2 & -3 & 1 & 0 & -2 \\
-2 & -1 & 2 & 0 & 0 & 0 \\
0 & -2 & 1 & 0 & 0 & 0 \\
0 & 0 & 1 & -2 & 0 & 0 \\
0 & 0 & -2 & 0 & -1 & 0 \\
0 & 0 & -2 & 0 & -2 & -1
\end{array}\right], \quad \tilde{B}=\frac{\sqrt{2}}{2}\left[\begin{array}{rr}
1 & -2 \\
0 & 0 \\
0 & 0 \\
-1 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right], \\
\tilde{C}=\frac{\sqrt{2}}{2}\left[\begin{array}{rrrrrr}
0 & 0 & 1 & -1 & 0 & 0 \\
0 & 0 & -2 & 0 & 0 & 0
\end{array}\right], \quad \tilde{D}=\frac{1}{2}\left[\begin{array}{ll}
1 & 0 \\
0 & 0
\end{array}\right] .
\end{array}
$$

Utilizing either the toolbox of Chen [9] or that of Lin [60], we find that $\Sigma_{d}$ is indeed controllable, observable and invertible, with one infinite zero of order 0 and one infinite zero of order 3 , i.e., $S_{\infty}^{\star}\left(\Sigma_{d}\right)=\{3\}$. $\Sigma_{d}$ also has two invariant zeros at $z=-3$ and $z=-1$ respectively, with structures $S_{-3}^{\star}\left(\Sigma_{d}\right)=\{1\}$ and $S_{-1}^{\star}\left(\Sigma_{d}\right)=\{2\}$.
2. If $a=2$, we obtain another discrete-time system $\Sigma_{d}$, characterized by

$$
\tilde{A}=\left[\begin{array}{rrrrrr}
0 & -2 & -5 & 3 & -3 & -3 \\
-2 & -1 & -2 & 2 & -2 & -2 \\
-1 & -2 & 0 & 1 & -1 & -1 \\
1 & 2 & 3 & -6 & 1 & 1 \\
-1 & -2 & -3 & 1 & -2 & -1 \\
-2 & -4 & -6 & 2 & -6 & -3
\end{array}\right], \quad \tilde{B}=\frac{1}{2}\left[\begin{array}{rr}
3 & -3 \\
2 & -2 \\
1 & -1 \\
-5 & 1 \\
1 & -1 \\
2 & -2
\end{array}\right]
$$

and

$$
\tilde{C}=\frac{1}{2}\left[\begin{array}{rrrrrr}
1 & 2 & 3 & -5 & 1 & 1 \\
-1 & -2 & -3 & 1 & -1 & -1
\end{array}\right], \quad \tilde{D}=\frac{1}{4}\left[\begin{array}{rr}
-1 & 1 \\
1 & -1
\end{array}\right]
$$

which is controllable, observable and invertible with one infinite zero of order 0 and one infinite zero of order 1, i.e., $S_{\infty}^{\star}\left(\Sigma_{d}\right)=\{1\}$. It also has two invariant zeros at $z=3$ and $z=-1$ respectively, with structures $S_{3}^{\star}\left(\Sigma_{d}\right)=\{3\}$ and $S_{-1}^{\star}\left(\Sigma_{d}\right)=\{2\}$, in accordance with Theorem 4.1.1. 国

### 4.1.2. Discrete-time to Continuous-time

We present in this subsection a similar result as in the previous subsection, but for the inverse bilinear transformation mapping a discrete-time system to a continuous-time system. We begin with a discrete-time linear time-invariant system $\tilde{\Sigma}_{d}$ characterized by

$$
\tilde{\Sigma}_{d}:\left\{\begin{align*}
x(k+1) & =\tilde{A} x(k)+\tilde{B} u(k),  \tag{4.1.26}\\
y(k) & =\tilde{C} x(k)+\tilde{D} u(k)
\end{align*}\right.
$$

where $x \in \mathbb{R}^{n}, y \in \mathbb{R}^{p}, u \in \mathbb{R}^{m}$ and $\tilde{A}, \tilde{B}, \tilde{C}$ and $\tilde{D}$ are matrices of appropriate dimensions. Without loss of any generality, we assume that both matrices $\left[\begin{array}{cc}\tilde{C} & \tilde{D}\end{array}\right]$ and $\left[\begin{array}{cc}\tilde{B}^{\prime} & \tilde{D}^{\prime}\end{array}\right]$ are of full rank. $\Sigma_{d}$ has a transfer function

$$
\begin{equation*}
H_{d}(z)=\tilde{C}(z I-\tilde{A})^{-1} \tilde{B}+\tilde{D} \tag{4.1.27}
\end{equation*}
$$

The inverse bilinear transformation corresponding to (4.1.4) replaces $z$ in the above equation (4.1.27) with

$$
\begin{equation*}
z=\frac{a+s}{a-s} \tag{4.1.28}
\end{equation*}
$$

to obtain the following continuous-time system:

$$
\begin{equation*}
H_{c}(s)=\tilde{C}\left(\frac{a+s}{a-s} I-\tilde{A}\right)^{-1} \tilde{B}+\tilde{D} \tag{4.1.29}
\end{equation*}
$$

The following lemma is analogous to Lemma 4.1.1, and provides a statespace realization of $H_{c}(s)$.

Lemma 4.1.2. A state-space realization of $H_{c}(s)$, the continuous-time counterpart of the discrete-time system $\tilde{\Sigma}_{d}$ of (4.1.26) under the inverse bilinear transformation (4.1.28), is given by

$$
\tilde{\Sigma}_{c}:\left\{\begin{array}{l}
\dot{x}=A x+B u  \tag{4.1.30}\\
y=C x+D u
\end{array}\right.
$$

where

$$
\left.\begin{array}{l}
A=a(\tilde{A}+I)^{-1}(\tilde{A}-I)  \tag{4.1.31}\\
B=\sqrt{2 a}(\tilde{A}+I)^{-1} \tilde{B} \\
C=\sqrt{2 a} \tilde{C}(\tilde{A}+I)^{-1} \\
D=\tilde{D}-\tilde{C}(\tilde{A}+I)^{-1} \tilde{B}
\end{array}\right\}
$$

or

$$
\left.\begin{array}{l}
A=a(\tilde{A}+I)^{-1}(\tilde{A}-I)  \tag{4.1.32}\\
B=2 a(\tilde{A}+I)^{-2} \tilde{B} \\
C=\tilde{C} \\
D=\tilde{D}-\tilde{C}(\tilde{A}+I)^{-1} \tilde{B}
\end{array}\right\}
$$

$$
\left.\begin{array}{l}
A=a(\tilde{A}+I)^{-1}(\tilde{A}-I)  \tag{4.1.33}\\
B=\tilde{B} \\
C=2 a \tilde{C}(\tilde{A}+I)^{-2} \\
D=\tilde{D}-\tilde{C}(\tilde{A}+I)^{-1} \tilde{B}
\end{array}\right\}
$$

Here we clearly assume that the matrix $\tilde{A}$ has no eigenvalue at -1 .
The following theorem is analogous to Theorem 4.1.1.

Theorem 4.1.2. Consider the discrete-time system $\tilde{\Sigma}_{d}$ of (4.1.26) characterized by the quadruple ( $\tilde{A}, \tilde{B}, \tilde{C}, \tilde{D}$ ) with matrix $\tilde{A}$ having no eigenvalue at -1 , and its continuous-time counterpart under the inverse bilinear transformation (4.1.28), i.e., $\tilde{\Sigma}_{c}$ of (4.1.30) characterized by the quadruple $(A, B, C, D)$ of (4.1.31). We have the following properties:

1. Controllability (stabilizability) and observability (detectability) of $\tilde{\Sigma}_{c}$ :
(a) The pair $(A, B)$ is controllable (stabilizable) if and only if the pair ( $\tilde{A}, \tilde{B}$ ) is controllable (stabilizable).
(b) The pair $(A, C)$ is observable (detectable) if and only if the pair ( $\tilde{A}, \tilde{C})$ is observable (detectable).
2. Effects of nonsingular state, output and input transformations, together with state feedback and output injection laws:
(a) For any given nonsingular state, output and input transformations $T_{s}, T_{o}$ and $T_{i}$, the quadruple

$$
\begin{equation*}
\left(T_{s}^{-1} A T_{s}, T_{s}^{-1} B T_{i}, T_{o}^{-1} C T_{s}, T_{o}^{-1} D T_{i}\right) \tag{4.1.34}
\end{equation*}
$$

is the continuous-time counterpart of the inverse bilinear transformation, i.e., (4.1.28), of the discrete-time system

$$
\begin{equation*}
\left(T_{s}^{-1} \tilde{A} T_{s}, T_{s}^{-1} \tilde{B} T_{i}, T_{o}^{-1} \tilde{C} T_{s}, T_{o}^{-1} \tilde{D} T_{i}\right) \tag{4.1.35}
\end{equation*}
$$

(b) For any $\tilde{F} \in \mathbb{R}^{m \times n}$ with $\tilde{A}+\tilde{B} \tilde{F}$ having no eigenvalue at -1 , define a nonsingular matrix

$$
\begin{equation*}
T_{i}:=I-\tilde{F}(I+\tilde{A}+\tilde{B} \tilde{F})^{-1} \tilde{B} \in \mathbb{R}^{m \times m} \tag{4.1.36}
\end{equation*}
$$

and a constant matrix

$$
\begin{equation*}
F:=\sqrt{2 a} \tilde{F}(I+\tilde{A}+\tilde{B} \tilde{F})^{-1} \in \mathbb{R}^{m \times n} \tag{4.1.37}
\end{equation*}
$$

Then a discrete-time system $\tilde{\Sigma}_{d F}$ characterized by

$$
\begin{equation*}
(\tilde{A}+\tilde{B} \tilde{F}, \tilde{B}, \tilde{C}+\tilde{D} \tilde{F}, \tilde{D}) \tag{4.1.38}
\end{equation*}
$$

is mapped to a continuous-time counterpart $\tilde{\Sigma}_{c \mathrm{~F}}$ characterized by

$$
\begin{equation*}
\left(A+B F, B T_{i}, C+D F, D T_{i}\right) \tag{4.1.39}
\end{equation*}
$$

under the inverse bilinear transformation (4.1.28). Note that $\tilde{\Sigma}_{d F}$ is the closed-loop system comprising $\tilde{\Sigma}_{d}$ and a state feedback law with
gain matrix $\tilde{F}$, and $\tilde{\Sigma}_{d F}$ is the closed-loop system comprising $\tilde{\Sigma}_{d}$ and a state feedback law with gain matrix $F$, together with a nonsingular input transformation $T_{i}$.
(c) For any $\tilde{K} \in \mathbf{R}^{n \times p}$ with $\tilde{A}+\tilde{K} \tilde{C}$ having no eigenvalue at -1 , define a nonsingular matrix

$$
\begin{equation*}
T_{o}:=\left[I-\tilde{C}(I+\tilde{A}+\tilde{K} \tilde{C})^{-1} \tilde{K}\right]^{-1} \in \mathbb{R}^{p \times p} \tag{4.1.40}
\end{equation*}
$$

and a constant matrix

$$
\begin{equation*}
K:=\sqrt{2 a}(I+\tilde{A}+\tilde{K} \tilde{C})^{-1} \tilde{K} \tag{4.1.41}
\end{equation*}
$$

Then a discrete-time system $\tilde{\Sigma}_{d \mathrm{~K}}$ characterized by

$$
\begin{equation*}
(\tilde{A}+\tilde{K} \tilde{C}, \tilde{B}+\tilde{K} \tilde{D}, \tilde{C}, \tilde{D}) \tag{4.1.42}
\end{equation*}
$$

is mapped to a continuous-time $\tilde{\Sigma}_{c K}$, characterized by

$$
\begin{equation*}
\left(A+K C, B+K D, T_{o}^{-1} C, T_{o}^{-1} D\right) \tag{4.1.43}
\end{equation*}
$$

under the inverse bilinear transformation (4.1.28). We note that $\tilde{\Sigma}_{d K}$ is the closed-loop system comprising $\tilde{\Sigma}_{d}$ and an output injection law with gain matrix $\tilde{K}$, and $\tilde{\Sigma}_{c k}$ is the closed-loop system comprising $\tilde{\Sigma}_{c}$ and an output injection law with gain matrix $K$, together with a nonsingular output transformation $T_{o}$.
3. Invertibility and structural invariant indices lists $\mathcal{I}_{2}$ and $\mathcal{I}_{3}$ of $\tilde{\Sigma}_{c}$ :
(a) $\mathcal{I}_{2}\left(\tilde{\Sigma}_{c}\right)=\mathcal{I}_{2}\left(\tilde{\Sigma}_{d}\right)$, and $\mathcal{I}_{3}\left(\tilde{\Sigma}_{c}\right)=\mathcal{I}_{3}\left(\tilde{\Sigma}_{d}\right)$.
(b) $\tilde{\Sigma}_{c}$ is left (right) invertible if and only if $\tilde{\Sigma}_{d}$ is left (right) invertible.
(c) $\tilde{\Sigma}_{c}$ is invertible (degenerate) if and only if $\tilde{\Sigma}_{d}$ is invertible (degenerate).
4. Invariant zeros of $\tilde{\Sigma}_{c}$ and their structures consist of the following two parts:
(a) Let the infinite zero structure (of order greater than 0 ) of $\tilde{\Sigma}_{d}$ be given by $S_{\infty}^{\star}\left(\tilde{\Sigma}_{d}\right)=\left\{q_{1}, q_{2}, \cdots, q_{m_{d}}\right\}$. Then $s=a$ is an invariant zero of $\tilde{\Sigma}_{c}$ with the multiplicity structure $S_{a}^{\star}\left(\tilde{\Sigma}_{c}\right)=\left\{q_{1}, q_{2}, \cdots, q_{m_{d}}\right\}$.
(b) Let $z=\alpha \neq-1$ be an invariant zero of $\tilde{\Sigma}_{d}$ with the multiplicity structure $S_{\alpha}^{\star}\left(\tilde{\Sigma}_{d}\right)=\left\{n_{\alpha, 1}, n_{\alpha, 2}, \cdots, n_{\alpha, \tau_{\alpha}}\right\}$. Then $s=\beta=a \frac{\alpha-1}{\alpha+1}$ is an invariant zero of $\tilde{\Sigma}_{c}$ with the multiplicity structure $S_{\beta}^{\star}\left(\tilde{\Sigma}_{c}\right)=$ $\left\{n_{\alpha, 1}, n_{\alpha, 2}, \cdots, n_{\alpha, \tau_{\alpha}}\right\}$.
5. The infinite zero structure of $\tilde{\Sigma}_{c}$ consists of the following two parts:
(a) Let $m_{0}$ be the number of infinite zeros of $\tilde{\Sigma}_{d}$ of order 0 , i.e., $m_{0}=$ rank $(\tilde{D})$, and let $m_{d}$ be the total number of infinite zeros of $\tilde{\Sigma}_{d}$ of order greater than 0 . Also, let $\tau_{-1}$ be the geometric multiplicity of the invariant zero of $\tilde{\Sigma}_{d}$ at $z=-1$. Then the total number of infinite zeros of $\tilde{\Sigma}_{c}$ of order 0 , i.e., $\operatorname{rank}(D)$, is equal to $m_{0}+m_{d}-\tau_{-1}$.
(b) Let $z=-1$ be an invariant zero of the given discrete-time system $\tilde{\Sigma}_{d}$ with the multiplicity structure $S_{-1}^{\star}\left(\tilde{\Sigma}_{d}\right)=\left\{n_{-1,1}, n_{-1,2}, \cdots, n_{-1, \tau_{-1}}\right\}$. Then $\tilde{\Sigma}_{c}$ has an infinite zero (of order greater than 0 ) structure $S_{\infty}^{\star}\left(\tilde{\Sigma}_{c}\right)=\left\{n_{-1,1}, n_{-1,2}, \cdots, n_{-1, \tau_{-1}}\right\}$.
6. The mappings of geometric subspaces:
(a) $\mathcal{V}^{\odot}\left(\tilde{\Sigma}_{d}\right)=\mathcal{S}^{+}\left(\tilde{\Sigma}_{c}\right)$.
(b) $\mathcal{S}^{\odot}\left(\tilde{\Sigma}_{d}\right)=\mathcal{V}^{+}\left(\tilde{\Sigma}_{c}\right)$.

Proof. The proof of this theorem is similar to that of Theorem 4.1.1.

We illustrate the result above with the following example.
Example 4.1.2. Consider a discrete-time linear time-invariant system $\tilde{\Sigma}_{d}$ characterized by the quadruple $(\tilde{A}, \tilde{B}, \tilde{C}, \tilde{D})$ with

$$
\tilde{A}=\left[\begin{array}{rrrrrrr}
-1 & 0 & 0 & 1 & 0 & 1 & 0  \tag{4.1.44}\\
0 & -1 & 1 & 1 & 0 & 1 & 1 \\
0 & 0 & -1 & 1 & 0 & 1 & 1 \\
0 & 0 & 0 & 1 & 0 & 1 & 1 \\
1 & 1 & 1 & 1 & 1 & 1 & 1 \\
1 & 1 & 1 & 1 & 1 & 1 & 1 \\
0 & 1 & 1 & 1 & 1 & 1 & 1
\end{array}\right], \quad \tilde{B}=\left[\begin{array}{lll}
0 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 1 \\
1 & 0 & 0 \\
0 & 1 & 0
\end{array}\right],
$$

and

$$
\tilde{C}=\left[\begin{array}{lllllll}
0 & 0 & 0 & 1 & 0 & 0 & 0  \tag{4.1.45}\\
0 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 1
\end{array}\right], \quad \tilde{D}=\left[\begin{array}{lll}
0 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right]
$$

Again the above system is already in the form of the special coordinate basis. It is simple to verify that $\tilde{\Sigma}_{d}$ is controllable, observable and is degenerate, i.e., neither left nor right invertible, with two infinite zeros of order 1, i.e., $S_{\infty}^{\star}\left(\tilde{\Sigma}_{d}\right)=\{1,1\}, \mathcal{I}_{2}\left(\tilde{\Sigma}_{d}\right)=\{1\}$ and $\mathcal{I}_{3}\left(\tilde{\Sigma}_{d}\right)=\{1\}$. It also has one invariant zero at $z=-1$ with a structure $S_{-1}^{\star}\left(\tilde{\Sigma}_{d}\right)=\{1,2\}$. Applying the result in

Lemma 4.1.2 (with $a=1$ ), we obtain $\tilde{\Sigma}_{c}$ which is characterized by $(A, B, C, D)$ with

$$
A=\left[\begin{array}{rrrrrrr}
5 & 0 & 0 & -2 & 0 & -2 & 2 \\
0 & 3 & 4 & -2 & 2 & -2 & -2 \\
0 & -2 & 3 & 0 & 0 & 0 & 0 \\
0 & 0 & 2 & -1 & 0 & 0 & 0 \\
-2 & 0 & -2 & 2 & -1 & 2 & 0 \\
-2 & 0 & -2 & 2 & 0 & 1 & 0 \\
2 & 0 & -2 & 0 & 0 & 0 & 1
\end{array}\right], \quad B=\sqrt{2}\left[\begin{array}{rrr}
1 & -1 & 0 \\
1 & 1 & -1 \\
0 & 0 & 0 \\
0 & 0 & 0 \\
-1 & 0 & 1 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right]
$$

and

$$
C=\sqrt{2}\left[\begin{array}{rrrrrrr}
0 & 0 & -1 & 1 & 0 & 0 & 0 \\
1 & 0 & 1 & -1 & 0 & 0 & 0 \\
-1 & 0 & 1 & 0 & 0 & 0 & 0
\end{array}\right], \quad D=\left[\begin{array}{lll}
0 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right]
$$

Then, it is straightforward to verify, using the software toolboxes of Chen [9] or $\operatorname{Lin}$ [60], for example, that $\tilde{\Sigma}_{c}$ is controllable, observable and degenerate with an infinite zero structure $S_{\infty}^{\star}\left(\tilde{\Sigma}_{c}\right)=\{1,2\}, \mathcal{I}_{2}\left(\tilde{\Sigma}_{c}\right)=\{1\}$ and $\mathcal{I}_{3}\left(\tilde{\Sigma}_{c}\right)=\{1\}$. Furthermore, $\tilde{\Sigma}_{c}$ has one invariant zero at $s=1$ with associated structure $S_{1}^{\star}\left(\tilde{\Sigma}_{c}\right)=\{1,1\}$, in accordance with Theorem 4.1.2.

Finally, we conclude this subsection by summarizing in graphical forms in Figures 4.1 .1 the structural mappings associated with the bilinear and inverse bilinear transformations.

### 4.1.3. Proof of Theorem 4.1.1

We present in this subsection the detailed proof of Theorem 4.1.1. For the sake of simplicity of presentation, and without loss of any generality, we assume that the constant $a$ in (4.1.4) is equal to unity, i.e., $a=2 / T=1$, throughout this proof. We will prove this theorem item-by-item.
$1(\mathrm{a})$. Let $\beta$ be an eigenvalue of $\tilde{A}$, i.e., $\beta \in \lambda(\tilde{A})$. It is straightforward to verify that $\beta \neq-1$, provided $A$ has no eigenvalue at $a=1$ and $\alpha=(\beta-1) /(\beta+1)$ is an eigenvalue of $A$, i.e., $\alpha \in \lambda(A)$. Next, we consider the matrix pencil

$$
\left.\left.\begin{array}{rl}
{\left[\begin{array}{lll}
\beta I-\tilde{A} & \tilde{B}
\end{array}\right]} & =\left[\begin{array}{ll}
\beta I-(I-A)^{-1}(I+A) & \sqrt{2}(I-A)^{-1} B
\end{array}\right] \\
& =(I-A)^{-1}[\beta(I-A)-(I+A) \\
\sqrt{2} B
\end{array}\right] .\right] .\left[\begin{array}{cc}
(\beta+1) I_{n} & 0 \\
0 & \sqrt{2} I_{m}
\end{array}\right] .
$$

Clearly, $\operatorname{rank}\left[\begin{array}{ll}\beta I-\tilde{A} & \tilde{B}\end{array}\right]=\operatorname{rank}\left[\begin{array}{ll}\alpha I-A & B\end{array}\right]$, and the result 1 (a) follows. 区


Figure 4.1.1: Structural mappings of bilinear bilinear transformations.

1(b). Dual of 1(a).
囚
2(a). It is trivial.
囚
2(b). It follows from Lemma 4.1.1 that the discrete-time counterpart $\Sigma_{d F}$ of the bilinear transformation of $\Sigma_{c F}$, characterized by $(A+B F, B, C+D F, D)$, is given by $\left(\tilde{A}_{\mathrm{F}}, \tilde{B}_{\mathrm{F}}, \tilde{C}_{\mathrm{F}}, \tilde{D}_{\mathrm{F}}\right)$ with

$$
\left.\begin{array}{l}
\tilde{A}_{\mathrm{F}}=(I+A+B F)(I-A-B F)^{-1}  \tag{4.1.46}\\
\tilde{B}_{\mathrm{F}}=\sqrt{2}(I-A-B F)^{-1} B \\
\tilde{C}_{\mathrm{F}}=\sqrt{2}(C+D F)(I-A-B F)^{-1} \\
\tilde{D}_{\mathrm{F}}=D+(C+D F)(I-A-B F)^{-1} B
\end{array}\right\}
$$

We first recall from the Appendix of Kailath [48] the following matrix identities that are frequently used in the derivation of our result:

$$
\begin{equation*}
(I+X Y)^{-1} X=X(I+Y X)^{-1} \tag{4.1.47}
\end{equation*}
$$

and

$$
\begin{equation*}
\left[I+X(s I-Z)^{-1} Y\right]^{-1}=I-X(s I-Z+Y X)^{-1} Y \tag{4.1.48}
\end{equation*}
$$

Next, we note that

$$
\begin{aligned}
\tilde{A}_{\mathrm{F}} & =(I+A+B F)(I-A-B F)^{-1} \\
& =(I+A+B F)(I-A)^{-1}\left[I-B F(I-A)^{-1}\right]^{-1} \\
& =\left[\tilde{A}+B F(I-A)^{-1}\right]\left[I-B F(I-A)^{-1}\right]^{-1} \\
& =\left[\tilde{A}+B F(I-A)^{-1}\right]\left[I+B F(I-A-B F)^{-1}\right] \\
& =\tilde{A}+\tilde{A} B F(I-A-B F)^{-1}+B F(I-A)^{-1}\left[I+B F(I-A-B F)^{-1}\right] \\
& =\tilde{A}+\tilde{A} B F(I-A-B F)^{-1}+B F(I-A)^{-1}(I-A)(I-A-B F)^{-1} \\
& =\tilde{A}+\tilde{A} B F(I-A-B F)^{-1}+B F(I-A-B F)^{-1} \\
& =\tilde{A}+(\tilde{A}+I) B F(I-A-B F)^{-1} \\
& =\tilde{A}+2(I-A)^{-1} B F(I-A-B F)^{-1} \\
& =\tilde{A}+\tilde{B} \tilde{F}
\end{aligned}
$$

and

$$
\begin{aligned}
\tilde{B}_{\mathrm{F}} & =\sqrt{2}(I-A-B F)^{-1} B \\
& =\sqrt{2}\left[I-(I-A)^{-1} B F\right]^{-1}(I-A)^{-1} B \\
& =\sqrt{2}(I-A)^{-1} B\left[I-F(I-A)^{-1} B\right]^{-1} \\
& =\tilde{B} \tilde{T}_{i} .
\end{aligned}
$$

Also, we have

$$
\begin{aligned}
\tilde{C}_{\mathrm{F}}= & \sqrt{2}(C+D F)(I-A-B F)^{-1} \\
= & \sqrt{2}(C+D F)(I-A)^{-1}\left[I-B F(I-A)^{-1}\right]^{-1} \\
= & \sqrt{2}(C+D F)(I-A)^{-1}\left[I+B F(I-A-B F)^{-1}\right] \\
= & \sqrt{2} C(I-A)^{-1}+\sqrt{2} D F(I-A)^{-1} \\
& +\sqrt{2}(C+D F)(I-A)^{-1} B F(I-A-B F)^{-1} \\
= & \tilde{C}+\sqrt{2}\left[D F(I-A)^{-1}(I-A-B F)\right. \\
& \left.+(C+D F)(I-A)^{-1} B F\right](I-A-B F)^{-1} \\
= & \tilde{C}+\sqrt{2}\left[D F-D F(I-A)^{-1} B F+C(I-A)^{-1} B F+D F(I-A)^{-1} B F\right] \\
& \times(I-A-B F)^{-1} \\
= & \tilde{C}+\left[D+C(I-A)^{-1} B\right] \sqrt{2} F(I-A-B F)^{-1} \\
= & \tilde{C}+\tilde{D} \tilde{F},
\end{aligned}
$$

and

$$
\begin{aligned}
\tilde{D}_{\mathrm{F}} & =D+(C+D F)(I-A-B F)^{-1} B \\
& =D+(C+D F)\left[I-(I-A)^{-1} B F\right]^{-1}(I-A)^{-1} B \\
& =D+(C+D F)(I-A)^{-1} B\left[I-F(I-A)^{-1} B\right]^{-1} \\
& =\left\{D\left[I-F(I-A)^{-1} B\right]+(C+D F)(I-A)^{-1} B\right\} \tilde{T}_{i} \\
& =\left\{D-D F(I-A)^{-1} B+C(I-A)^{-1} B+D F(I-A)^{-1} B\right\} \tilde{T}_{i} \\
& =\tilde{D} \tilde{T}_{i},
\end{aligned}
$$

which completes the proof of 2 (b).
2(c). Dual of 2(b).
With the benefit of properties of $2(\mathrm{a})-2(\mathrm{c})$, the remainder of the proof is considerably simplified. It is well known that the structural invariant indices lists of Morse, which correspond precisely to the structures of finite and infinite zeros as well as invertibility, are invariant under nonsingular state, output and input transformations, state feedback laws and output injections. We can thus apply appropriate nonsingular state, output and input transformations, as well as state feedback and output injection, to $\Sigma_{c}$ and so obtain a new system, say $\Sigma_{c}^{*}$. If this new system has $\Sigma_{d}^{*}$ as its discrete-time counterpart under bilinear transformation, then from properties 2(a)-2(c), it follows that $\Sigma_{d}^{*}$ and $\Sigma_{d}$ have the same structural invariant properties. It is therefore sufficient for the remainder of the proof that we show $3(\mathrm{a})-5(\mathrm{~b})$ are properties of $\Sigma_{d}^{*}$.

Let us first apply nonsingular state, output and input transformations $\Gamma_{s}$, $\Gamma_{o}$ and $\Gamma_{i}$ to $\Sigma_{c}$ such that the resulting system is in the form of the special coordinate basis as in Theorem 2.3.1, or, equivalently, the compact form in (2.3.20)-(2.3.23) with $A_{a a}$ and $C_{0 a}$ being given by (2.3.25), $E_{d a}$ and $E_{c a}$ being given by (2.3.26), and $B_{0 a}, L_{a b}$ and $L_{a d}$ being given by (2.3.28). We will further assume that $A_{a a}$ is already in the Jordan form of (2.1.1) and (2.3.32), and that matrices $A_{a a}, L_{a d}, B_{a 0}, E_{d a}, C_{0 a}, E_{c a}$ and $L_{a b}$ are partitioned as follows:

$$
\begin{gather*}
A_{a a}=\left[\begin{array}{cc}
A_{a a}^{a} & 0 \\
0 & A_{a a}^{*}
\end{array}\right], L_{a d}=\left[\begin{array}{c}
L_{a d}^{a} \\
L_{a d}^{*}
\end{array}\right], B_{a 0}=\left[\begin{array}{c}
B_{a 0}^{a} \\
B_{a 0}^{*}
\end{array}\right], L_{a b}=\left[\begin{array}{c}
L_{a b}^{a} \\
L_{a b}^{*}
\end{array}\right],  \tag{4.1.49}\\
E_{d a}=\left[\begin{array}{ll}
E_{d a}^{a} & E_{d a}^{*}
\end{array}\right], \quad C_{0 a}=\left[\begin{array}{ll}
C_{0 a}^{a} & C_{0 a}^{*}
\end{array}\right], \quad E_{c a}=\left[\begin{array}{ll}
E_{c a}^{a} & E_{c a}^{*}
\end{array}\right] \tag{4.1.50}
\end{gather*}
$$

where matrix $A_{a a}^{a}$ has all its eigenvalues at $a=1$, i.e.,

$$
A_{a a}^{a}=I+\left[\begin{array}{ccccc}
0 & I_{n_{a, 1}-1} & \cdots & 0 & 0  \tag{4.1.51}\\
0 & 0 & \cdots & 0 & 0 \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & 0 & I_{n_{a, \tau_{a}-1}} \\
0 & 0 & \cdots & 0 & 0
\end{array}\right]
$$

and $A_{a a}^{*}$ contains the remaining invariant zeros of $\Sigma_{c}$. Furthermore, we assume that the pair $\left(A_{c c}, B_{c}\right)$ is in the Brunovsky canonical form of (2.3.37), as is the pair $\left(A_{b b}^{\prime}, C_{b}^{\prime}\right)$. Next, define a state feedback gain matrix

$$
F=-\Gamma_{i}\left[\begin{array}{ccccc}
C_{0 a}^{a}-C_{2}^{a} & C_{0 a}^{*} & C_{0 b} & C_{0 c} & C_{0 d}  \tag{4.1.52}\\
E_{d a}^{a}-C_{1}^{a} & E_{d a}^{*} & E_{d b} & E_{d c} & E_{d d} \\
E_{c a}^{a} & E_{c a}^{*} & 0 & E_{c c} & 0
\end{array}\right] \Gamma_{s}^{-1}
$$

and an output injection gain matrix

$$
K=-\Gamma_{s}\left[\begin{array}{ccc}
B_{a 0}^{a}-B_{2}^{a} & L_{a d}^{a}-B_{1}^{a} & L_{a b}^{a}  \tag{4.1.53}\\
B_{a 0}^{*} & L_{a d}^{*} & L_{a b}^{*} \\
B_{b 0} & L_{b d} & L_{b b} \\
B_{c 0} & L_{c d} & L_{c b} \\
B_{d 0} & L_{d d} & 0
\end{array}\right] \Gamma_{o}^{-1}
$$

Here, $E_{c c}$ is chosen such that all $\star$ 's in (2.3.37) are cleaned out, i.e.,

$$
\begin{equation*}
A_{c c}^{*}:=A_{c c}-B_{c} E_{c c}, \tag{4.1.54}
\end{equation*}
$$

is in Jordan form with all diagonal elements equal to 0 . Similarly, $L_{b b}$ is chosen such that

$$
\begin{equation*}
\left(A_{b b}^{*}\right)^{\prime}:=\left(A_{b b}-L_{b b} C_{b}\right)^{\prime} \tag{4.1.55}
\end{equation*}
$$

is in Jordan form with with all diagonal elements equal to 0 . Likewise, $E_{d d}$ and $L_{d d}$ are chosen such that

$$
\begin{equation*}
A_{d d}^{*}:=A_{d d}-L_{d d} C_{d}-B_{d} E_{d d}, \tag{4.1.56}
\end{equation*}
$$

is in Jordan form with all diagonal elements equal to 0 , which in turn implies

$$
\begin{equation*}
C_{d}\left(I-A_{d d}^{*}\right)^{-1} B_{d}=I_{m_{d}} . \tag{4.1.57}
\end{equation*}
$$

The matrices $B_{1}^{a}, B_{2}^{a}, C_{1}^{a}$ and $C_{2}^{a}$ are chosen in conformity with $A_{a a}^{a}$ of (4.1.51) as follows:

$$
B^{a}:=\left[\begin{array}{ll}
B_{2}^{a} & B_{1}^{a}
\end{array}\right]:=\left[\begin{array}{cccc}
0 & 0 & \cdots & 0  \tag{4.1.58}\\
0 & 1 & \cdots & 0 \\
\vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & 0 \\
0 & 0 & \cdots & 1
\end{array}\right]
$$

and

$$
C^{a}:=\left[\begin{array}{l}
C_{2}^{a}  \tag{4.1.59}\\
C_{1}^{a}
\end{array}\right]:=\left[\begin{array}{ccccc}
0 & 0 & \cdots & 0 & 0 \\
1 & 0 & \cdots & 0 & 0 \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & 1 & 0
\end{array}\right]
$$

This can always be done, as a consequence of the assumption that the matrix $A$ has no eigenvalue at $a=1$, which implies that the invariant zero at $a=1$ of $\Sigma_{c}$ is completely controllable and observable.

Finally, we obtain a continuous-time system $\Sigma_{c}^{*}$ characterized by the quadruple $\left(A^{*}, B^{*}, C^{*}, D^{*}\right)$, where

$$
\begin{gather*}
A^{*}=P^{-1} \Gamma_{s}^{-1}(A+B F+K C+K D F) \Gamma_{s} P \\
=\left[\begin{array}{ccccc}
A_{a a}^{*} & 0 & 0 & 0 & 0 \\
0 & A_{b b}^{*} & 0 & 0 & 0 \\
0 & 0 & A_{c c}^{*} & 0 & 0 \\
0 & 0 & 0 & A_{d d}^{*} & B_{d} C_{a}^{a} \\
0 & 0 & 0 & B_{1}^{a} C_{d} & A_{a a}^{a}+B_{2}^{a} C_{2}^{a}
\end{array}\right]  \tag{4.1.60}\\
B^{*}=P^{-1} \Gamma_{s}^{-1}(B+K D) \Gamma_{i}=\left[\begin{array}{ccc}
0 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & B_{c} \\
0 & B_{d} & 0 \\
B_{2}^{a} & 0 & 0
\end{array}\right]  \tag{4.1.61}\\
\quad C^{*}=\Gamma_{o}^{-1}(C+D F) \Gamma_{s} P=\left[\begin{array}{ccccc}
0 & 0 & 0 & 0 & C_{2}^{a} \\
0 & 0 & 0 & C_{d} & 0 \\
0 & C_{b} & 0 & 0 & 0
\end{array}\right] \tag{4.1.62}
\end{gather*}
$$

and

$$
D^{*}=\Gamma_{o}^{-1} D \Gamma_{i}=\left[\begin{array}{ccc}
I_{m_{0}} & 0 & 0  \tag{4.1.63}\\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right]
$$

where $P$ is a permutation matrix that transforms $A_{a a}^{a}$ from its original position, i.e., block $(1,1)$, to block $(5,5)$ in (4.1.60).

Next, define a subsystem ( $A_{s}, B_{s}, C_{s}, D_{s}$ ) with

$$
A_{s}:=\left[\begin{array}{cc}
A_{d d}^{*} & B_{d} C_{1}^{a}  \tag{4.1.64}\\
B_{1}^{a} C_{d} & A_{a a}^{a}+B_{2}^{a} C_{2}^{a}
\end{array}\right], \quad B_{s}:=\left[\begin{array}{cc}
0 & B_{d} \\
B_{2}^{a} & 0
\end{array}\right],
$$

and

$$
C_{s}:=\left[\begin{array}{cc}
0 & C_{2}^{a}  \tag{4.1.65}\\
C_{d} & 0
\end{array}\right], \quad D_{s}:=\left[\begin{array}{cc}
I_{m_{0}} & 0 \\
0 & 0
\end{array}\right] .
$$

It is straightforward to verify that with the choice of $B^{a}$ and $C^{a}$ as in (4.1.58) and (4.1.59), $A_{s}$ has no eigenvalue at $a=1$. Hence $A^{*}$ has no eigenvalue at $a=1$ either, since both $A_{b b}^{*}$ and $A_{c c}^{*}$ have all eigenvalues at 0 , and $A_{a a}^{*}$ contains only the invariant zeros of $\Sigma_{c}$ which are not equal to $a=1$. Applying the bilinear transformation (4.1.4) to $\Sigma_{c}^{*}$, it follows from Lemma 4.1.1 that we obtain a discrete-time system $\Sigma_{d}^{*}$, characterized by ( $\tilde{A}^{*}, \tilde{B}^{*}, \tilde{C}^{*}, \tilde{D}^{*}$ ), with

$$
\begin{gather*}
\tilde{A}^{*}=\left[\begin{array}{cccc}
\left(I+A_{a a}^{*}\right)\left(I-A_{a a}^{*}\right)^{-1} & 0 & 0 & 0 \\
0 & \left(I+A_{b b}^{*}\right)\left(I-A_{b b}^{*}\right)^{-1} & 0 & 0 \\
0 & 0 & \left(I+A_{c c}^{*}\right)\left(I-A_{c c}^{*}\right)^{-1} & 0 \\
0 & 0 & 0 & \left(I+A_{s}\right)\left(I-A_{s}\right)^{-1}
\end{array}\right], \\
\tilde{B}^{*}=\sqrt{2}\left[\begin{array}{cc}
0 & 0 \\
0 & 0 \\
0 & \left(I-A_{c c}^{*}\right)^{-1} B_{c} \\
\left(I-A_{s}\right)^{-1} B_{s} & 0
\end{array}\right], \tag{4.1.66}
\end{gather*}
$$

and

$$
\tilde{D}^{*}=\left[\begin{array}{cc}
D_{s}+C_{s}\left(I-A_{s}\right)^{-1} B_{s} & 0  \tag{4.1.69}\\
0 & 0
\end{array}\right] .
$$

Our next task is to find appropriate transformations, state feedback, and output injection laws, so as to transform the above system into the form of the special coordinate basis displaying the properties $3(\mathrm{a})-5(\mathrm{~b})$.

To simplify the presentation, we first focus on the subsystem $\left(\tilde{A}_{s}, \tilde{B}_{s}, \tilde{C}_{s}, \tilde{D}_{s}\right)$ with

$$
\begin{equation*}
\tilde{A}_{s}:=\left(I+A_{s}\right)\left(I-A_{s}\right)^{-1}, \quad \tilde{B}_{s}:=\sqrt{2}\left(I-A_{s}\right)^{-1} B_{s}, \tag{4.1.70}
\end{equation*}
$$

and

$$
\begin{equation*}
\tilde{C}_{s}:=\sqrt{2} C_{s}\left(I-A_{s}\right)^{-1}, \quad \tilde{D}_{s}:=D_{s}+C_{s}\left(I-A_{s}\right)^{-1} B_{s} \tag{4.1.71}
\end{equation*}
$$

Using (4.1.57) in conjunction with Appendix A. 22 of Kailath [48], it is straightforward to compute $\left(I-A_{s}\right)^{-1}=$

$$
\left[\begin{array}{cc}
X_{1} & \left(I-A_{d d}^{*}\right)^{-1} B_{d} C_{1}^{a}\left(I-A_{a a}-B^{a} C^{a}\right)^{-1}  \tag{4.1.72}\\
\left(I-A_{a a}-B^{a} C^{a}\right)^{-1} B_{1}^{a} C_{d}\left(I-A_{d d}^{*}\right)^{-1} & \left(I-A_{a a}-B^{a} C^{a}\right)^{-1}
\end{array}\right]
$$

where

$$
X_{1}=\left(I-A_{d d}^{*}\right)^{-1}+\left(I-A_{d d}^{*}\right)^{-1} B_{d} C_{1}^{a}\left(I-A_{a a}-B^{a} C^{a}\right)^{-1} B_{1}^{a} C_{d}\left(I-A_{d d}^{*}\right)^{-1}
$$

and hence

$$
\tilde{A}_{s}=\left[\begin{array}{c}
X_{2} \\
2\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{1}^{a} C_{d}\left(I-A_{d d}^{*}\right)^{-1}  \tag{4.1.73}\\
2\left(I-A_{d d}^{*}\right)^{-1} B_{d} C_{1}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} \\
\left(I+A_{a a}^{a}+B^{a} C^{a}\right)\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1}
\end{array}\right],
$$

where

$$
\left.\begin{array}{c}
X_{2}=\left(I+A_{d d}^{*}\right)\left(I-A_{d d}^{*}\right)^{-1}+2\left(I-A_{d d}^{*}\right)^{-1} B_{d} C_{1}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{1}^{a} C_{d}\left(I-A_{d d}^{*}\right)^{-1}, \\
\tilde{B}_{s}=\sqrt{2}\left[\left(I-A_{d d}^{*}\right)^{-1} B_{d} C_{1}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{2}^{a}\right. \\
\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{2}^{a} \\
\left(I-A_{d d}^{*}\right)^{-1} B_{d}\left[I+C_{1}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{1}^{a}\right] \\
\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{1}^{a}
\end{array}\right], ~(4.1 .74), ~\left[\begin{array}{c}
\tilde{C}_{s}=\sqrt{2}\left[\begin{array}{c}
C_{2}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{1}^{a} C_{d}\left(I-A_{d d}^{*}\right)^{-1} \\
{\left[I+C_{1}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{1}^{a}\right] C_{d}\left(I-A_{d d}^{*}\right)^{-1}} \\
C_{2}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} \\
C_{1}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1}
\end{array}\right] \tag{4.1.75}
\end{array}\right.
$$

and

$$
\tilde{D}_{s}=\left[\begin{array}{cc}
I+C_{2}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{2}^{a} & C_{2}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{1}^{a}  \tag{4.1.76}\\
C_{1}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{2}^{a} & I+C_{1}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{1}^{a}
\end{array}\right] .
$$

Noting the structure of $A_{a a}^{a}$ in (4.1.51), and the structures of $B^{a}$ and $C^{a}$ in (4.1.58) and (4.1.59), we have

$$
\left(I-A_{a a}-B^{a} C^{a}\right)^{-1}=\left[\begin{array}{ccccc}
0 & -1 & \cdots & 0 & 0  \tag{4.1.77}\\
-I_{n_{a, 1}-1} & 0 & \cdots & 0 & 0 \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & 0 & -1 \\
0 & 0 & \cdots & -I_{n_{a, \tau_{a}}-1} & 0
\end{array}\right]
$$

$$
\begin{equation*}
C_{1}^{a}\left(I-A_{a a}-B^{a} C^{a}\right)^{-1} B_{2}^{a}=0, \quad C_{2}^{a}\left(I-A_{a a}-B^{a} C^{a}\right)^{-1} B_{1}^{a}=0, \tag{4.1.78}
\end{equation*}
$$

and

$$
C^{a}\left(I-A_{a a}-B^{a} C^{a}\right)^{-1} B^{a}=\left[\begin{array}{cc}
0 & 0  \tag{4.1.79}\\
0 & -I_{\tau_{a}}
\end{array}\right] .
$$

Thus, $\tilde{B}_{s}, \tilde{C}_{s}$ and $\tilde{D}_{s}$ are reduced to the following forms:

$$
\tilde{B}_{s}=\sqrt{2}\left[\begin{array}{cc}
0 & \left(I-A_{d d}^{*}\right)^{-1} B_{d}\left[I+C_{1}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{1}^{a}\right]  \tag{4.1.80}\\
\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{2}^{a} & \left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{1}^{a}
\end{array}\right],
$$

$$
\tilde{C}_{s}=\sqrt{2}\left[\begin{array}{c}
0 \\
{\left[I+C_{1}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{1}^{a}\right] C_{d}\left(I-A_{d d}^{*}\right)^{-1}}  \tag{4.1.81}\\
\\
C_{2}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} \\
C_{1}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1}
\end{array}\right],
$$

and

$$
\tilde{D}_{s}=\left[\begin{array}{cc}
I+C_{2}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{2}^{a} & 0  \tag{4.1.82}\\
0 & I+C_{1}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{1}^{a}
\end{array}\right] .
$$

Next, define

$$
\tilde{F}_{s}:=\sqrt{2}\left[\begin{array}{cc}
0 & 0  \tag{4.1.83}\\
-C_{d}\left(I-A_{d d}^{*}\right)^{-1} & 0
\end{array}\right],
$$

and

$$
\tilde{K}_{s}:=\sqrt{2}\left[\begin{array}{cc}
0 & -\left(I-A_{d d}^{*}\right)^{-1} B_{d}  \tag{4.1.84}\\
0 & 0
\end{array}\right],
$$

from which it follows that

$$
\begin{aligned}
\tilde{A}_{s c} & =\tilde{A}_{s}+\tilde{B}_{s} \tilde{F}_{s}+\tilde{K}_{s} \tilde{C}_{s}+\tilde{K}_{s} \tilde{D}_{s} \tilde{F}_{s} \\
& =\left[\begin{array}{cc}
\tilde{A}_{a a}^{* *} & 0 \\
0 & \left(I+A_{a a}^{a}+B^{a} C^{a}\right)\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1}
\end{array}\right],
\end{aligned}
$$

where

$$
\begin{gather*}
\tilde{A}_{a a}^{* *}:=\left(I+A_{d d}^{*}\right)\left(I-A_{d d}^{*}\right)^{-1}-2\left(I-A_{d d}^{*}\right)^{-1} B_{d} C_{d}\left(I-A_{d d}^{*}\right)^{-1},  \tag{4.1.85}\\
\tilde{B}_{s c}=\tilde{B}_{s}+\tilde{K}_{s} \tilde{D}_{s}=\sqrt{2}\left[\begin{array}{cc}
0 & 0 \\
\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{2}^{a} & \left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} B_{1}^{a}
\end{array}\right],
\end{gather*}
$$

and

$$
\tilde{C}_{s c}=\tilde{C}_{s}+\tilde{D}_{s} \tilde{F}_{s}=\sqrt{2}\left[\begin{array}{cc}
0 & C_{2}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1} \\
0 & C_{1}^{a}\left(I-A_{a a}^{a}-B^{a} C^{a}\right)^{-1}
\end{array}\right] .
$$

Next, repartition $B^{a}$ and $C^{a}$ of (4.1.58) and (4.1.59) as follows:

$$
B^{a}=\left[\begin{array}{ll}
0 & \tilde{B}_{a}
\end{array}\right] \quad \text { and } \quad C^{a}=\left[\begin{array}{c}
0  \tag{4.1.86}\\
\tilde{C}_{a}
\end{array}\right]
$$

where both $\tilde{B}_{a}$ and $\tilde{C}_{a}$ are of maximal rank. We thus obtain

$$
\begin{gathered}
\tilde{A}_{s c}=\left[\begin{array}{cc}
\tilde{A}_{a a}^{* *} & 0 \\
0 & \left(I+A_{a a}^{a}+\tilde{B}_{a} \tilde{C}_{a}\right)\left(I-A_{a a}^{a}-\tilde{B}_{a} \tilde{C}_{a}\right)^{-1}
\end{array}\right] \\
\tilde{B}_{s c}=\sqrt{2}\left[\begin{array}{cc}
0 & 0 \\
0 & \left(I-A_{a a}-\tilde{B}_{a} \tilde{C}_{a}\right)^{-1} \tilde{B}_{a}
\end{array}\right]
\end{gathered}
$$

and

$$
\tilde{C}_{s c}=\sqrt{2}\left[\begin{array}{cc}
0 & 0 \\
0 & \tilde{C}_{a}\left(I-A_{a a}-\tilde{B}_{a} \tilde{C}_{a}\right)^{-1}
\end{array}\right], \quad \tilde{D}_{s c}=\tilde{D}_{s}=\left[\begin{array}{cc}
I_{m_{0}+m_{d}-\tau_{a}} & 0 \\
0 & 0
\end{array}\right]
$$

Using (4.1.51) and (4.1.77), straightforward manipulations yield

$$
\begin{aligned}
&\left(I+A_{a a}^{a}+\tilde{B}_{a} \tilde{C}_{a}\right)\left(I-A_{a a}^{a}-\tilde{B}_{a} \tilde{C}_{a}\right)^{-1} \\
&= {\left[\begin{array}{ccc}
{\left[\begin{array}{cc}
0 & -2 \\
-2 I_{n_{a, 1}-1} & 0
\end{array}\right]-I_{n_{a, 1}}} & \cdots & \\
\vdots & \ddots & 0 \\
0 & \cdots & {\left[\begin{array}{cc}
0 & \\
-2 I_{n_{a, \tau_{a}-1}} & 0
\end{array}\right]-I_{n_{a, \tau_{a}}}}
\end{array}\right] } \\
&\left(I-A_{a a}^{a}-\tilde{B}_{a} \tilde{C}_{a}\right)^{-1} \tilde{B}_{a}=-\left[\begin{array}{ccc}
1 & \cdots & 0 \\
0 & \cdots & 0 \\
\vdots & \ddots & \vdots \\
0 & \cdots & 1 \\
0 & \cdots & 0
\end{array}\right]
\end{aligned}
$$

and

$$
\tilde{C}_{a}\left(I-A_{a a}^{a}-\tilde{B}_{a} \tilde{C}_{a}\right)^{-1}=-\left[\begin{array}{ccccc}
0 & 1 & \cdots & 0 & 0 \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & 0 & 1
\end{array}\right]
$$

Moreover, it can be readily verified that each subsystem $\left(\tilde{A}_{a i}, \tilde{B}_{a i}, \tilde{C}_{a i}\right), i=$ $1, \cdots, \tau_{a}$, with

$$
\tilde{A}_{a i}=-I_{n_{a, i}}+\left[\begin{array}{cc}
0 & -2 \\
-2 I_{n_{a, i}-1} & 0
\end{array}\right], \quad \tilde{B}_{a i}=\left[\begin{array}{r}
-1 \\
0
\end{array}\right], \quad \tilde{C}_{a i}=\left[\begin{array}{ll}
0 & -1
\end{array}\right]
$$

has the following properties:

$$
\tilde{C}_{a i} \tilde{B}_{a i}=\tilde{C}_{a i} \tilde{A}_{a i} \tilde{B}_{a i}=\cdots=\tilde{C}_{a i}\left(\tilde{A}_{a i}\right)^{n_{a, i}-2} \tilde{B}_{a i}=0
$$

and

$$
\tilde{C}_{a i}\left(\tilde{A}_{a i}\right)^{n_{a, i}-1} \tilde{B}_{a i} \neq 0
$$

It follows from Theorem 2.3.1 that there exist nonsingular transformations $\Gamma_{s a}$, $\Gamma_{o a}$ and $\Gamma_{i a}$ such that

$$
\begin{gather*}
\tilde{A}_{d}=\Gamma_{s a}^{-1}\left[\left(I+A_{a a}^{a}+\tilde{B}_{a} \tilde{C}_{a}\right)\left(I-A_{a a}^{a}-\tilde{B}_{a} \tilde{C}_{a}\right)^{-1}\right] \Gamma_{s a} \\
=\left[\begin{array}{ccccc}
\star & I_{n_{a, 1}-1} & \cdots & 0 & 0 \\
\star & \star & \cdots & 0 & 0 \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & \star & I_{n_{a, \tau_{a}-1}} \\
0 & 0 & \cdots & \star & \star
\end{array}\right],  \tag{4.1.87}\\
\tilde{B}_{d}=\Gamma_{s a}^{-1}\left[\left(I-A_{a a}^{a}-\tilde{B}_{a} \tilde{C}_{a}\right)^{-1} \tilde{B}_{a}\right] \Gamma_{i a}=\left[\begin{array}{ccc}
0 & \cdots & 0 \\
1 & \cdots & 0 \\
\vdots & \ddots & \vdots \\
0 & \cdots & 0 \\
0 & \cdots & 1
\end{array}\right] \tag{4.1.88}
\end{gather*}
$$

and

$$
\tilde{C}_{d}=\Gamma_{o a}^{-1}\left[\tilde{C}_{a}\left(I-A_{a a}^{a}-\tilde{B}_{a} \tilde{C}_{a}\right)^{-1}\right] \Gamma_{s a}=\left[\begin{array}{ccccc}
1 & 0 & \cdots & 0 & 0  \tag{4.1.89}\\
\vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & 1 & 0
\end{array}\right]
$$

Now, let us return to $\Sigma_{d}^{*}$ characterized by $\left(\tilde{A}^{*}, \tilde{B}^{*}, \tilde{C}^{*}, \tilde{D}^{*}\right)$ as in (4.1.66) to (4.1.69). Using the properties of the subsystem $\left(\tilde{A}_{s}, \tilde{B}_{s}, \tilde{C}_{s}, \tilde{D}_{s}\right)$ just derived, we are in a position to define appropriate state feedback and output injection gain matrices, say $\tilde{F}^{*}$ and $\tilde{K}^{*}$, together with nonsingular state, output and input transformations $\tilde{\Gamma}_{s}^{*}, \tilde{\Gamma}_{o}^{*}$ and $\tilde{\Gamma}_{i}^{*}$, such that

$$
\begin{align*}
\tilde{A}_{\mathrm{sCB}}^{*} & :=\left(\tilde{\Gamma}_{s}^{*}\right)^{-1}\left(\tilde{A}^{*}+\tilde{B}^{*} \tilde{F}^{*}+\tilde{K}^{*} \tilde{C}^{*}+\tilde{K}^{*} \tilde{D}^{*} \tilde{F}^{*}\right) \tilde{\Gamma}_{s}^{*} \\
& =\left[\begin{array}{ccccc}
\left(I+A_{a a}^{*}\right)\left(I-A_{a a}^{*}\right)^{-1} & 0 & 0 & 0 & 0 \\
0 & \left(I+A_{b b}^{*}\right)\left(I-A_{b b}^{*}\right)^{-1} & 0 & 0 & 0 \\
0 & 0 & \left(I+A_{c c}^{*}\right)\left(I-A_{c c}^{*}\right)^{-1} & 0 & 0 \\
0 & 0 & 0 & \tilde{A}_{a a}^{* *} & 0 \\
0 & 0 & 0 & 0 & \tilde{A}_{d}
\end{array}\right],
\end{align*}
$$

with $\tilde{A}_{a a}^{* *}$ given by (4.1.85), and

$$
\begin{align*}
& \tilde{B}_{\mathrm{sCB}}^{*}:=\left(\tilde{\Gamma}_{s}^{*}\right)^{-1}\left(\tilde{B}^{*}+\tilde{K}^{*} \tilde{D}^{*}\right) \tilde{\Gamma}_{i}^{*}=\left[\begin{array}{ccc}
0 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & \left(I-A_{c c}^{*}\right)^{-1} B_{c} \\
0 & 0 & 0 \\
0 & \tilde{B}_{d} & 0
\end{array}\right],  \tag{4.1.91}\\
& \tilde{C}_{\mathrm{SCB}}^{*}:=\left(\tilde{\Gamma}_{o}^{*}\right)^{-1}\left(\tilde{C}^{*}+\tilde{D}^{*} \tilde{F}^{*}\right) \tilde{\Gamma}_{s}^{*}=\left[\begin{array}{ccccc}
0 & 0 & 0 & 0 & 0 \\
0 & C_{b}\left(I-A_{b b}^{*}\right)^{-1} & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & \tilde{C}_{d}
\end{array}\right], \tag{4}
\end{align*}
$$

and

$$
\tilde{D}_{\mathrm{SCB}}^{*}:=\left(\tilde{\Gamma}_{o}^{*}\right)^{-1} \tilde{D}^{*} \tilde{\Gamma}_{i}^{*}=\left[\begin{array}{ccc}
I_{m_{0}+m_{d}-\tau_{a}} & 0 & 0  \tag{4.1.93}\\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right]
$$

Clearly, $\Sigma_{\mathrm{SCB}}^{*}$ characterized by $\left(\tilde{A}_{\mathrm{SCB}}^{*}, \tilde{B}_{\mathrm{SCB}}^{*}, \tilde{C}_{\mathrm{SCB}}^{*}, \tilde{D}_{\mathrm{SCB}}^{*}\right)$ has the same structural invariant indices lists as does $\Sigma_{d}^{*}$, which in turn has the same structural invariant indices lists as $\Sigma_{d}$. Most importantly, however, $\Sigma_{\mathrm{SCB}}^{*}$ is in the form of the special coordinate basis, and we are now ready to prove properties 3 (a)-5(b) of the theorem.

3(a). First, we note that $\mathcal{I}_{2}\left(\Sigma_{d}\right)=\mathcal{I}_{2}\left(\Sigma_{\mathrm{sCB}}^{*}\right)$. From (4.1.90) to (4.1.93) and the properties of the special coordinate basis, we know that $\mathcal{I}_{2}\left(\Sigma_{\mathrm{SCB}}^{*}\right)$ is given by the controllability index of the pair

$$
\left(\left(I+A_{c c}^{*}\right)\left(I-A_{c c}^{*}\right)^{-1},\left(I-A_{c c}^{*}\right)^{-1} B_{c}\right) \text { or }\left(\left(I+A_{c c}^{*}\right)\left(I-A_{c c}^{*}\right)^{-1}, B_{c}\right)
$$

Recalling the definitions of $A_{c c}^{*}$ and $B_{c}$ :

$$
A_{c c}^{*}=\left[\begin{array}{ccccc}
0 & I_{\ell_{1}-1} & \cdots & 0 & 0 \\
0 & 0 & \cdots & 0 & 0 \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & 0 & I_{\ell_{m_{c}-1}} \\
0 & 0 & \cdots & 0 & 0
\end{array}\right], \quad B_{c}=\left[\begin{array}{ccc}
0 & \cdots & 0 \\
1 & \cdots & 0 \\
\vdots & \ddots & \vdots \\
0 & \cdots & 0 \\
0 & \cdots & 1
\end{array}\right]
$$

it is straightforward to verify that the controllability index of

$$
\left(\left(I+A_{c c}^{*}\right)\left(I-A_{c c}^{*}\right)^{-1}, B_{c}\right)
$$

is also given by $\left\{\ell_{1}, \cdots, \ell_{m_{c}}\right\}$, and thus $\mathcal{I}_{2}\left(\Sigma_{d}\right)=\mathcal{I}_{2}\left(\Sigma_{c}\right)$.
Likewise, the proof that $\mathcal{I}_{3}\left(\Sigma_{d}\right)=\mathcal{I}_{3}\left(\Sigma_{c}\right)$ follows along similar lines. 区 3(b)-3(c). These follow directly from 3(a).
4(a). It follows from the properties of the special coordinate basis that the invariant zero structure of $\tilde{\Sigma}_{\text {sCB }}^{*}$, or equivalently $\Sigma_{d}$, is given by the eigenvalues of $\tilde{A}_{a a}^{* *}$ and $\left(I+A_{a a}^{*}\right)\left(I-A_{a a}^{*}\right)^{-1}$, together with their associated Jordan blocks. Property 4(a) corresponds with the eigenvalues of $\tilde{A}_{a a}^{* *}$ of (4.1.85), together with their associated Jordan blocks. First, we note that for any $z \in \mathbb{C}$,

$$
\begin{equation*}
z I-\tilde{A}_{a a}^{* *}=\left[(z-1) I-(z+1) A_{d d}^{*}+2\left(I-A_{d d}^{*}\right)^{-1} B_{d} C_{d}\right]\left(I-A_{d d}^{*}\right)^{-1} \tag{4.1.94}
\end{equation*}
$$

Recall the definitions of $A_{d d}^{*}, B_{d}$ and $C_{d}$ :

$$
A_{d d}^{*}=\left[\begin{array}{ccccc}
0 & I_{n_{q_{1}}-1} & \cdots & 0 & 0 \\
0 & 0 & \cdots & 0 & 0 \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & 0 & I_{q_{m_{d}}-1} \\
0 & 0 & \cdots & 0 & 0
\end{array}\right], \quad B_{d}=\left[\begin{array}{ccc}
0 & \cdots & 0 \\
1 & \cdots & 0 \\
\vdots & \ddots & \vdots \\
0 & \cdots & 0 \\
0 & \cdots & 1
\end{array}\right]
$$

and

$$
C_{d}=\left[\begin{array}{ccccc}
1 & 0 & \cdots & 0 & 0 \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & 1 & 0
\end{array}\right]
$$

It can be shown that

$$
(z-1) I-(z+1) A_{d d}^{*}+2\left(I-A_{d d}^{*}\right)^{-1} B_{d} C_{d}=\operatorname{blkdiag}\left\{Q_{1}(z), \cdots, Q_{i}(z)\right\}
$$

where $Q_{i}(z) \in \mathbb{C}^{n_{q_{i}} \times n_{q_{i}}}$ is given by

$$
Q_{i}(z)=\left[\begin{array}{cccccc}
z+1 & -(z+1) & 0 & \cdots & 0 & 0  \tag{4.1.95}\\
2 & z-1 & -(z+1) & \cdots & 0 & 0 \\
2 & 0 & z-1 & \ddots & 0 & 0 \\
\vdots & \vdots & \vdots & \ddots & \ddots & \vdots \\
2 & 0 & 0 & \cdots & z-1 & -(z+1) \\
2 & 0 & 0 & \cdots & 0 & z-1
\end{array}\right]
$$

for $i=1, \cdots, m_{d}$ ．It follows from（4．1．94）that the eigenvalue of $\tilde{A}_{a a}^{* *}$ is the scalar $z$ that causes the rank of

$$
\operatorname{blkdiag}\left\{Q_{1}(z), \cdots, Q_{m_{d}}(z)\right\}
$$

to drop below $n_{d}=\sum_{i=1}^{m_{d}} q_{i}$ ．Using the particular form of $Q_{i}(z)$ ，it is clear that the only such scalar $z \in \mathbb{C}$ which causes $Q_{i}(z)$ to drop rank is $z=-1$ ． Moreover， $\operatorname{rank}\left\{Q_{i}(-1)\right\}=n_{q_{i}}-1$ ，i．e．，$Q_{i}(-1)$ has only one linearly indepen－ dent eigenvector．Hence，$z=-1$ is the eigenvalue of $\tilde{A}_{a a}^{* *}$ ，or equivalently the invariant zero of $\Sigma_{d}$ ，with the multiplicity structure

$$
S_{-1}^{\star}\left(\Sigma_{d}\right)=\left\{q_{1}, \cdots, q_{m_{d}}\right\}=S_{\infty}^{\star}\left(\Sigma_{c}\right),
$$

thereby proving $4(\mathrm{a})$ ．
4（b）．This part of the infinite zero structure corresponds to the invariant zeros of the matrix $\left(I+A_{a a}^{*}\right)\left(I-A_{a a}^{*}\right)^{-1}$ ．With $A_{a a}^{*}$ in Jordan form，Property 4（b） follows by straightforward manipulations．

5（a）．It follows directly from（4．1．93）．
区
5（b）．This follows from the structure of $\left(\tilde{A}_{d}, \tilde{B}_{d}, \tilde{C}_{d}\right)$ in（4．1．87）to（4．1．89），in conjunction with Property 2.3 .3 of the special coordinate basis．

区
6（a）－6（b）．We let the state space of the system（4．1．2）be $\mathcal{X}$ and be partitioned in its SCB subsystems as follows：

$$
\begin{equation*}
\mathcal{X}=\mathcal{X}_{a}^{-} \oplus \mathcal{X}_{a}^{0} \oplus \mathcal{X}_{a}^{+} \oplus \mathcal{X}_{b} \oplus \mathcal{X}_{c} \oplus \mathcal{X}_{d} \tag{4.1.96}
\end{equation*}
$$

We further partition $\mathcal{X}_{a}^{+}$as

$$
\begin{equation*}
\mathcal{X}_{a}^{+}=\mathcal{X}_{a 1}^{+} \oplus \mathcal{X}_{a *}^{+} \tag{4.1.97}
\end{equation*}
$$

where $\mathcal{X}_{a 1}^{+}$is associated with the zero dynamics of the unstable zero of (4.1.2) at $s=a=1$ and $\mathcal{X}_{a *}^{+}$is associated with the rest of unstable zero dynamics of (4.1.2). Similarly, we let the state space of the transformed system (4.1.7) be $\tilde{\mathcal{X}}$ and be partitioned in its SCB subsystems as follows:

$$
\begin{equation*}
\tilde{\mathcal{X}}=\tilde{\mathcal{X}}_{a}^{-} \oplus \tilde{\mathcal{X}}_{a}^{0} \oplus \tilde{\mathcal{X}}_{a}^{+} \oplus \tilde{\mathcal{X}}_{b} \oplus \tilde{\mathcal{X}}_{c} \oplus \tilde{\mathcal{X}}_{d} \tag{4.1.98}
\end{equation*}
$$

with $\tilde{\mathcal{X}}_{a}^{0}$ being further partitioned as

$$
\begin{equation*}
\tilde{\mathcal{X}}_{a}^{0}=\tilde{\mathcal{X}}_{a 1}^{0} \oplus \tilde{\mathcal{X}}_{a *}^{0}, \tag{4.1.99}
\end{equation*}
$$

where $\tilde{\mathcal{X}}_{a 1}^{0}$ is associated with the zero dynamics of the invariant zero of (4.1.7) at $z=-1$ and $\tilde{\mathcal{X}}_{a *}^{0}$ is associated the rest of the zero dynamics of the zeros of (4.1.7) on the unit circle. Then, from the above derivations of 1 (a) to 5 (b), we have the following mappings between the subsystems of $\Sigma_{c}$ of (4.1.2) and those of $\Sigma_{d}$ of (4.1.7):


Noting that both geometric subspaces $\mathcal{V}^{\mathrm{x}}$ and $\mathcal{S}^{\mathrm{x}}$ are invariant under any nonsingular output and input transformations, as well as any state feedback and output injection laws, we have

$$
\begin{equation*}
\mathcal{V}^{+}\left(\Sigma_{c}\right)=\mathcal{X}_{a *}^{+} \oplus \mathcal{X}_{a 1}^{+} \oplus \mathcal{X}_{c}=\tilde{\mathcal{X}}_{a}^{+} \oplus \tilde{\mathcal{X}}_{d} \oplus \tilde{\mathcal{X}}_{c}=\mathcal{S}^{\ominus}\left(\Sigma_{d}\right) \tag{4.1.101}
\end{equation*}
$$

and

$$
\mathcal{S}^{+}\left(\Sigma_{c}\right)=\mathcal{X}_{a}^{-} \oplus \mathcal{X}_{a}^{0} \oplus \mathcal{X}_{c} \oplus \mathcal{X}_{d}=\tilde{\mathcal{X}}_{a}^{-} \oplus \tilde{\mathcal{X}}_{a *}^{0} \oplus \tilde{\mathcal{X}}_{c} \oplus \tilde{\mathcal{X}}_{a 1}^{0}=\mathcal{V}^{\odot}\left(\Sigma_{d}\right)
$$

Unfortunately, other geometric subspaces do not have such clear relationships as above.

This concludes the proof of Theorem 4.1.1 and this section.

### 4.2. Solutions to Discrete-time Riccati Equations

The discrete-time algebraic Riccati equation (DARE) has been investigated extensively in the literature (see, for example [7,51,54,77,82,99]). Here, most of the work was based on the discrete-time algebraic Riccati equation appearing in a linear quadratic control problem (hereafter we will refer to such a DARE as the $H_{2}$-DARE). Recently, the problem of $H_{\infty}$ control and that of differential games for discrete-time systems, have been studied by a number of researchers including $[4,47,59]$. This work gives rise to a different kind of algebraic Riccati equation (hereafter we call it an $H_{\infty}$-DARE). Analyzing and solving such an $H_{\infty}$-DARE are very difficult primarily because of an indefinite nonlinear term and because we cannot a-priori guarantee the existence of solutions. In this section, we recall the result of Chen et al [30] on a non-recursive method for solving general DARE's, as well as $H_{2}$-DARE's and $H_{\infty}$-DARE's. We cast the problem of solving a given DARE to the problem of solving an auxiliary continuous-time algebraic Riccati equation (CARE). The latter can be solved using the well known non-recursive methods available in the literature. The advantages of this approach over the recursive method are three-fold: (a) it reduces the computation involved while giving much more accurate solutions, (b) it brings a clear intuition to the conditions associated with the $H_{\infty}$-DARE, and (c) some of the properties of the $H_{\infty}$-DARE follow readily from the continuoustime counterpart.

### 4.2.1. Solution to a General DARE

We first introduce in this subsection a non-recursive method for solving the following discrete-time algebraic Riccati equation, which is even more general than the $H_{\infty}$-DARE and which plays a critical role in solving the $H_{\infty}$-DARE,

$$
\begin{equation*}
P=A^{\prime} P A-\left(A^{\prime} P M+N\right)\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)+Q \tag{4.2.1}
\end{equation*}
$$

where $A, M, N, R$ and $Q$ are real matrices of dimensions $n \times n, n \times m, n \times m$, $m \times m$ and $n \times n$, respectively, and with $Q$ and $R$ being symmetric matrices. We will show that the DARE of (4.2.1) can be converted to a continuous-time Riccati equation. Assume that matrix $A$ has no eigenvalue at -1 . We define

$$
\left.\begin{array}{rl}
F & :=(A+I)^{-1}(A-I)  \tag{4.2.2}\\
G:=2(A+I)^{-2} M \\
W:=R+M^{\prime}\left(A^{\prime}+I\right)^{-1} Q(A+I)^{-1} M \\
& -N^{\prime}(A+I)^{-1} M-M^{\prime}\left(A^{\prime}+I\right)^{-1} N \\
H: & =-Q(A+I)^{-1} M+N
\end{array}\right\}
$$

We have the following theorem.
Theorem 4.2.1. Assume that matrix $A$ has no eigenvalue at -1 . Then the following two statements are equivalent.

1. $P$ is a symmetric solution to the DARE (4.2.1) and $W$ is nonsingular.
2. $\tilde{P}$ is a symmetric solution to the continuous algebraic Riccati equation,

$$
\begin{equation*}
\tilde{P} F+F^{\prime} \tilde{P}-(\tilde{P} G+H) W^{-1}(\tilde{P} G+H)^{\prime}+Q=0 \tag{4.2.3}
\end{equation*}
$$

and $R+2 G^{\prime}\left(I-F^{\prime}\right)^{-1} \tilde{P}(I-F)^{-1} G$ is nonsingular.
Moreover, $P$ and $\tilde{P}$ are related by $P=2\left(A^{\prime}+I\right)^{-1} \tilde{P}(A+I)^{-1}$.
Proof. First, let us consider the following reductions:

$$
\begin{align*}
A^{\prime} P A & -P+Q=2 A^{\prime}\left(A^{\prime}+I\right)^{-1} \tilde{P}(A+I)^{-1} A-2\left(A^{\prime}+I\right)^{-1} \tilde{P}(A+I)^{-1}+Q \\
& =2\left(A^{\prime}+I\right)^{-1} A^{\prime} \tilde{P} A(A+I)^{-1}-2\left(A^{\prime}+I\right)^{-1} \tilde{P}(A+I)^{-1}+Q \\
& =\left(A^{\prime}+I\right)^{-1}\left(2 A^{\prime} \tilde{P} A-2 \tilde{P}\right)(A+I)^{-1}+Q \\
& =\left(A^{\prime}+I\right)^{-1}\left[\left(A^{\prime}+I\right) \tilde{P}(A-I)+\left(A^{\prime}-I\right) \tilde{P}(A+I)\right](A+I)^{-1}+Q \\
& =\tilde{P}(A-I)(A+I)^{-1}+\left(A^{\prime}+I\right)^{-1}\left(A^{\prime}-I\right) \tilde{P}+Q \\
& =\tilde{P} F+F^{\prime} \tilde{P}+Q . \tag{4.2.4}
\end{align*}
$$

(1. $\Rightarrow 2$.) Let us start with the following trivial equality,

$$
A^{\prime} P A-P+\left(A^{\prime}+I\right) P(A+I)-\left(A^{\prime}+I\right) P A-A^{\prime} P(A+I)=0
$$

which implies that

$$
\begin{aligned}
P-P A(A & +I)^{-1}-\left(A^{\prime}+I\right)^{-1} A^{\prime} P \\
& +\left(A^{\prime}+I\right)^{-1} A^{\prime} P A(A+I)^{-1}-\left(A^{\prime}+I\right)^{-1} P(A+I)^{-1}=0
\end{aligned}
$$

Then we have

$$
\begin{aligned}
W= & R+M^{\prime}\left(A^{\prime}+I\right)^{-1} Q(A+I)^{-1} M-N^{\prime}(A+I)^{-1} M-M^{\prime}\left(A^{\prime}+I\right)^{-1} N \\
= & R+M^{\prime}\left(A^{\prime}+I\right)^{-1} Q(A+I)^{-1} M-N^{\prime}(A+I)^{-1} M-M^{\prime}\left(A^{\prime}+I\right)^{-1} N \\
& +M^{\prime} P M-M^{\prime} P A(A+I)^{-1} M-M^{\prime}\left(A^{\prime}+I\right)^{-1} A^{\prime} P M \\
& +M^{\prime}\left(A^{\prime}+I\right)^{-1} A^{\prime} P A(A+I)^{-1} M-M^{\prime}\left(A^{\prime}+I\right)^{-1} P(A+I)^{-1} M \\
= & R+M^{\prime} P M-\left(M^{\prime} P A+N^{\prime}\right)(A+I)^{-1} M-M^{\prime}\left(A^{\prime}+I\right)^{-1}\left(A^{\prime} P M+N\right)
\end{aligned}
$$

$$
\begin{align*}
& +M^{\prime}\left(A^{\prime}+I\right)^{-1}\left(A^{\prime} P A+Q-P\right)(A+I)^{-1} M  \tag{4.2.5}\\
= & R+M^{\prime} P M-\left(M^{\prime} P A+N^{\prime}\right)(A+I)^{-1} M-M^{\prime}\left(A^{\prime}+I\right)^{-1}\left(A^{\prime} P M+N\right) \\
& +M^{\prime}\left(A^{\prime}+I\right)^{-1}\left(A^{\prime} P M+N\right)\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right) \\
& \times(A+I)^{-1} M  \tag{4.2.6}\\
= & {\left[I-M^{\prime}\left(A^{\prime}+I\right)^{-1}\left(A^{\prime} P M+N\right)\left(R+M^{\prime} P M\right)^{-1}\right] } \\
& \times\left(R+M^{\prime} P M\right)\left[I-\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)(A+I)^{-1} M\right] . \tag{4.2.7}
\end{align*}
$$

Here we note that we have used (4.2.1) to get (4.2.6) from (4.2.5). By the assumption that $W$ is nonsingular, we have

$$
\begin{aligned}
R+M^{\prime} P M=[I & \left.-M^{\prime}\left(A^{\prime}+I\right)^{-1}\left(A^{\prime} P M+N\right)\left(R+M^{\prime} P M\right)^{-1}\right]^{-1} W \\
& \times\left[I-\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)(A+I)^{-1} M\right]^{-1}
\end{aligned}
$$

Hence,

$$
\begin{align*}
\left(A^{\prime} P\right. & M+N)\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right) \\
= & \left(A^{\prime} P M+N\right)\left[I-\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)(A+I)^{-1} M\right] W^{-1} \\
& \times\left[I-\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)(A+I)^{-1} M\right]^{\prime}\left(M^{\prime} P A+N^{\prime}\right) \\
= & {\left[A^{\prime} P M-\left(A^{\prime} P M+N\right)\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)(A+I)^{-1} M+N\right] W^{-1} } \\
& \times\left[A^{\prime} P M-\left(A^{\prime} P M+N\right)\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)\right. \\
& \left.\times(A+I)^{-1} M+N\right]^{\prime}  \tag{4.2.8}\\
= & {\left[A^{\prime} P M+\left(P-A^{\prime} P A-Q\right)(A+I)^{-1} M+N\right] W^{-1} } \\
& \times\left[A^{\prime} P M+\left(P-A^{\prime} P A-Q\right)(A+I)^{-1} M+N\right]^{\prime}  \tag{4.2.9}\\
= & {\left[\left(A^{\prime} P+P-Q\right)(A+I)^{-1} M+N\right] W^{-1}\left[\left(A^{\prime} P+P-Q\right)(A+I)^{-1} M+N\right]^{\prime} } \\
= & {\left[\left(A^{\prime}+I\right) P(A+I)(A+I)^{-2} M-Q(A+I)^{-1} M+N\right] W^{-1} } \\
& \times\left[\left(A^{\prime}+I\right) P(A+I)(A+I)^{-2} M-Q(A+I)^{-1} M+N\right]^{\prime} \\
= & (\tilde{P} G+H) W^{-1}(\tilde{P} G+H)^{\prime} . \tag{4.2.10}
\end{align*}
$$

Again, we have used (4.2.1) to get (4.2.9) from (4.2.8). Finally, (4.2.1), (4.2.4) and (4.2.10) imply that

$$
\tilde{P} F+F^{\prime} \tilde{P}-(\tilde{P} G+H) W^{-1}(\tilde{P} G+H)^{\prime}+Q=0
$$

(2. $\Rightarrow$ 1.) It follows from (4.2.2) that

$$
\begin{align*}
A= & (I+F)(I-F)^{-1}, \\
M= & 2(I-F)^{-2} G, \\
H= & -Q(I-F)^{-1} G+N, \\
P= & \left(I-F^{\prime}\right) \tilde{P}(I-F) / 2  \tag{4.2.11}\\
W= & R+G^{\prime}\left(I-F^{\prime}\right)^{-1} Q(I-F)^{-1} G \\
& \quad-N^{\prime}(I-F)^{-1} G-G^{\prime}\left(I-F^{\prime}\right)^{-1} N, \\
R+M^{\prime} P M= & R+2 G^{\prime}\left(I-F^{\prime}\right)^{-1} \tilde{P}(I-F)^{-1} G .
\end{align*}
$$

Then we have

$$
\begin{align*}
& R+ M^{\prime} P M=R+G^{\prime}\left(I-F^{\prime}\right)^{-1}[Q+(\tilde{P}-\tilde{P} F-Q) \\
&\left.\quad+\left(\tilde{P}-F^{\prime} \tilde{P}-Q\right)+\left(\tilde{P} F+F^{\prime} \tilde{P}+Q\right)\right](I-F)^{-1} G \\
&= R+G^{\prime}\left(I-F^{\prime}\right)^{-1} Q(I-F)^{-1} G-N^{\prime}(I-F)^{-1} G-G^{\prime}\left(I-F^{\prime}\right)^{-1} N \\
& \quad+G^{\prime}\left(I-F^{\prime}\right)^{-1}\left[\tilde{P} G-Q(I-F)^{-1} G+N\right]+\left[\tilde{P} G-Q(I-F)^{-1} G+N\right]^{\prime} \\
& \quad \times(I-F)^{-1} G+G^{\prime}\left(I-F^{\prime}\right)^{-1}\left(\tilde{P} F+F^{\prime} \tilde{P}+Q\right)(I-F)^{-1} G  \tag{4.2.12}\\
&=W+G^{\prime}\left(I-F^{\prime}\right)^{-1}(\tilde{P} G+H)+(\tilde{P} G+H)^{\prime}(I-F)^{-1} G \\
& \quad+G^{\prime}\left(I-F^{\prime}\right)^{-1}(\tilde{P} G+H) W^{-1}(\tilde{P} G+H)^{\prime}(I-F)^{-1} G  \tag{4.2.13}\\
&= {\left[I+W^{-1}(\tilde{P} G+H)^{\prime}(I-F)^{-1} G\right]^{\prime} W\left[I+W^{-1}(\tilde{P} G+H)^{\prime}(I-F)^{-1} G\right] . } \tag{4.2.14}
\end{align*}
$$

Here we note that we have used (4.2.3) to get (4.2.13) from (4.2.12).
By assumption, we have $R+M^{\prime} P M$ nonsingular. Thus, we can rewrite (4.2.14) as,

$$
\begin{aligned}
W=\left[I+G^{\prime}\left(I-F^{\prime}\right)^{-1}(\tilde{P} G\right. & \left.+H) W^{-1}\right]^{-1}\left(R+M^{\prime} P M\right) \\
& \times\left[I+W^{-1}(\tilde{P} G+H)^{\prime}(I-F)^{-1} G\right]^{-1}
\end{aligned}
$$

We have the following reductions,

$$
\begin{align*}
(\tilde{P} G+ & H) W^{-1}(\tilde{P} G+H)^{\prime} \\
= & (\tilde{P} G+H)\left[I+W^{-1}(\tilde{P} G+H)^{\prime}(I-F)^{-1} G\right] \\
& \quad \times\left(R+M^{\prime} P M\right)^{-1}\left[I+W^{-1}(\tilde{P} G+H)^{\prime}(I-F)^{-1} G\right]^{\prime}(\tilde{P} G+H)^{\prime} \\
= & {\left[\tilde{P} G+H+(\tilde{P} G+H) W^{-1}(\tilde{P} G+H)^{\prime}(I-F)^{-1} G\right]\left(R+M^{\prime} P M\right)^{-1} } \\
& \quad \times\left[\tilde{P} G+H+(\tilde{P} G+H) W^{-1}(\tilde{P} G+H)^{\prime}(I-F)^{-1} G\right]^{\prime}  \tag{4.2.15}\\
= & {\left[\tilde{P} G-Q(I-F)^{-1} G+\left(\tilde{P} F+F^{\prime} \tilde{P}+Q\right)(I-F)^{-1} G+N\right]\left(R+M^{\prime} P M\right)^{-1} } \\
& \quad \times\left[\tilde{P} G-Q(I-F)^{-1} G+\left(\tilde{P} F+F^{\prime} \tilde{P}+Q\right)(I-F)^{-1} G+N\right]^{\prime}  \tag{4.2.16}\\
= & {\left[\left(I+F^{\prime}\right) \tilde{P}(I-F)^{-1} G+N\right]\left(R+M^{\prime} P M\right)^{-1}\left[G^{\prime}\left(I-F^{\prime}\right)^{-1} \tilde{P}(I+F)+N^{\prime}\right] } \\
= & \left(A^{\prime} P M+N\right)\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right) . \tag{4.2.17}
\end{align*}
$$

Again, we have used (4.2.3) to get (4.2.16) from (4.2.15). Finally, it follows from (4.2.3), (4.2.4) and (4.2.17) that

$$
A^{\prime} P A-\left(A^{\prime} P M+N\right)\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)+Q-P=0
$$

This completes the proof of Theorem 4.2.1.
因

### 4.2.2. Solution to an $H_{\infty}$-DARE

In this subsection we present a non-recursive procedure that generates symmetric positive semi-definite matrices $P$ such that

$$
\begin{align*}
& V:=B^{\prime} P B+D_{2}^{\prime} D_{2}>0  \tag{4.2.18}\\
& R:=\gamma^{2} I-D_{22}^{\prime} D_{22}-E^{\prime} P E \\
&  \tag{4.2.19}\\
& \quad+\left(E^{\prime} P B+D_{22}^{\prime} D_{2}\right) V^{-1}\left(B^{\prime} P E+D_{2}^{\prime} D_{22}\right)>0
\end{align*}
$$

and such that the following discrete-time algebraic Riccati equation (DARE) is satisfied:

$$
P=A^{\prime} P A+C_{2}^{\prime} C_{2}-\left[\begin{array}{c}
B^{\prime} P A+D_{2}^{\prime} C_{2}  \tag{4.2.20}\\
E^{\prime} P A+D_{22}^{\prime} C_{2}
\end{array}\right]^{\prime} G^{-1}\left[\begin{array}{c}
B^{\prime} P A+D_{2}^{\prime} C_{2} \\
E^{\prime} P A+D_{22}^{\prime} C_{2}
\end{array}\right]
$$

where

$$
G:=\left[\begin{array}{cc}
D_{2}^{\prime} D_{2}+B^{\prime} P B & D_{2}^{\prime} D_{22}+B^{\prime} P E  \tag{4.2.21}\\
D_{22}^{\prime} D_{2}+E^{\prime} P B & E^{\prime} P E+D_{22}^{\prime} D_{22}-\gamma^{2} I
\end{array}\right] .
$$

The conditions (4.2.18) and (4.2.19) guarantee that the matrix $G$ is invertible. We are particularly interested in solutions $P$ of (4.2.18), (4.2.19) and (4.2.20) such that all the eigenvalues of the matrix $A_{\mathrm{cl}}$ are inside the unit circle, where

$$
A_{\mathrm{cl}}:=A-\left[\begin{array}{ll}
B & E
\end{array}\right] G^{-1}\left[\begin{array}{c}
B^{\prime} P A+D_{2}^{\prime} C_{2}  \tag{4.2.22}\\
E^{\prime} P A+D_{22}^{\prime} C_{2}
\end{array}\right]
$$

The interest in this particular Riccati equation stems from the discretetime $H_{\infty}$ control theory (see Corollary 3.2.1). Also, it is simple to see that by letting $E=0$ and $D_{22}=0,(4.2 .18),(4.2 .19)$ and (4.2.20) reduce to the well-known Riccati equation from linear quadratic control theory. For clarity, we first recall the relation between the above Riccati equation and the discretetime full information feedback $H_{\infty}$ control problem. Let us define a system $\Sigma_{\text {FI }}$ by

$$
\Sigma_{\mathrm{FI}}:\left\{\begin{array}{rl}
x(k+1) & =A \quad x(k)+B u(k)+E \quad w(k),  \tag{4.2.23}\\
y(k) & =\binom{I}{0} x(k) \\
h(k) & \left.=C_{2} x(k)+D_{2} u(k)+\begin{array}{c}
0 \\
I
\end{array}\right) w(k), \\
D_{22} & w(k),
\end{array}\right.
$$

where $x \in \mathbb{R}^{n}$ is the state, $u \in \mathbb{R}^{m}$ is the control input, $w \in \mathbf{R}^{q}$ the disturbance input, $h \in \mathbb{R}^{\ell}$ the controlled output and $y \in \mathbb{R}^{n+q}$ the measurement. Then the following lemma follows from Corollary 3.2.1.

Lemma 4.2.1. Consider a given system (4.2.23). Assume that $\left(A, B, C_{2}, D_{2}\right)$ is left invertible and has no invariant zero on the unit circle. Then the following two statements are equivalent:

1. There exists a static feedback $u=K_{1} x+K_{2} w$, which stabilizes $\Sigma_{\text {FI }}$ and makes the $H_{\infty}$ norm of the closed-loop transfer function from $w$ to $h$ less than $\gamma$.
2. There exists a symmetric positive semi-definite solution $P$ to (4.2.18), (4.2.19) and (4.2.20) such that matrix $A_{\mathrm{cl}}$ of (4.2.22) has all its eigenvalues inside the unit circle.

In what follows, we provide a non-recursive method for computing the stabilizing solution to the $H_{\infty}$-DARE for the full information problem, i.e., (4.2.18), (4.2.19) and (4.2.20). We first define an auxiliary $H_{\infty}$-CARE from the given system data and we connect the stabilizing solution for the given $H_{\infty}$-DARE to the stabilizing solution for the auxiliary $H_{\infty}$-CARE, for which non-recursive methods of obtaining solutions are available.

We first choose any constant matrix $F$ such that $A+B F$ has no eigenvalue at -1 . We note that this can always be done as $(A, B)$ is stabilizable with respect to $\mathbb{C}^{\mathbf{O}} \cup \mathbb{C}^{\otimes}$. Next, define an auxiliary $H_{\infty}$-CARE,

$$
0=\tilde{P} \tilde{A}+\tilde{A}^{\prime} \tilde{P}+\tilde{C}_{2}^{\prime} \tilde{C}_{2}-\left[\begin{array}{c}
\tilde{B}^{\prime} \tilde{P}+\tilde{D}_{2}^{\prime} \tilde{C}_{2}  \tag{4.2.24}\\
\tilde{E}^{\prime} \tilde{P}+\tilde{D}_{22}^{\prime} \tilde{C}_{2}
\end{array}\right]^{\prime} \tilde{G}^{-1}\left[\begin{array}{c}
\tilde{B}^{\prime} \tilde{P}+\tilde{D}_{2}^{\prime} \tilde{C}_{2} \\
\tilde{E}^{\prime} \tilde{P}+\tilde{D}_{22}^{\prime} \tilde{C}_{2}
\end{array}\right]
$$

with the associated condition

$$
\begin{equation*}
\tilde{D}_{22}^{\prime}\left(I-\tilde{D}_{2}\left(\tilde{D}_{2}^{\prime} \tilde{D}_{2}\right)^{-1} \tilde{D}_{2}^{\prime}\right) \tilde{D}_{22}<\gamma^{2} I \tag{4.2.25}
\end{equation*}
$$

where

$$
\left.\begin{array}{rl}
\tilde{A} & :=(A+B F+I)^{-1}(A+B F-I)  \tag{4.2.26}\\
\tilde{B} & :=2(A+B F+I)^{-2} B \\
\tilde{E} & :=2(A+B F+I)^{-2} E \\
\tilde{C}_{2} & :=C_{2}+D_{2} F \\
\tilde{D}_{2} & :=D_{2}-C_{2}(A+B F+I)^{-1} B \\
\tilde{D}_{2} & :=D_{2}-C_{2}(A+B F+I)^{-1} E
\end{array}\right\}
$$

and

$$
\tilde{G}:=\left[\begin{array}{cc}
\tilde{D}_{2}^{\prime} \tilde{D}_{2} & \tilde{D}_{2}^{\prime} \tilde{D}_{22}  \tag{4.2.27}\\
\tilde{D}_{22}^{\prime} \tilde{D}_{2} & \tilde{D}_{22}^{\prime} \tilde{D}_{22}-\gamma^{2} I
\end{array}\right]
$$

If matrix $\tilde{D}_{2}$ is injective, then condition (4.2.25) implies $\tilde{G}$ in (4.2.27) is invertible. Again, we are particularly interested in solution $\tilde{P}$ of (4.2.24) such that the eigenvalues of $\tilde{A}_{\mathrm{cl}}$ are in the open-left plane, where

$$
\tilde{A}_{\mathrm{cl}}:=\tilde{A}-\left[\begin{array}{ll}
\tilde{B} & \tilde{E}
\end{array}\right] \tilde{G}^{-1}\left[\begin{array}{c}
\tilde{B}^{\prime} \tilde{P}+\tilde{D}_{2}^{\prime} \tilde{C}_{2}  \tag{4.2.28}\\
\tilde{E}^{\prime} \tilde{P}+\tilde{D}_{22}^{\prime} \tilde{C}_{2}
\end{array}\right]
$$

We note that under the conditions when $\tilde{D}_{2}$ is injective, $\left(\tilde{A}, \tilde{B}, \tilde{C}_{2}, \tilde{D}_{2}\right)$ has no invariant zero on the $j \omega$ axis, and (4.2.25), the above $H_{\infty}$-CARE (4.2.24) is related to the continuous-time $H_{\infty} \gamma$-suboptimal full information feedback control problem for the following system,

$$
\tilde{\Sigma}_{\mathrm{FI}}:\left\{\begin{array}{l}
\dot{x}=\tilde{A} x+\tilde{B} u+\tilde{E} w  \tag{4.2.29}\\
y=\binom{I}{0} x \\
h=\tilde{C}_{2} x+\tilde{D}_{2} u+\tilde{D}_{22} w \\
I
\end{array}\right) w,
$$

The following lemma follows from Corollary 3.1.2.
Lemma 4.2.2. Consider a given system (4.2.29). Assume that $\tilde{D}_{2}$ is injective and $\left(\tilde{A}, \tilde{B}, \tilde{C}_{2}, \tilde{D}_{2}\right)$ has no invariant zero on the $j \omega$ axis. Then the following two statements are equivalent:

1. There exists a static feedback law $u=\tilde{K}_{1} x+\tilde{K}_{2} w$, which stabilizes $\tilde{\Sigma}_{\mathrm{FI}}$ and makes the $H_{\infty}$ norm of the closed-loop transfer function from $w$ to $h$ less than $\gamma$.
2. Condition (4.2.25) holds and there exists a symmetric $\tilde{P} \geq 0$ such that (4.2.24) is satisfied and such that the matrix $\tilde{A}_{\mathrm{cl}}$ of (4.2.28) has all its eigenvalues in the open left-half plane.

Now, we are ready to present our main results.
Theorem 4.2.2. Assume that $A$ has no eigenvalue at -1 . Then the following two statements are equivalent:

1. $(A, B)$ is stabilizable and $\left(A, B, C_{2}, D_{2}\right)$ is left invertible with no invariant zero on the unit circle. Moreover, there exists a symmetric positive semidefinite matrix $P$ such that (4.2.18), (4.2.19) and (4.2.20) are satisfied along with the matrix $A_{\mathrm{cl}}$ of (4.2.22) having all its eigenvalues inside the unit circle.
2. $(\tilde{A}, \tilde{B})$ is stabilizable, $\tilde{D}_{2}$ is injective and $\left(\tilde{A}, \tilde{B}, \tilde{C}_{2}, \tilde{D}_{2}\right)$ has no invariant zero on the $j \omega$ axis, and (4.2.25) holds. Moreover, there exists a symmetric
positive semi-definite solution $\tilde{P}$ of the $H_{\infty}$-CARE (4.2.24) such that the eigenvalues of $\tilde{A}_{\mathrm{cl}}$, where $\tilde{A}_{\mathrm{cl}}$ is as in (4.2.28), are in the open left-half complex plane.

Furthermore, $P$ and $\tilde{P}$ are related by $P=2\left(A^{\prime}+I\right)^{-1} \tilde{P}(A+I)^{-1}$.
Proof. We note that the constant matrix $F$, a pre-state feedback, is introduced merely to overcome the situation when $A$ has eigenvalues at -1 . It is well-known in the literature that a pre-state feedback law does not affect the solution of the Riccati equation (4.2.20). Hence, for simplicity of presentation, we prove Theorem 4.2.2 for the case that $F=0$ and $\gamma=1$.
$\left(1 . \Rightarrow 2\right.$.) It follows from Lemma 4.1.2 that the quadruple $\left(\tilde{A}, \tilde{B}, \tilde{C}_{2}, \tilde{D}_{2}\right)$ is an inverse bilinear transformation of the quadruple ( $A, B, C_{2}, D_{2}$ ) with $a=1$. Hence, it follows from Theorem 4.1.2 that ( $\tilde{A}, \tilde{B}$ ) is stabilizable (see Item 1.a of Theorem 4.1.2) and ( $\tilde{A}, \tilde{B}, \tilde{C}_{2}, \tilde{D}_{2}$ ) is left invertible (see Item 3.b of Theorem 4.1.2) with no invariant zero on the $j \omega$ axis (see Item 4 of Theorem 4.1.2) and with no infinite zero of order higher than 0 (see Item 5 of Theorem 4.1.2). Hence, $\tilde{D}_{2}$ is injective as ( $A, B, C_{2}, D_{2}$ ) has no invariant zero at -1 .

Next, we will show that (4.2.25) holds. Let

$$
\left.\begin{array}{rl}
M & :=\left[\begin{array}{ll}
B & E
\end{array}\right], \\
N: & C_{2}^{\prime}\left[\begin{array}{ll}
D_{2} & D_{22}
\end{array}\right], \\
R & :=\left[\begin{array}{ll}
D_{2}^{\prime} D_{2} & D_{2}^{\prime} D_{22} \\
D_{22}^{\prime} D_{2} & D_{22}^{\prime} D_{22}-I
\end{array}\right], \\
Q:= & C_{2}^{\prime} C_{2}, \\
F & :=\tilde{A},  \tag{4.2.30}\\
G: & 2(A+I)^{-2} M, \\
H:=-Q(A+I)^{-1} M+N, \\
W:= & R+M^{\prime}\left(A^{\prime}+I\right)^{-1} Q(A+I)^{-1} M-N^{\prime}(A+I)^{-1} M \\
& -M^{\prime}\left(A^{\prime}+I\right)^{-1} N, \\
X:= & I-\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)(A+I)^{-1} M .
\end{array}\right\}
$$

It is simple to verify that

$$
W=\left[\begin{array}{cc}
\tilde{D}_{2}^{\prime} \tilde{D}_{2} & \tilde{D}_{2}^{\prime} \tilde{D}_{22} \\
\tilde{D}_{22}^{\prime} \tilde{D}_{2} & \tilde{D}_{22}^{\prime} \tilde{D}_{22}-I
\end{array}\right] .
$$

Then, (4.2.20) and (4.2.24) are, respectively, reduced to (4.2.1) and (4.2.3), and (4.2.22) and (4.2.28) can be written, respectively, as

$$
\begin{equation*}
A_{\mathrm{cl}}=A-M\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right), \tag{4.2.31}
\end{equation*}
$$

and

$$
\begin{equation*}
\tilde{A}_{\mathrm{cl}}=F-G W^{-1}(\tilde{P} G+H)^{\prime} \tag{4.2.32}
\end{equation*}
$$

Noting that

$$
\begin{aligned}
\operatorname{det}[X] & =\operatorname{det}\left[I-\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)(A+I)^{-1} M\right] \\
& =\operatorname{det}\left[I-M\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)(A+I)^{-1}\right] \\
& =\operatorname{det}\left[I+A_{\mathrm{cl}}\right] \cdot \operatorname{det}\left[(A+I)^{-1}\right],
\end{aligned}
$$

it follows that $X$ is nonsingular provided that the eigenvalues of $A_{\text {cl }}$ are inside the unit circle. Recalling (4.2.7) in the proof of Theorem 4.2.1, we have $W$ nonsingular and

$$
\begin{equation*}
W^{-1}=X^{-1}\left(R+M^{\prime} P M\right)^{-1}\left(X^{-1}\right)^{\prime} \tag{4.2.33}
\end{equation*}
$$

which implies that the inertia of $W^{-1}$ is equal to the inertia of $\left(R+M^{\prime} P M\right)^{-1}$ (see e.g., Theorem 4.9 of [3]). Again, noting that
$W^{-1}=\left[\begin{array}{cc}I & Y \\ 0 & I\end{array}\right]\left[\begin{array}{cc}\left(\tilde{D}_{2}^{\prime} \tilde{D}_{2}\right)^{-1} & 0 \\ 0 & {\left[\tilde{D}_{22}^{\prime}\left(I-\tilde{D}_{2}\left(\tilde{D}_{2}^{\prime} \tilde{D}_{2}\right)^{-1} \tilde{D}_{2}^{\prime}\right) \tilde{D}_{22}-I\right]^{-1}}\end{array}\right]\left[\begin{array}{cc}I & Y \\ 0 & I\end{array}\right]^{\prime}$,
and

$$
\left(R+M^{\prime} P M\right)^{-1}=\left[\begin{array}{cc}
I & Z \\
0 & I
\end{array}\right]\left[\begin{array}{cc}
V^{-1} & 0 \\
0 & -R^{-1}
\end{array}\right]\left[\begin{array}{cc}
I & Z \\
0 & I
\end{array}\right]^{\prime},
$$

where $Y=-\left(\tilde{D}_{2}^{\prime} \tilde{D}_{2}\right)^{-1} \tilde{D}_{2}^{\prime} \tilde{D}_{22}$ and $Z=-V^{-1} B^{\prime} P E$, together with (4.2.33) and the facts that $V>0$ and $R>0$, it follows that

$$
\tilde{D}_{22}^{\prime}\left(I-\tilde{D}_{2}\left(\tilde{D}_{2}^{\prime} \tilde{D}_{2}\right)^{-1} \tilde{D}_{2}^{\prime}\right) \tilde{D}_{22}<I
$$

Using the fact that $W$ is nonsingular, it follows from Theorem 4.2 . 1 that $\tilde{P}$ is a positive semi-definite solution of (4.2.24).

Finally, we are ready to prove that $\tilde{A}_{\mathrm{cl}}$ has all its eigenvalues in the open left-half complex plane. It follows from (4.2.10) in the proof of Theorem 4.2.1 that

$$
\begin{aligned}
\tilde{A}_{\mathrm{cl}}= & F-G W^{-1}(\tilde{P} G+H)^{\prime} \\
= & F-G X^{-1}\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right) \\
= & (A+I)^{-1}(A-I)-2(A+I)^{-2} M\left[I-\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)\right. \\
& \left.\quad \times(A+I)^{-1} M\right]^{-1}\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right) \\
= & (A+I)^{-1}\left\{A-I-2\left[I-(A+I)^{-1} M\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)\right]^{-1}\right. \\
& \left.\quad \times(A+I)^{-1} M\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)\right\}
\end{aligned}
$$

$$
\begin{align*}
= & (A+I)^{-1}\left\{A-I-2\left[I+A-M\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)\right]^{-1}\right. \\
& \left.\times M\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)\right\} \\
= & (A+I)^{-1}\left(A_{\mathrm{cl}}+I\right)^{-1}\left\{\left[I+A-M\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)\right]\right. \\
& \left.\times(A-I)-2 M\left(R+M^{\prime} P M\right)^{-1}\left(M^{\prime} P A+N^{\prime}\right)\right\} \\
= & (A+I)^{-1}\left(A_{\mathrm{cl}}+I\right)^{-1}\left(A_{\mathrm{cl}}-I\right)(A+I), \tag{4.2.34}
\end{align*}
$$

which implies that the eigenvalues of $\tilde{A}_{\text {cl }}$ are in the open left-half plane provided that the eigenvalues of $A_{\mathrm{cl}}$ are inside the unit circle.
$(2 . \Rightarrow 1$.$) First, following the results of Theorem 4.1.1, it is straightforward$ to show that $(A, B)$ is stabilizable and $\left(A, B, C_{2}, D_{2}\right)$ is left invertible with no invariant zero on the unit circle, provided that $(\tilde{A}, \tilde{B})$ is stabilizable, $\tilde{D}_{2}$ is injective and $\left(\tilde{A}, \tilde{B}, \tilde{C}_{2}, \tilde{D}_{2}\right)$ has no invariant zero on the $j \omega$ axis. Next, noting that

$$
\begin{aligned}
\operatorname{det}\left[I+W^{-1}(\tilde{P} G\right. & \left.+H)^{\prime}(I-F)^{-1} G\right] \\
& =\operatorname{det}\left[I+G W^{-1}(\tilde{P} G+H)^{\prime}(I-F)^{-1}\right] \\
& =\operatorname{det}\left[I-F+G W^{-1}(\tilde{P} G+H)^{\prime}\right] \cdot \operatorname{det}\left[(I-F)^{-1}\right] \\
& =\operatorname{det}\left[I-\tilde{A}_{\mathrm{cl}}\right] \cdot \operatorname{det}\left[(I-F)^{-1}\right]
\end{aligned}
$$

and $\tilde{A}_{\mathrm{cl}}$ has all its eigenvalue in the open left-half plane, it follows from (4.2.14) that $R+M^{\prime} P M$ is nonsingular. Thus, the condition in part 2 of Theorem 4.2.1 holds. The rest of the proof in the reverse direction of Theorem 4.2.2 follows from an almost identical procedure as ( $1 . \Rightarrow 2$.). This completes our proof. 因

Remark 4.2.1. We should point out that the left invertibility of ( $A, B, C_{2}, D_{2}$ ) is a necessary condition for the existence of the stabilizing solution to the $H_{\infty^{-}}$ DARE for the full information problem (see [100]). Moreover, following the proof of Theorem 4.2.2 and the properties of the continuous-time algebraic Riccati equation, it is easy to show that the condition that $\left(A, B, C_{2}, D_{2}\right)$ has no invariant zero on the unit circle is also necessary for the existence of the stabilizing solution to the $H_{\infty}$-DARE for the full information problem. 园

Remark 4.2.2. From Theorem 4.2.2, a non-iterative method of obtaining the stabilizing solution $P$ to the $H_{\infty}$-DARE for the full information problem can be established as follows:

1. Obtain the auxiliary $H_{\infty}$-CARE;
2. Obtain the stabilizing solution $\tilde{P}$ to the $H_{\infty}$-CARE using some wellknown non-iterative methods. For clarity, we recall in the following a so-called Schur method (see [55]): Define a Hamiltonian matrix

$$
H_{\mathrm{m}}=\left[\begin{array}{ll}
H_{11} & H_{12}  \tag{4.2.35}\\
H_{21} & H_{22}
\end{array}\right],
$$

where

$$
\left.\begin{array}{l}
H_{11}=\tilde{A}-\left[\begin{array}{ll}
\tilde{B} & \tilde{E}
\end{array}\right] \tilde{G}^{-1}\left[\begin{array}{ll}
\tilde{D}_{2} & \tilde{D}_{22}
\end{array}\right]^{\prime} \tilde{C}_{2}  \tag{4.2.36}\\
H_{12}=-\left[\begin{array}{ll}
\tilde{B} & \tilde{E}] \tilde{G}^{-1}\left[\begin{array}{ll}
\tilde{B} & \tilde{E}
\end{array}\right]^{\prime} \\
H_{21} & =-\tilde{C}_{2}^{\prime}\left\{I-\left[\begin{array}{lll}
\tilde{D}_{2} & \tilde{D}_{22}
\end{array}\right] \tilde{G}^{-1}\left[\begin{array}{ll}
\tilde{D}_{2} & \tilde{D}_{22}
\end{array}\right]^{\prime}\right\} \tilde{C}_{2} \\
H_{22} & =-\left\{\tilde{A}-\left[\begin{array}{lll}
\tilde{B} & \tilde{E}
\end{array}\right] \tilde{G}^{-1}\left[\begin{array}{ll}
\tilde{D}_{2} & \tilde{D}_{22}
\end{array}\right]^{\prime} \tilde{C}_{2}\right\}^{\prime}
\end{array}\right\}, ~
\end{array}\right\}
$$

Find an orthogonal matrix $T_{\mathrm{m}} \in \mathbb{R}^{2 n \times 2 n}$ that puts $H_{\mathrm{m}}$ in the real Schur form

$$
T_{\mathrm{m}}^{\prime} H_{\mathrm{m}} T_{\mathrm{m}}=\left[\begin{array}{cc}
S_{11} & S_{12}  \tag{4.2.37}\\
0 & S_{22}
\end{array}\right]
$$

where $S_{11} \in \mathbb{R}^{n \times n}$ is a stable matrix and $S_{22} \in \mathbb{R}^{n \times n}$ is an anti-stable matrix. Partition $T_{\mathrm{m}}$ into four $n \times n$ blocks:

$$
T_{\mathrm{m}}=\left[\begin{array}{ll}
T_{11} & T_{12}  \tag{4.2.38}\\
T_{21} & T_{22}
\end{array}\right]
$$

Then $\tilde{P}$ is given by $\tilde{P}=T_{21} T_{11}^{-1}$.
3. The stabilizing solution to the $H_{\infty}$-DARE for the full information problem is given by $P=2\left(A^{\prime}+I\right)^{-1} \tilde{P}(A+I)^{-1}$.

It is well-known that the $H_{\infty}$-DARE is the generalization of the $H_{2}$-DARE. Namely, by letting $\gamma=\infty$, or equivalently $E=0$ and $D_{2}=0$, we obtain the general $\mathrm{H}_{2}$-DARE. For completeness, we give the following corollary that provides a non-iterative method of solving the general $H_{2}$-DARE.

Corollary 4.2.1. Assume that $A$ has no eigenvalue at -1 . Then the following two statements are equivalent:

1. $(A, B)$ is stabilizable and $\left(A, B, C_{2}, D_{2}\right)$ is left invertible with no invariant zero on the unit circle. Moreover, there exists a positive semi-definite matrix $P$ such that

$$
\begin{equation*}
B^{\prime} P B+D_{2}^{\prime} D_{2}>0 \tag{4.2.39}
\end{equation*}
$$

$$
\begin{equation*}
P=A^{\prime} P A+C_{2}^{\prime} C_{2}-\left(A^{\prime} P B+C_{2}^{\prime} D_{2}\right)\left(D_{2}^{\prime} D_{2}+B^{\prime} P B\right)^{-1}\left(A^{\prime} P B+C_{2}^{\prime} D_{2}\right)^{\prime} \tag{4.2.40}
\end{equation*}
$$

and such that the eigenvalues of the matrix $A_{\mathrm{cl}}$ are inside the unit circle, where

$$
\begin{equation*}
A_{\mathrm{cl}}=A-B\left(D_{2}^{\prime} D_{2}+B^{\prime} P B\right)^{-1}\left(A^{\prime} P B+C_{2}^{\prime} D_{2}\right)^{\prime} \tag{4.2.41}
\end{equation*}
$$

2. $(\tilde{A}, \tilde{B})$ is stabilizable, $\tilde{D}_{2}$ is injective and ( $\left.\tilde{A}, \tilde{B}, \tilde{C}_{2}, \tilde{D}_{2}\right)$ has no invariant zero on the $j \omega$ axis. Moreover, there exists a positive semi-definite solution $\tilde{P}$ of the following CARE
$0=\tilde{P} \tilde{A}+\tilde{A}^{\prime} \tilde{P}+\tilde{C}_{2}^{\prime} \tilde{C}_{2}-\left(\tilde{P} \tilde{B}+\tilde{C}_{2}^{\prime} \tilde{D}_{2}\right)\left(\tilde{D}_{2}^{\prime} \tilde{D}_{2}\right)^{-1}\left(\tilde{P} \tilde{B}+\tilde{C}_{2}^{\prime} \tilde{D}_{2}\right)^{\prime}$,
such that the eigenvalues of $\tilde{A}_{\mathrm{cl}}$ are in the open left-half complex plane, where

$$
\begin{equation*}
\tilde{A}_{\mathrm{cl}}=\tilde{A}-\tilde{B}\left(\tilde{D}_{2}^{\prime} \tilde{D}_{2}\right)^{-1}\left(\tilde{P} \tilde{B}+\tilde{C}_{2}^{\prime} \tilde{D}_{2}\right)^{\prime} \tag{4.2.43}
\end{equation*}
$$

Furthermore, $P$ and $\tilde{P}$ are related by $P=2\left(A^{\prime}+I\right)^{-1} \tilde{P}(A+I)^{-1}$.
Lemmas 4.2.1 and 4.2.2, and Theorem 4.2.2 show the interconnection between the $H_{\infty} \gamma$-suboptimal control problem for the discrete-time system $\Sigma_{\text {FI }}$ and the continuous-time system $\tilde{\Sigma}_{\mathrm{FI}}$. This connection is formalized in the following lemma.

Lemma 4.2.3. Assume that $(A, B)$ is stabilizable and $\left(A, B, C_{2}, D_{2}\right)$ is left invertible with no invariant zero on the unit circle. Then the following statements are equivalent:

1. The full information feedback discrete-time system $\Sigma_{\text {FI }}$ of (4.2.23) has at least one $\gamma$-suboptimal control law. Namely, for a given $\gamma$, there exists a static full information feedback $u=K_{1} x+K_{2} w$ such that the closed-loop transfer function from $w$ to $h$ has an $H_{\infty}$-norm less than $\gamma$.
2. The full information feedback continuous-time system $\tilde{\Sigma}_{\text {FI }}$ of (4.2.29) has at least one $\gamma$-suboptimal control law. Namely, for a given $\gamma$, there exists a static full information feedback $u=\tilde{K}_{1} x+\tilde{K}_{2} w$ such that the closed-loop transfer function from $w$ to $h$ has an $H_{\infty}$-norm less than $\gamma$.

Remark 4.2.3. The results of Lemma 4.2 .3 can easily be obtained by a different route. It is well known that the Hankel norm and the $H_{\infty}$ norm of a transfer function are invariant under a bilinear transformation (see e.g., [45]).

Hence one can re-cast the $H_{\infty} \gamma$-suboptimal control problem for the discretetime system $\Sigma_{\text {FI }}$ into an equivalent $H_{\infty} \gamma$-suboptimal control problem for an auxiliary continuous-time system obtained by performing bilinear transformation on $\Sigma_{\mathrm{FI}}$. It can be shown that one of the state space realizations of this auxiliary continuous-time system, $\Sigma_{\mathrm{BL}}$, is given by
where $\tilde{D}_{3}=-(A+I)^{-1} B, \tilde{D}_{4}=-(A+I)^{-1} E$, and $\tilde{A}, \tilde{B}, \tilde{E}, \tilde{C}_{2}, \tilde{D}_{2}$ and $\tilde{D}_{22}$ are as defined in (4.2.26). Consequently the $H_{\infty} \gamma$-suboptimal control problem for the discrete-time $\Sigma_{\text {FI }}$ has a solution if and only if the $H_{\infty} \gamma$-suboptimal control problem for the continuous-time system $\Sigma_{\mathrm{BL}}$ has a solution. However, we note that $\Sigma_{\text {BL }}$ is not completely in the full information form. This difficulty can easily be removed by redefining the measurement output in $\Sigma_{\text {BL }}$ as

$$
\tilde{y}:=\left[\begin{array}{cc}
I & -\tilde{D}_{4}  \tag{4.2.45}\\
0 & I
\end{array}\right]\left(y-\left[\begin{array}{c}
\tilde{D}_{3} \\
0
\end{array}\right] u\right)=\binom{I}{0} x+\binom{0}{I} w
$$

It is now obvious that $\Sigma_{\text {BL }}$ with the new measurement output $\tilde{y}$ is in fact the same as $\tilde{\Sigma}_{\mathrm{FI}}$. Also, it is easy to show that the $H_{\infty} \gamma$-suboptimal problem for $\Sigma_{\mathrm{BL}}$ has a solution if and only if the $H_{\infty} \gamma$-suboptimal problem for $\Sigma_{\text {FI }}$ has a solution and hence the result of Lemma 4.2 .3 follows. It is important to note that the bilinear transformation approach does not establish a relationship between the stabilizing solution of the $H_{\infty}$-CARE associated with the continuous-time system $\tilde{\Sigma}_{\mathrm{FI}}$, obtained by performing a bilinear transformation on discrete-time system $\Sigma_{\text {FI }}$ and defining the new measurement as in (4.2.45), and the $H_{\infty^{-}}$ DARE associated with the given discrete-time system $\Sigma_{\mathrm{FI}}$. In fact, the main contribution of Theorem 4.2.2 is to establish such a relationship.

We present in the following a numerical example to illustrate our results.

Example 4.2.1. Let us consider a discrete-time $H_{\infty}$-DARE for the full information problem with

$$
A=\left[\begin{array}{rrrrr}
1 & 1 & 0 & 1 & 0  \tag{4.2.46}\\
1 & 1 & 1 & 0 & 0 \\
0 & 1 & 0 & 1 & 0 \\
0 & 0 & 1 & 1 & 1 \\
0 & -1 & 0 & 1 & 1
\end{array}\right], B=\left[\begin{array}{ll}
1 & 0 \\
0 & 1 \\
1 & 0 \\
0 & 1 \\
1 & 0
\end{array}\right], E=\left[\begin{array}{r}
1 \\
0 \\
-1 \\
0 \\
1
\end{array}\right]
$$

$$
C_{2}=\left[\begin{array}{lllll}
0 & 0 & 0 & 0 & 0  \tag{4.2.47}\\
1 & 0 & 1 & 0 & 1 \\
0 & 1 & 0 & 1 & 0
\end{array}\right], D_{2}=\left[\begin{array}{ll}
1 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right], D_{22}=\left[\begin{array}{c}
0 \\
0 \\
0.5
\end{array}\right]
$$

and $\gamma=1$. It is simple to verify that $\left(A, B, C_{2}, D_{2}\right)$ is left invertible with an invariant zero at 0 . Following (4.2.26), we obtain the auxiliary $H_{\infty}$ - CARE with

$$
\begin{gathered}
\tilde{A}=\left[\begin{array}{rrrrr}
1 & -2 & 6 & -4 & 2 \\
-1 & 3 & -8 & 6 & -3 \\
2 & -4 & 11 & -8 & 4 \\
-1 & 2 & -4 & 3 & -1 \\
0 & 0 & -2 & 2 & -1
\end{array}\right], \quad \tilde{B}=\left[\begin{array}{rr}
68 & -50 \\
-92 & 68 \\
128 & -94 \\
-52 & 38 \\
-18 & 14
\end{array}\right], \quad \tilde{E}=\left[\begin{array}{r}
-20 \\
28 \\
-40 \\
16 \\
6
\end{array}\right], \\
\tilde{C}_{2}=\left[\begin{array}{rrrrr}
0 & 0 & 0 & 0 & 0 \\
1 & 0 & 1 & 0 & 1 \\
0 & 1 & 0 & 1 & 0
\end{array}\right], \quad \tilde{D}_{2}=\left[\begin{array}{rr}
1 & 0 \\
10 & -8 \\
-9 & 6
\end{array}\right], \quad \tilde{D}_{22}=\left[\begin{array}{r}
0.0 \\
-4.0 \\
3.5
\end{array}\right] .
\end{gathered}
$$

Solving (4.2.3) in MATLAB, we obtain the stabilizing solution to the auxiliary $H_{\infty}$-CARE as

$$
\tilde{P}=10^{3} \times\left[\begin{array}{rrrrr}
0.767767 & 1.110081 & 0.180720 & -0.307296 & -0.617828 \\
1.110081 & 1.607297 & 0.260775 & -0.448623 & -0.897322 \\
0.180720 & 0.260775 & 0.046343 & -0.064704 & -0.139318 \\
-0.307296 & -0.448623 & -0.064704 & 0.143150 & 0.264285 \\
-0.617828 & -0.897322 & -0.139318 & 0.264285 & 0.511644
\end{array}\right],
$$

and the stabilizing solution to the $H_{\infty}$-DARE for the full information problem is given by,

$$
P=\left[\begin{array}{ccccc}
127.143494 & 187.057481 & 1 & -84.671880 & -134.864680 \\
187.057481 & 278.730887 & 0 & -124.061419 & -201.396153 \\
1 & 0 & 1 & 0 & 1 \\
-84.671880 & -124.061419 & 0 & 61.078015 & 92.569717 \\
-134.864680 & -201.396153 & 1 & 92.569717 & 147.982935
\end{array}\right]
$$

It is straightforward to verify that the above $P$ satisfies (4.2.18), (4.2.19) and (4.2.20). Moreover, the eigenvalues of $A_{\mathrm{cl}}$ are given by $\{0.4125 \pm j 0.0733,0,0,0\}$, which are inside the unit circle.

## Chapter 5

## Infima in Continuous-time $H_{\infty}$ Optimization

In THIS CHAPTER, we address the problem of computing infima in $H_{\infty}$ optimization for continuous-time systems. The $H_{\infty}$-CARE based approach to this problem simply provides an iterative scheme of approximating the infimum, $\gamma^{*}$, of the $H_{\infty}$-norm of the closed-loop transfer function. For example, in the regular measurement feedback case and utilizing the results of Doyle et al [39] (see also Corollary 3.1.1), an iterative procedure for approximating $\gamma^{*}$ would proceed as follows: one starts with a value of $\gamma$ and determines whether $\gamma>\gamma^{*}$ by solving two "indefinite" algebraic Riccati equations and checking the positive semi-definiteness and stabilizing properties of these solutions. In the case when such positive semi-definite solutions exist and satisfy a coupling condition, then we have $\gamma>\gamma^{*}$ and one simply repeats the above steps using a smaller value of $\gamma$. In principle, one can approximate the infimum $\gamma^{*}$ to within any degree of accuracy in this manner. However this search procedure is exhaustive and can be very costly. More significantly, due to the possible high-gain occurrence as $\gamma$ gets close to $\gamma^{*}$, numerical solutions for these $H_{\infty}$-CARE's can become highly sensitive and ill-conditioned. This difficulty also arises in the coupling condition. Namely, as $\gamma$ decreases, evaluation of the coupling condition would generally involve finding eigenvalues of stiff matrices. These numerical difficulties are likely to be more severe for problems associated with the singular case. So in general the iterative procedure for the computation of $\gamma^{*}$ based on ARE's is not reliable.

Our goal here is to develop non-iterative procedures to compute exactly the value of $\gamma^{*}$ for a fairly large class of systems, which are associated with the singular case and satisfy certain geometric conditions. The computation of $\gamma^{*}$
in our procedure involves solving two well-defined Riccati and two Lyapunov equations, which are independent of $\gamma$. The algorithm has been implemented efficiently in a MATLAB-software environment for numerical solutions. The results of this chapter are based on those reported in Chen [14] and Chen et al [19,23-25].

The outline of this chapter is as follows: In Section 5.1, we will present a non-iterative algorithm that computes the infimum, $\gamma^{*}$, for the continuous-time $H_{\infty}$ optimization problem under full information feedback, which is equivalent to that under full state feedback if the direct feedthrough term from the disturbance to the controlled output is equal to zero. Section 5.2 deals with the computation of $\gamma^{*}$ for the measurement feedback case. Both Sections 5.1 and 5.2 require the given systems having no invariant zero on the imaginary axis and satisfying certain geometric conditions. Finally, in Section 5.3, we will remove the constraints on the imaginary axis invariant zeros, i.e., we will present a non-iterative computational algorithm for finding $\gamma^{*}$ for systems with invariant zeros on the imaginary axis.

### 5.1. Full Information Feedback Case

We consider in this section the $H_{\infty}$ optimization problem for the class of continuous-time systems characterized by

$$
\left\{\begin{array}{l}
\dot{x}=A x+B u+E \quad w,  \tag{5.1.1}\\
y=\binom{I}{0} x+\binom{0}{I} w \\
h=C_{2} x+D_{2} u+D_{22} w
\end{array}\right.
$$

where $x \in \mathbb{R}^{n}$ is the state, $u \in \mathbb{R}^{m}$ is the control input, $w \in \mathbb{R}^{q}$ is the external disturbance input, $y \in \mathbb{R}^{n+q}$ is the measurement output, and $h \in \mathbb{R}^{\ell}$ is the controlled output of $\Sigma$. It is labelled a full information problem in the literature because all information about the system, i.e., both $x$ and $w$, are available for feedback. For the purpose of easy reference in future developments, we define $\Sigma_{\mathrm{P}}$ to be the subsystem characterized by the matrix quadruple ( $A, B, C_{2}, D_{2}$ ).

We first make the following assumptions:
Assumption 5.F.1: $(A, B)$ is stabilizable;
Assumption 5.F.2: $\Sigma_{\mathbf{P}}$ has no invariant zero on the imaginary axis;
Assumption 5.F.3: $\operatorname{Im}(E) \subset \mathcal{V}^{-}\left(\Sigma_{\mathrm{P}}\right)+\mathcal{S}^{-}\left(\Sigma_{\mathrm{P}}\right)$; and
Assumption 5.F.4: $D_{22}=0$.

Remark 5.1.1. Here we note that the first assumption, i.e., $(A, B)$ is stabilizable, is necessary for the existence of any stabilizing controller. The second assumption will be removed in Section 5.3. Also, Assumption 5.F. 3 will be automatically satisfied if $\Sigma_{P}$ is right invertible. In fact, in this case, Assumption 5.F. 4 will be no longer necessary. This will be treated as a special case at the end of this section (see Remark 5.1.4).

We have the following non-iterative algorithm for computing the infimum, $\gamma^{*}$, of the full information system (5.1.1).

Step 5.F.1. Without loss of generality, we assume that $\left(A, B, C_{2}, D_{2}\right)$, i.e., $\Sigma_{\mathrm{P}}$, has been partitioned in the form of (2.3.4). Then, transform $\Sigma_{P}$ into the special coordinate basis as described in Chapter 2 (see also (2.3.20) to (2.3.23) for the compact form of the special coordinate basis). In this algorithm, for easy reference in future developments, we introduce an additional permutation matrix to the state transformation $\Gamma_{s}$ such that the new state variables are ordered as follows:

$$
\tilde{x}=\left(\begin{array}{c}
x_{a}^{+}  \tag{5.1.2}\\
x_{b} \\
x_{a}^{-} \\
x_{c} \\
x_{d}
\end{array}\right)
$$

We also choose the output transformation $\Gamma_{o}$ to have the following form:

$$
\Gamma_{o}=\left[\begin{array}{cc}
I_{m_{0}} & 0  \tag{5.1.3}\\
0 & \Gamma_{o r}
\end{array}\right]
$$

where $m_{0}=\operatorname{rank}\left(D_{2}\right)$. Next, we compute

$$
\Gamma_{s}^{-1} E=\left[\begin{array}{c}
E_{a}^{+}  \tag{5.1.4}\\
E_{b} \\
E_{a}^{-} \\
E_{c} \\
E_{d}
\end{array}\right]
$$

It is simple to verify from the properties of the special coordinate basis that Assumption 5.F. 3 is equivalent to $E_{b}=0$. Also, for economy of notation, we denote $n_{x}$ the dimension of $\mathbb{R}^{n} / \mathcal{S}^{+}\left(\Sigma_{\mathrm{P}}\right)$, which is equivalent to $n_{a}^{+}+n_{b}$. We note that $n_{x}=0$ if and only if the system $\Sigma_{\mathrm{P}}$ is right invertible and is of minimum phase.

Step 5.F.2. Next, we define

$$
\begin{gather*}
A_{11}:=\left[\begin{array}{cc}
A_{a a}^{+} & L_{a b}^{+} C_{b} \\
0 & A_{b b}
\end{array}\right], B_{11}:=\left[\begin{array}{c}
B_{0 a}^{+} \\
B_{0 b}
\end{array}\right], A_{13}:=\left[\begin{array}{c}
L_{a d}^{+} \\
L_{b d}
\end{array}\right],  \tag{5.1.5}\\
C_{21}:=\Gamma_{o r}\left[\begin{array}{cc}
0 & 0 \\
0 & C_{b}
\end{array}\right], \quad C_{23}:=\Gamma_{o r}\left[\begin{array}{c}
C_{d} C_{d}^{\prime} \\
0
\end{array}\right] \tag{5.1.6}
\end{gather*}
$$

and

$$
\begin{gather*}
A_{x}:=A_{11}-A_{13}\left(C_{23}^{\prime} C_{23}\right)^{-1} C_{23}^{\prime} C_{21},  \tag{5.1.7}\\
B_{x} B_{x}^{\prime}:=B_{11} B_{11}^{\prime}+A_{13}\left(C_{23}^{\prime} C_{23}\right)^{-1} A_{13}^{\prime}  \tag{5.1.8}\\
C_{x}^{\prime} C_{x}:=C_{21}^{\prime} C_{21}-C_{21}^{\prime} C_{23}\left(C_{23}^{\prime} C_{23}\right)^{-1} C_{23}^{\prime} C_{21} \tag{5.1.9}
\end{gather*}
$$

Then we solve for the positive definite solution $S_{x}$ of the algebraic Riccati equation,

$$
\begin{equation*}
A_{x} S_{x}+S_{x} A_{x}^{\prime}-B_{x} B_{x}^{\prime}+S_{x} C_{x}^{\prime} C_{x} S_{x}=0 \tag{5.1.10}
\end{equation*}
$$

together with the matrix $T_{x}$ defined by

$$
T_{x}:=\left[\begin{array}{cc}
T_{a x} & 0  \tag{5.1.11}\\
0 & 0
\end{array}\right]
$$

where $T_{a x}$ is the unique solution of the algebraic Lyapunov equation,

$$
\begin{equation*}
A_{a a}^{+} T_{a x}+T_{a x}\left(A_{a a}^{+}\right)^{\prime}=E_{a}^{+}\left(E_{a}^{+}\right)^{\prime} \tag{5.1.12}
\end{equation*}
$$

Here we should note that $\left(-A_{x}, C_{x}\right)$ is detectable since $-A_{a a}^{+}$is stable and $\left(A_{b b}, C_{b}\right)$ is observable. Furthermore, Assumption 5.F. 1 implies that $\left(A_{x}, B_{x}\right)$ is stabilizable. Hence the existence and uniqueness of the solutions $S_{x}$ and $T_{a x}$ follow from the results of Richardson and Kwong [83].

Step 5.F.3. The infimum, $\gamma^{*}$, is given by

$$
\begin{equation*}
\gamma^{*}=\sqrt{\lambda_{\max }\left(T_{x} S_{x}^{-1}\right)} \tag{5.1.13}
\end{equation*}
$$

It can be shown using the result of Wielandt [110] that all the eigenvalues of $T_{x} S_{x}^{-1}$ are real and nonnegative.

We have the following theorem.
Theorem 5.1.1. Consider the full information system given by (5.1.1). Then under Assumptions 5.F. 1 to 5.F.4,

1. $\gamma^{*}$ given by (5.1.13) is indeed its infimum, and
2. for $\gamma>\gamma^{*}$, the positive semi-definite matrix $P(\gamma)$ given by

$$
P(\gamma)=\left(\Gamma_{s}^{-1}\right)^{\prime}\left[\begin{array}{cc}
\left(S_{x}-T_{x} / \gamma^{2}\right)^{-1} & 0  \tag{5.1.14}\\
0 & 0
\end{array}\right] \Gamma_{s}^{-1}
$$

is the unique solution that satisfies conditions 2.(a)-2.(c) of Theorem 3.1.1. Moreover, such a solution $P(\gamma)$ does not exist when $\gamma<\gamma^{*}$.

Proof. As stated in Step 5.F. 1 of the algorithm, we assume that $\Sigma_{\mathrm{P}}$ has been partitioned as in (2.3.4). Hence, the full information system of (5.1.1) can be rewritten as

$$
\left\{\begin{array}{c}
\dot{x}=A \quad x+\left[\begin{array}{ll}
B_{0} & B_{1}
\end{array}\right]\binom{u_{0}}{u_{1}}+\quad E \quad w  \tag{5.1.15}\\
\binom{h_{0}}{h_{1}}=\left[\begin{array}{l}
C_{2,0} \\
C_{2,1}
\end{array}\right] x+\left[\begin{array}{cc}
I_{m_{0}} & 0 \\
0 & 0
\end{array}\right]\binom{u_{0}}{u_{1}}+\left[\begin{array}{l}
D_{22,0} \\
D_{22,1}
\end{array}\right] w
\end{array}\right.
$$

where in this proof, we consider both $D_{22,0}=0$ and $D_{22,1}=0$. Let us apply a pre-feedback law,

$$
\begin{equation*}
u_{0}=-C_{2,0} x+v_{0} \tag{5.1.16}
\end{equation*}
$$

to the above system. Then it is trivial to write the new system as,

$$
\left\{\begin{align*}
\dot{x} & =\left(A-B_{0} C_{2,0}\right) x+\left[\begin{array}{ll}
B_{0} & B_{1}
\end{array}\right]\binom{v_{0}}{u_{1}}+E w  \tag{5.1.17}\\
\binom{h_{0}}{h_{1}} & =\left[\begin{array}{c}
0 \\
C_{2,1}
\end{array}\right] \quad x+\left[\begin{array}{cc}
I_{m_{0}} & 0 \\
0 & 0
\end{array}\right]\binom{v_{0}}{u_{1}}
\end{align*}\right.
$$

It follows from the theorem of the special coordinate basis, i.e., Theorem 2.3.1, that there exist non-singular transformations, $\Gamma_{s}, \Gamma_{o}$ and $\Gamma_{i}$ such that

$$
\binom{v_{0}}{u_{1}}=\Gamma_{i}\left(\begin{array}{c}
v_{0} \\
u_{d} \\
u_{c}
\end{array}\right), \quad x=\Gamma_{a}\left(\begin{array}{c}
x_{a}^{+} \\
x_{b} \\
x_{a}^{-} \\
x_{c} \\
x_{d}
\end{array}\right), \quad\binom{h_{0}}{h_{1}}=\Gamma_{o}\left(\begin{array}{c}
h_{0} \\
h_{d} \\
h_{b}
\end{array}\right)
$$

By Assumption 5.F.2, i.e., $\Sigma_{\mathrm{P}}$ has no invariant zero on the imaginary axis, the state component $x_{a}^{0}$ is nonexistent and the transformed system is given by

$$
\left(\begin{array}{c}
\dot{x}_{a}^{+} \\
\dot{x}_{b} \\
\dot{x}_{a}^{-} \\
\dot{x}_{c} \\
\dot{x}_{d}
\end{array}\right)=\left[\begin{array}{ccccc}
A_{a a}^{+} & L_{a b}^{+} C_{b} & 0 & 0 & L_{a d}^{+} C_{d} \\
0 & A_{b b} & 0 & 0 & L_{b d} C_{d} \\
0 & L_{a b}^{-} C_{b} & A_{a a}^{-a} & 0 & L_{a d}^{-} C_{d} \\
B_{c} E_{c a}^{+} & L_{c b} C_{b} & B_{c} E_{c a}^{-} & A_{c c} & L_{c d} C_{d} \\
B_{d} E_{d a}^{+} & B_{d} E_{d b} & B_{d} E_{d a}^{-} & B_{d} E_{d c} & A_{d d}
\end{array}\right]\left(\begin{array}{c}
x_{a}^{+} \\
x_{b} \\
x_{a}^{-} \\
x_{c} \\
x_{d}
\end{array}\right)
$$

$$
+\left[\begin{array}{ccc}
B_{0 a}^{+} & 0 & 0  \tag{5.1.18}\\
B_{0 b} & 0 & 0 \\
B_{0 a}^{-} & 0 & 0 \\
B_{0 c} & 0 & B_{c} \\
B_{0 d} & B_{d} & 0
\end{array}\right]\left(\begin{array}{c}
v_{0} \\
u_{d} \\
u_{c}
\end{array}\right)+\left[\begin{array}{c}
E_{a}^{+} \\
E_{b} \\
E_{a}^{-} \\
E_{c} \\
E_{d}
\end{array}\right] w
$$

where $E_{b}=0$, and

$$
\left(\begin{array}{l}
h_{0}  \tag{5.1.19}\\
h_{d} \\
h_{b}
\end{array}\right)=\left[\begin{array}{cc}
I_{m_{0}} & 0 \\
0 & \Gamma_{o r}
\end{array}\right]\left[\begin{array}{ccccc}
0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & C_{d} \\
0 & C_{b} & 0 & 0 & 0
\end{array}\right]\left(\begin{array}{c}
x_{a}^{+} \\
x_{b} \\
x_{a}^{-} \\
x_{c} \\
x_{d}
\end{array}\right)+\left[\begin{array}{ccc}
I_{m_{0}} & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right]\left(\begin{array}{c}
v_{0} \\
u_{d} \\
u_{c}
\end{array}\right)
$$

The above transformation of the system with a pre-state feedback law,

$$
u_{0}=-C_{2,0} x+v_{0}
$$

along with the non-singular state and control input transformations does not change our solution since it does not affect the value of $\gamma^{*}$. We need to introduce the following lemmas in order to prove the theorem.

Lemma 5.1.1. Given the system of (5.1.1), which satisfies Assumptions 5.F.1, 5.F. 2 and 5.F.4, and $\gamma>0$, then there exists a full information feedback control law $u=F_{1} x+F_{2} w$ such that the resulting $\left\|T_{h w}\right\|_{\infty}<\gamma$ and $\lambda(A+B F) \subset \mathbb{C}^{-}$ if and only if there exists a real symmetric solution $P_{x}>0$ to the algebraic Riccati equation

$$
\begin{equation*}
P_{x} A_{x}+A_{x}^{\prime} P_{x}+P_{x} E_{x} E_{x}^{\prime} P_{x} / \gamma^{2}-P_{x} B_{x} B_{x}^{\prime} P_{x}+C_{x}^{\prime} C_{x}=0 \tag{5.1.20}
\end{equation*}
$$

where $A_{x}, B_{x}$ and $C_{x}$ are as defined in (5.1.7) to (5.1.9), and

$$
E_{x}=\left[\begin{array}{c}
E_{a}^{+}  \tag{5.1.21}\\
E_{b}
\end{array}\right]
$$

with no restriction on $E_{b}$. Note that $E_{b}=0$ if Assumption 5.F. 3 holds.
Proof. Without loss of generality, we assume that the given system has been transformed into the form of (5.1.18) and (5.1.19). Now let us define the new state variables,

$$
x_{1}:=\binom{x_{a}^{+}}{x_{b}}, \quad\binom{x_{2}}{x_{3}}:=\left(\begin{array}{c}
x_{a}^{-}  \tag{5.1.22}\\
x_{c} \\
x_{d}
\end{array}\right)
$$

where $x_{3}$ contains only the $m_{d}$ states of $x_{d}$ which are directly associated with the controlled output $h_{d}$ while $x_{2}$ contains $x_{a}^{-}, x_{c}$ and the remaining states of
$x_{d}$. Hence, the dynamics of the transformed system in (5.1.18) and (5.1.19) can be partitioned as follows,

$$
\begin{gather*}
\dot{x}_{1}=A_{11} x_{1}+\left[\begin{array}{ll}
B_{11} & A_{13}
\end{array}\right]\binom{v_{0}}{x_{3}}+E_{x} w,  \tag{5.1.23}\\
\binom{\dot{x}_{2}}{\dot{x}_{3}}=\left[\begin{array}{ll}
A_{22} & A_{23} \\
A_{32} & A_{33}
\end{array}\right]\binom{x_{2}}{x_{3}}+\left[\begin{array}{l}
B_{22} \\
B_{32}
\end{array}\right] u_{1}+\left[\begin{array}{ll}
B_{21} & A_{21} \\
B_{31} & A_{31}
\end{array}\right]\binom{v_{0}}{x_{1}}+\left[\begin{array}{c}
E_{2} \\
E_{3}
\end{array}\right] w \\
\binom{h_{0}}{h_{1}}=\left[\begin{array}{c}
0 \\
C_{21}
\end{array}\right] x_{1}+\left[\begin{array}{cc}
I_{m_{0}} & 0 \\
0 & C_{23}
\end{array}\right]\binom{v_{0}}{x_{3}}, \tag{5.1.24}
\end{gather*}
$$

where $A_{11}, B_{11}, A_{13}, C_{21}$ and $C_{23}$ are as defined in (5.1.5) to (5.1.6), while $A_{22}, A_{23}, \cdots, E_{3}$ are the matrices with appropriate dimensions. It is now straightforward to verify using the properties of the special coordinate basis that the quadruple characterized by

$$
\left(\left[\begin{array}{ll}
A_{22} & A_{23}  \tag{5.1.26}\\
A_{32} & A_{33}
\end{array}\right],\left[\begin{array}{l}
B_{22} \\
B_{32}
\end{array}\right],\left[\begin{array}{ll}
0 & I
\end{array}\right], 0\right)
$$

is right invertible and of minimum phase. Moreover, the state space $\mathcal{X}_{2} \oplus \mathcal{X}_{3}$ spans the strongly controllable subspace $\mathcal{S}^{+}\left(\Sigma_{\mathrm{P}}\right)$. On the other hand, the subsystem characterized by the quadruple

$$
\left(A_{11}, B_{11},\left[\begin{array}{c}
0  \tag{5.1.27}\\
C_{21}
\end{array}\right],\left[\begin{array}{cc}
I_{m_{0}} & 0 \\
0 & C_{23}
\end{array}\right]\right)
$$

is left invertible with no infinite zero and with no stable invariant zero. The result of Lemma 5.1.1 follows from Corollary 5.2 and Theorem 6.2 of [104]. 区

Lemma 5.1.2. Given the system of (5.1.1) which satisfies Assumptions 5.F. 1 to 5.F.4, then the algebraic Riccati equation of (5.1.20) has a symmetric solution $P_{x}>0$ if and only if $S_{x}>T_{x} / \gamma^{2}$, where $S_{x}$ and $T_{x}$ are respectively given by (5.1.10) and (5.1.11).

Proof. First, we note that $T_{x}$ of (5.1.11) is in fact the solution to the following Lyapunov equation

$$
\begin{equation*}
A_{x} T_{x}+T_{x} A_{x}^{\prime}=E_{x} E_{x} \tag{5.1.28}
\end{equation*}
$$

where

$$
E_{x}=\left[\begin{array}{c}
E_{a}^{+} \\
0
\end{array}\right]
$$

since Assumption 5.F. 3 holds. Also note that

$$
\begin{equation*}
T_{x} C_{21}=0 \quad \text { and } \quad T_{x} C_{x}^{\prime} C_{x} T_{x}=0 \tag{5.1.29}
\end{equation*}
$$

Now, suppose that $S_{x}>T_{x} / \gamma^{2}$ and define a positive definite matrix,

$$
X:=S_{x}-T_{x} / \gamma^{2}
$$

It follows from (5.1.10), (5.1.28) and (5.1.29) that

$$
\begin{equation*}
A_{x} X+X A_{x}^{\prime}+E_{x} E_{x}^{\prime} / \gamma^{2}-B_{x} B_{x}^{\prime}+X C_{x}^{\prime} C_{x} X=0 \tag{5.1.30}
\end{equation*}
$$

Now, let us pre- and post-multiply (5.1.30) by $P_{x}:=X^{-1}$, we obtain

$$
\begin{equation*}
P_{x} A_{x}+A_{x}^{\prime} P_{x}+P_{x} E_{x} E_{x}^{\prime} P_{x} / \gamma^{2}-P_{x} B_{x} B_{x}^{\prime} P_{x}+C_{x}^{\prime} C_{x}=0 \tag{5.1.31}
\end{equation*}
$$

Hence, $P_{x}>0$ is a solution to (5.1.20).
Conversely, suppose that (5.1.20) has a solution $P_{x}>0$. Let $X:=P_{x}^{-1}>0$. We have

$$
\begin{equation*}
A_{x} X+X A_{x}^{\prime}+E_{x} E_{x}^{\prime} / \gamma^{2}-B_{x} B_{x}^{\prime}+X C_{x}^{\prime} C_{x} X=0 \tag{5.1.32}
\end{equation*}
$$

Also, let $T_{x}$ be the solution to the Lyapunov equation

$$
\begin{equation*}
A_{x} T_{x}+T_{x} A_{x}^{\prime}=E_{x} E_{x}^{\prime} \tag{5.1.33}
\end{equation*}
$$

which has the special form as in (5.1.11). Thus, (5.1.29) holds. Next, we define $\bar{S}_{x}=T_{x} / \gamma^{2}+X$. Clearly, we have $\bar{S}_{x}>T_{x} / \gamma^{2}$ and $\bar{S}_{x} \geq X>0$. Then, we have

$$
\begin{aligned}
& A_{x} \bar{S}_{x}+ \bar{S}_{x} A_{x}^{\prime}-B_{x} B_{x}^{\prime}+\bar{S}_{x} C_{x}^{\prime} C_{x} \bar{S}_{x}=A_{x}\left(T_{x} / \gamma^{2}+X\right) \\
& \quad+\left(T_{x} / \gamma^{2}+X\right) A_{x}^{\prime}-B_{x} B_{x}^{\prime}+\left(T_{x} / \gamma^{2}+X\right) C_{x}^{\prime} C_{x}\left(T_{x} / \gamma^{2}+X\right) \\
&=\left(A_{x} T_{x}+T_{x} A_{x}^{\prime}-E_{x} E_{x}^{\prime}\right) / \gamma^{2} \\
& \quad+A_{x} X+X A_{x}^{\prime}+E_{x} E_{x}^{\prime} / \gamma^{2}-B_{x} B_{x}^{\prime}+X C_{x}^{\prime} C_{x} X \\
&=0
\end{aligned}
$$

which implies that $\bar{S}_{x}>0$ is a solution of the Riccati equation (5.1.10). Since (5.1.10) can only have one positive definite solution, thus we have $\bar{S}_{x}=S_{x}$ and $S_{x}>T_{x} / \gamma^{2}$. This completes our proof of Lemma 5.1.2.

Now, let us get back to the proof of Theorem 5.1.1. Suppose that $\gamma>\gamma^{*}$. It is easy to verify that

$$
P(\gamma)=\left(\Gamma_{s}^{-1}\right)^{\prime}\left[\begin{array}{cc}
\left(S_{x}-T_{x} / \gamma^{2}\right)^{-1} & 0 \\
0 & 0
\end{array}\right] \Gamma_{s}^{-1}
$$

satisfies conditions 2.(a)-2.(c) of Theorem 3.1.1. Hence, there exists a state feedback law $u=F x$ with $F \in \mathbb{R}^{m \times n}$ (and obviously there exists a full information feedback law $u=F_{1} x+F_{2} w$ ) such that the $H_{\infty}$-norm of the resulting
closed-loop system from the disturbance $w$ to the controlled output $h, T_{h w}(s)$, is less than $\gamma$ and $\lambda(A+B F) \subset \mathbf{C}^{-}$.

The converse part of the theorem follows immediately from Lemmas 5.1.1 and 5.1.2 since the condition $\gamma>\left\{\lambda_{\max }\left(T_{x} S_{x}^{-1}\right)\right\}^{\frac{1}{2}}$ is equivalent to $S_{x}>T_{x} / \gamma^{2}$. This completes our proof of Theorem 5.1.1.

The following remarks are in order.
Remark 5.1.2. For the continuous-time systems, the infimum for the full information system of (5.1.1) with $D_{22}=0$ is equivalent to the infimum for the full state feedback system, i.e.,

$$
\left\{\begin{array}{l}
\dot{x}=A x+B u+E w  \tag{5.1.34}\\
y=x \\
h=C_{2} x+D_{2} u+D_{22} w
\end{array}\right.
$$

Thus, the infimum for the above full state feedback system is also given by $\gamma^{*}$ in (5.1.13).

Remark 5.1.3. If Assumption 5.F.3, i.e., the geometric condition, is not satisfied, then an iterative scheme might be used to determine the infimum. This can be done by finding the smallest scalar, say $\tilde{\gamma}^{*}$, such that the Riccati equation

$$
\begin{equation*}
\tilde{P}_{x} A_{x}+A_{x}^{\prime} \tilde{P}_{x}+\tilde{P}_{x} E_{x} E_{x}^{\prime} \tilde{P}_{x} /\left(\tilde{\gamma}^{*}\right)^{2}-\tilde{P}_{x} B_{x} B_{x}^{\prime} \tilde{P}_{x}+C_{x}^{\prime} C_{x}=0 \tag{5.1.35}
\end{equation*}
$$

has a positive definite solution $\tilde{P}_{x}>0$. One could also apply the result of Scherer [94] directly to the Riccati equation (5.1.20) to develop an iterative algorithm of the Newton type to compute an approximation of $\gamma^{*}$. The algorithm of Scherer has a quadratic convergent rate.

Remark 5.1.4. If $\Sigma_{P}$ is right invertible, then Assumption 5.F. 3 is automatically satisfied. Moreover, Assumption 5.F. 4 is no longer necessary and the infimum $\gamma^{*}$ for the full information feedback system (5.1.1) can be obtained as follows:

$$
\gamma^{*}=\left(\lambda_{\max }\left\{\left[\begin{array}{cc}
D_{22,1}^{\prime} D_{22,1} & 0  \tag{5.1.36}\\
0 & \tilde{T}_{x} \tilde{S}_{x}^{-1}
\end{array}\right]\right\}\right)^{\frac{1}{2}}
$$

where $\tilde{T}_{x}$ and $\tilde{S}_{x}$ are the positive semi-definite and positive definite solutions of the following Lyapunov equations,

$$
\begin{align*}
A_{a a}^{+} \tilde{T}_{x}+\tilde{T}_{x}\left(A_{a a}^{+}\right)^{\prime} & =\left(E_{a}^{+}-B_{0 a}^{+} D_{22,0}-L_{a d}^{+} \Gamma_{o r}^{-1} D_{22,1}\right) \\
& \times\left(E_{a}^{+}-B_{0 a}^{+} D_{22,0}-L_{a d}^{+} \Gamma_{o r}^{-1} D_{22,1}\right)^{\prime}  \tag{5.1.37}\\
A_{a a}^{+} \tilde{S}_{x}+\tilde{S}_{x}\left(A_{a a}^{+}\right)^{\prime} & =B_{0 a}^{+}\left(B_{0 a}^{+}\right)^{\prime}+L_{a d}^{+} \Gamma_{o r}^{-1}\left(L_{a d}^{+} \Gamma_{o r}^{-1}\right)^{\prime} \tag{5.1.38}
\end{align*}
$$

respectively, and $D_{22,0}$ and $D_{22,1}$ are as defined in (5.1.15) but for nonzero $D_{22}$. On the other hand, the infimum for the full state feedback system (5.1.34) is different from (5.1.36) and is given by

$$
\gamma^{*}=\left(\lambda_{\max }\left\{\left[\begin{array}{cc}
D_{22}^{\prime} D_{22} & 0  \tag{5.1.39}\\
0 & \tilde{T}_{x} \tilde{S}_{x}^{-1}
\end{array}\right]\right\}\right)^{\frac{1}{2}}
$$

where $\tilde{T}_{x}$ and $\tilde{S}_{x}$ are again the positive semi-definite and positive definite solutions of the Lyapunov equations (5.1.37) and (5.1.38), respectively. These claims can be verified using similar arguments as in the proof of Theorem 5.1.1. The detailed proofs can be found in Chen [14].

We conclude this section with the following illustrative examples.
Example 5.1.1. Consider a full information system (5.1.1) and a full state feedback system (5.1.34) characterized by

$$
A=\left[\begin{array}{lllll}
1 & 1 & 1 & 0 & 1  \tag{5.1.40}\\
0 & 1 & 0 & 0 & 1 \\
0 & 1 & 1 & 0 & 1 \\
1 & 1 & 1 & 1 & 1 \\
1 & 1 & 1 & 1 & 0
\end{array}\right], \quad B=\left[\begin{array}{lll}
0 & 0 & 0 \\
0 & 0 & 0 \\
1 & 0 & 0 \\
0 & 0 & 1 \\
0 & 1 & 0
\end{array}\right], \quad E=\left[\begin{array}{ll}
5 & 1 \\
0 & 0 \\
0 & 0 \\
2 & 3 \\
1 & 4
\end{array}\right],
$$

and

$$
C_{2}=\left[\begin{array}{lllll}
0 & 0 & 1 & 0 & 0  \tag{5.1.41}\\
0 & 0 & 0 & 0 & 1 \\
0 & 1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0
\end{array}\right], \quad D_{2}=\left[\begin{array}{lll}
1 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right], \quad D_{22}=0
$$

It is simple to verify that the subsystem $\left(A, B, C_{2}, D_{2}\right)$ is neither left- nor rightinvertible with one unstable invariant zero at $s=1$. Moreover, it is already in the form of special coordinate basis with

$$
\begin{gathered}
\Gamma_{s}=I_{5}, \quad \Gamma_{o r}=I_{3}, \quad n_{x}=3 \\
A_{x}=\left[\begin{array}{lll}
1 & 1 & 1 \\
0 & 1 & 0 \\
0 & 1 & 0
\end{array}\right], \quad B_{x} B_{x}^{\prime}=\left[\begin{array}{lll}
1 & 1 & 1 \\
1 & 1 & 1 \\
1 & 1 & 2
\end{array}\right], \quad C_{x}^{\prime} C_{x}=\left[\begin{array}{lll}
0 & 0 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{array}\right],
\end{gathered}
$$

and

$$
A_{a a}^{+}=1, \quad E_{a}^{+}=\left[\begin{array}{ll}
5 & 1
\end{array}\right]
$$

Then solving equations (5.1.10) and (5.1.12), we obtain

$$
S_{x}=\left[\begin{array}{rrr}
0.556281 & 0.185427 & -0.305593 \\
0.185427 & 0.395142 & 0.231469 \\
-0.305593 & 0.231469 & 1.217984
\end{array}\right], \quad T_{x}=\left[\begin{array}{rrr}
13 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right]
$$

and for both systems (5.1.1) and (5.1.34), the infima are given by

$$
\begin{equation*}
\gamma^{*}=\sqrt{\lambda_{\max }\left(T_{x} S_{x}^{-1}\right)}=6.4679044 \tag{E}
\end{equation*}
$$

Example 5.1.2. Consider a full information system (5.1.1) and a full state feedback system (5.1.34) characterized by

$$
A=\left[\begin{array}{llll}
3 & 0 & 0 & 1  \tag{5.1.42}\\
1 & 1 & 0 & 1 \\
1 & 1 & 0 & 0 \\
0 & 0 & 1 & 0
\end{array}\right], \quad B=\left[\begin{array}{lll}
1 & 0 & 0 \\
0 & 0 & 1 \\
0 & 1 & 0 \\
0 & 0 & 0
\end{array}\right], \quad E=\left[\begin{array}{l}
4 \\
3 \\
2 \\
1
\end{array}\right],
$$

and

$$
C_{2}=\left[\begin{array}{llll}
1 & 0 & 0 & 0  \tag{5.1.43}\\
0 & 0 & 0 & 1
\end{array}\right], \quad D_{2}=\left[\begin{array}{lll}
1 & 0 & 0 \\
0 & 0 & 0
\end{array}\right], \quad D_{22}=\left[\begin{array}{l}
2 \\
1
\end{array}\right] .
$$

It is simple to verify that the subsystem $\left(A, B, C_{2}, D_{2}\right)$ or $\Sigma_{\mathbf{P}}$ is controllable and right invertible with one unstable invariant zero at 2 and one infinite zero of order 2. Following Remark 5.1.4, we obtain

$$
\begin{gathered}
\Gamma_{s}=I_{4}, \quad \Gamma_{o r}=1, \quad n_{x}=1, \quad A_{a a}^{+}=2, \quad B_{0 a}^{+}=1 \\
L_{a d}^{+}=1, \quad E_{a}^{+}=4, \quad D_{22,0}=2, \quad D_{22,1}=1
\end{gathered}
$$

and

$$
\tilde{S}_{x}=0.5, \quad \tilde{T}_{x}=0.25
$$

Then, the infimum for the full information feedback system is given

$$
\gamma^{*}=\left(\lambda_{\max }\left\{\left[\begin{array}{cc}
D_{22,1}^{\prime} D_{22,1} & 0 \\
0 & \tilde{T}_{x} \tilde{S}_{x}^{-1}
\end{array}\right]\right\}\right)^{\frac{1}{2}}=\left(\lambda_{\max }\left\{\left[\begin{array}{cc}
1 & 0 \\
0 & 0.5
\end{array}\right]\right\}\right)^{\frac{1}{2}}=1
$$

and the infimum for the full state feedback system is

$$
\gamma^{*}=\left(\lambda_{\max }\left\{\left[\begin{array}{cc}
D_{22}^{\prime} D_{22} & 0 \\
0 & \tilde{T}_{x} \tilde{S}_{x}^{-1}
\end{array}\right]\right\}\right)^{\frac{1}{2}}=\left(\lambda_{\max }\left\{\left[\begin{array}{cc}
5 & 0 \\
0 & 0.5
\end{array}\right]\right\}\right)^{\frac{1}{2}}=\sqrt{5}
$$

Clearly, they are different.

### 5.2. Output Feedback Case

We present in this section an elegant well-conditioned non-iterative algorithm for the exact computation of $\gamma^{*}$ of the following measurement feedback system,

$$
\left\{\begin{array}{l}
\dot{x}=A x+B u+E w  \tag{5.2.1}\\
y=C_{1} x+D_{1} w \\
h=C_{2} x+D_{2} u+D_{22} w
\end{array}\right.
$$

where $x \in \mathbb{R}^{n}$ is the state, $u \in \mathbb{R}^{m}$ is the control input, $w \in \mathbb{R}^{q}$ is the external disturbance input, $y \in \mathbb{R}^{p}$ is the measurement output, and $h \in \mathbf{R}^{\ell}$ is the controlled output of $\Sigma$. Again, for the purpose of easy reference, we define $\Sigma_{\mathrm{P}}$ to be the subsystem characterized by the matrix quadruple $\left(A, B, C_{2}, D_{2}\right)$ and $\Sigma_{\mathrm{Q}}$ to be the subsystem characterized by the matrix quadruple $\left(A, E, C_{1}, D_{1}\right)$. We first make the following assumptions:

Assumption 5.M.1: $(A, B)$ is stabilizable;
Assumption 5.M.2: $\Sigma_{\mathrm{P}}$ has no invariant zero on the imaginary axis;
Assumption 5.M.3: $\operatorname{Im}(E) \subset \mathcal{V}^{-}\left(\Sigma_{\mathrm{P}}\right)+\mathcal{S}^{-}\left(\Sigma_{\mathrm{P}}\right)$;
Assumption 5.M.4: $\left(A, C_{1}\right)$ is detectable;
Assumption 5.M.5: $\Sigma_{\mathbf{Q}}$ has no invariant zero on the imaginary axis;
Assumption 5.M.6: $\operatorname{Ker}\left(C_{2}\right) \supset \mathcal{V}^{-}\left(\Sigma_{\mathbf{Q}}\right) \cap \mathcal{S}^{-}\left(\Sigma_{\mathbf{Q}}\right)$; and
Assumption 5.M.7: $D_{22}=0$.
Remark 5.2.1. Here we note that Assumptions 5.M.1 and 5.M.4, i.e., $(A, B)$ is stabilizable and $\left(A, C_{1}\right)$ is detectable, are necessary for the existence of any stabilizing controller. Assumptions 5.M.2 and 5.M. 5 will be removed later in Section 5.3. Also, Assumptions 5.M. 3 and 5.M. 6 will be automatically satisfied if $\Sigma_{P}$ is right invertible and if $\Sigma_{Q}$ is left invertible. Moreover, in this case, $D_{22}=0$, i.e., Assumption 5.M.7, can be removed without any difficulties (see Remark 5.2.3 later in this section).

We have the following non-iterative algorithm for computing the infimum, $\gamma^{*}$, of the general measurement feedback system (5.2.1).

Step 5.M.1. Define an auxiliary full information system

$$
\left\{\begin{array}{l}
\dot{x}=A x+B u+E \quad w  \tag{5.2.2}\\
y=\binom{0}{I} x+\binom{I}{0} w \\
h=C_{2} x+D_{2} u+D_{22} w
\end{array}\right.
$$

and perform Steps 5.F. 1 and 5.F. 2 of the algorithm as given in Section 5.1. For easy reference in future development, we append a subscript ' P ' to all sub-matrices and transformations in the special coordinate basis associated with the system (5.2.2). In particular, we rename the state transformation of the special coordinate basis for $\Sigma_{\mathrm{P}}$ as $\Gamma_{s \mathrm{P}}$, and the dimension of $\mathbb{R}^{n} / \mathcal{S}^{+}\left(\Sigma_{\mathrm{P}}\right)$ as $n_{x \mathrm{P}}$. Furthermore, $S_{x}$ of (5.1.10) and $T_{x}$ of (5.1.11) are respectively renamed to $S_{x \mathrm{P}}$ and $T_{x \mathrm{p}}$.

Step 5.M.2. Define another auxiliary full information system

$$
\left\{\begin{array}{l}
\dot{x}=A^{\prime} x+C_{1}^{\prime} u+C_{2}^{\prime} w  \tag{5.2.3}\\
y=\binom{0}{I} x+\binom{I}{0} w \\
z=E^{\prime} x+D_{1}^{\prime} u+D_{22}^{\prime} w
\end{array}\right.
$$

and again perform Steps 5.F. 1 and 5.F. 2 of the algorithm as given in Section 5.1 one more time but for this auxiliary system. To all sub-matrices and transformations in the special coordinate basis of $\Sigma_{4}^{\star}$, where $\Sigma_{\mathrm{y}}^{\star}$ is the dual system of $\Sigma_{Q}$ and is characterized by quadruple ( $A^{\prime}, C_{1}^{\prime}, E^{\prime}, D_{1}^{\prime}$ ), we append a subscript ' $Q$ ' to signify their relation to the system $\Sigma_{6}^{\star}$. In particular, we rename the state transformation of the special coordinate basis for this case as $\Gamma_{s Q}$, and the dimension of $\mathbb{R}^{n} / \mathcal{S}^{+}\left(\Sigma_{\mathrm{q}}^{\star}\right)$ as $n_{x Q}$. As in Step 5.M.1, we also rename $S_{x}$ of (5.1.10) and $T_{x}$ of (5.1.11) as $S_{x Q}$ and $T_{x Q}$, respectively.

Step 5.M.3. Partition

$$
\Gamma_{s \mathrm{P}}^{-1}\left(\Gamma_{s \mathrm{Q}}^{-1}\right)^{\prime}=\left[\begin{array}{ll}
\Gamma & \star  \tag{5.2.4}\\
\star & \star
\end{array}\right],
$$

where $\Gamma$ is a $n_{x \mathrm{P}} \times n_{x \mathrm{Q}}$ matrix, and define a constant matrix

$$
M=\left[\begin{array}{cc}
T_{x \mathrm{P}} S_{x \mathrm{P}}^{-1}+\Gamma S_{x \mathrm{Q}}^{-1} \Gamma^{\prime} S_{x \mathrm{P}}^{-1} & -\Gamma S_{x \mathrm{Q}}^{-1}  \tag{5.2.5}\\
-T_{x \mathrm{Q}} S_{x \mathrm{Q}}^{-1} \Gamma^{\prime} S_{x \mathrm{P}}^{-1} & T_{x \mathrm{Q}} S_{x \mathrm{Q}}^{-1}
\end{array}\right] .
$$

Step 5.M.4. The infimum $\gamma^{*}$ for the measurement feedback system (5.2.1) is then given by

$$
\begin{equation*}
\gamma^{*}=\sqrt{\lambda_{\max }(M)} \tag{5.2.6}
\end{equation*}
$$

It will be shown later in Proposition 5.2.4 that the matrix $M$ of (5.2.5) has only real and nonnegative eigenvalues.

The proof of the above algorithm is rather involved. We would have to introduce several lemmas before proceeding to its final proof. Let us first define

$$
\begin{gather*}
\gamma_{\mathrm{P}}^{*}:=\left\{\lambda_{\max }\left(T_{x \mathrm{P}} S_{x \mathrm{P}}^{-1}\right)\right\}^{\frac{1}{2}} \quad \text { and } \quad \gamma_{\mathrm{Q}}^{*}:=\left\{\lambda_{\max }\left(T_{x \mathrm{Q}} S_{x \mathrm{Q}}^{-1}\right)\right\}^{\frac{1}{2}},  \tag{5.2.7}\\
P(\gamma):=\left(\Gamma_{s \mathrm{P}}^{-1}\right)^{\prime}\left[\begin{array}{cc}
\left(S_{x \mathrm{P}}-T_{x \mathrm{P}} / \gamma^{2}\right)^{-1} & 0 \\
0 & 0
\end{array}\right] \Gamma_{s \mathrm{P}}^{-1}, \tag{5.2.8}
\end{gather*}
$$

and

$$
Q(\gamma):=\left(\Gamma_{s Q}^{-1}\right)^{\prime}\left[\begin{array}{cc}
\left(S_{x Q}-T_{x Q} / \gamma^{2}\right)^{-1} & 0  \tag{5.2.9}\\
0 & 0
\end{array}\right] \Gamma_{s Q}^{-1}
$$

We have the following lemma.

Lemma 5.2.1. Consider the system (5.2.1), which satisfies Assumptions 5.M.1 to 5.M.7. Then we have

1. For $\gamma>\gamma_{\mathrm{p}}^{*}$, the positive semi-definite matrix $P(\gamma)$ given by (5.2.8) is the unique solution to the matrix inequality $F_{\gamma}(P) \geq 0$, i.e., condition 2.(a) of Theorem 3.1.1, and satisfies both rank conditions 2.(b) and 2.(c) of Theorem 3.1.1. Moreover, such a solution $P(\gamma)$ does not exist when $\gamma<\gamma_{\mathrm{P}}^{*}$.
2. For $\gamma>\gamma_{Q}^{*}$, the positive semi-definite matrix $Q(\gamma)$ given by (5.2.9) is the unique solution to the matrix inequality $G_{\gamma}(Q) \geq 0$, i.e., condition 2.(d) of Theorem 3.1.1, and satisfies both rank conditions 2.(e) and 2.(f) of Theorem 3.1.1. Moreover, such a solution $Q(\gamma)$ does not exist when $\gamma<\gamma_{\mathrm{Q}}^{*}$.

Proof. It follows from Theorem 5.1.1.

The next lemma gives an equivalence of the infimum, $\gamma^{*}$, for the measurement feedback system (5.2.1).

Lemma 5.2.2. Let $\gamma_{\mathrm{PQ}}^{*}:=\max \left\{\gamma_{\mathrm{P}}^{*}, \gamma_{\mathrm{Q}}^{*}\right\}$. Then the infimum for the given measurement feedback system (5.2.1) is equivalent to

$$
\begin{equation*}
\gamma^{*}=\inf \left\{\gamma \in\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right) \mid f(\gamma)<\gamma^{2}\right\} \tag{5.2.10}
\end{equation*}
$$

where the scalar function

$$
\begin{equation*}
f(\gamma):=\rho\{P(\gamma) Q(\gamma)\} \tag{5.2.11}
\end{equation*}
$$

and $P(\gamma)$ and $Q(\gamma)$ are given by (5.2.8) and (5.2.9) respectively.
Proof. It follows Lemma 5.2.2 that $\gamma^{*} \geq \gamma_{\mathrm{PQ}}^{*}$. Next, for any $\hat{\gamma} \in\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$ such that $f(\hat{\gamma})<\hat{\gamma}^{2}$, i.e. $\rho\{P(\hat{\gamma}) Q(\hat{\gamma})\}<\hat{\gamma}^{2}$, then the corresponding $P(\hat{\gamma})$ and $Q(\hat{\gamma})$ as given in (5.2.8) and (5.2.9) satisfy the conditions of Theorem 3.1.1. Hence, $\hat{\gamma}>\gamma^{*}$ and $\gamma^{*}$ is equivalent to that of (5.2.10).

It is then straightforward to show that the scalar function $f(\gamma)$ of (5.2.11) is given by

$$
\begin{equation*}
f(\gamma)=\lambda_{\max }\left\{\left(S_{x \mathrm{P}}-\gamma^{-2} T_{x \mathrm{P}}\right)^{-1} \Gamma\left(S_{x \mathrm{Q}}-\gamma^{-2} T_{x \mathrm{Q}}\right)^{-1} \Gamma^{\prime}\right\} \tag{5.2.12}
\end{equation*}
$$

The function $f(\gamma)$ of (5.2.12) is a well-defined mapping from $\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$ to $[0, \infty)$. Its evaluation involves the computation of the maximum eigenvalue of a matrix
of dimension $n_{x \mathrm{P}} \times n_{x \mathrm{P}}$, which is normally of a much smaller dimension than the original product $P(\gamma) Q(\gamma)$. We establish some important properties of the function $f(\gamma)$ in the following propositions.

Proposition 5.2.1. $f(\gamma)$ is a continuous, nonnegative and non-increasing function of $\gamma$ on $\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$.

Proof. We first show that $P_{x}(\gamma):=\left(S_{x \mathrm{P}}-\gamma^{-2} T_{x \mathrm{P}}\right)^{-1}$ is non-increasing, i.e., if $\gamma_{2}>\gamma_{1}$ then $P_{x}\left(\gamma_{2}\right) \leq P_{x}\left(\gamma_{1}\right)$. Recall that $S_{x \mathrm{P}}>0$ and $T_{x \mathrm{P}} \geq 0$, we have for all $\gamma_{2}>\gamma_{1}>\gamma_{\mathrm{PQ}}^{*}$

$$
\left(\gamma_{1}^{-2}-\gamma_{2}^{-2}\right) T_{x \mathrm{P}} \geq 0
$$

which implies that

$$
S_{x \mathrm{P}}-\gamma_{1}^{-2} T_{x \mathrm{P}} \leq S_{x \mathrm{P}}-\gamma_{2}^{-2} T_{x \mathrm{p}}
$$

Hence,

$$
P_{x}\left(\gamma_{2}\right) \leq P_{x}\left(\gamma_{1}\right), \quad \text { for } \gamma_{2}>\gamma_{1}
$$

Similarly, one can show that $Q_{x}(\gamma):=\left(S_{x Q}-\gamma^{-2} T_{x Q}\right)^{-1}$ is non-increasing. This implies that $\Gamma Q_{x}(\gamma) \Gamma^{\prime}$ is also non-increasing. Then clearly $f(\gamma)$ is a continuous, non-negative and non-increasing function of $\gamma$ on $\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$.

The function $f(\gamma)$ defined above can be extended as a mapping from $\left[\gamma_{\mathrm{PQ}}^{*}, \infty\right)$ to $[0, \infty)$ by setting

$$
\begin{equation*}
f\left(\gamma_{\mathrm{PQ}}^{*}\right)=\lim _{\gamma \rightarrow \gamma_{\mathrm{PQ}}^{*}} f(\gamma) \tag{5.2.13}
\end{equation*}
$$

It follows from Proposition 5.2.1 that the limit $f\left(\gamma_{\mathrm{PQ}}^{*}\right)$ exists and could be finite or infinite.

Proposition 5.2.2. $f(\gamma)=\gamma^{2}$ has either no solution or a unique solution in the interval $\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$.

Proof. The result follows from Proposition 5.2.1 and the fact that $\gamma^{2}$ is strictly increasing for positive $\gamma$.

Proposition 5.2.3. If $f(\gamma)=\gamma^{2}$ has no solution in the interval $\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$ then $\gamma^{*}$ is equal to $\gamma_{\mathrm{PQ}}^{*}$. Otherwise, $\gamma^{*}$ is equal to the unique solution of $f(\gamma)=\gamma^{2}$ in the interval $\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$.

Proof. If $f(\gamma)=\gamma^{2}$ has no solution in the interval $\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$, then $f(\gamma)<\gamma^{2}$ for all $\gamma \in\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$ and hence according to Lemma 5.2.2, $\gamma^{*}=\gamma_{\mathrm{PQ}}^{*}$. On the other hand, it is obvious that $\gamma^{*}$ is equal to the unique solution of $f(\gamma)=\gamma^{2}$ when such a solution exists.

At first glance, it seems that the solution of $f(\gamma)=\gamma^{2}$ would involve the rooting of a highly nonlinear algebraic equation in $\gamma$. Actually its solution can be achieved in one step. Namely the problem of solving $f(\gamma)=\gamma^{2}$, if such a solution exists in the interval $\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$, can be converted to the problem of calculating the maximum eigenvalue of a constant matrix, i.e., $M$ of (5.2.5). In fact, we would also show that, when $f(\gamma)=\gamma^{2}$ has no solution in the interval $\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$, the maximum eigenvalue of this matrix $M$ is equal to $\gamma_{\mathrm{PQ}}^{*}$, which is $\gamma^{*}$ as well. To prove this, we would have to introduce a matrix function of $\gamma$,

$$
\begin{equation*}
N(\gamma):=\left(S_{x \mathrm{P}}-\gamma^{-2} T_{x \mathrm{P}}\right)^{-1} \Gamma\left(S_{x \mathrm{Q}}-\gamma^{-2} T_{x \mathrm{Q}}\right)^{-1} \Gamma^{\prime}-\gamma^{2} I . \tag{5.2.14}
\end{equation*}
$$

We have the following propositions on the properties of the matrices $M$ and $N(\gamma)$.

Proposition 5.2.4. The eigenvalues of the matrix $M$ of (5.2.5) are real and non-negative.

Proof. First, we have

$$
\begin{align*}
\lambda\{M\} & =\lambda\left\{\left[\begin{array}{cc}
I & 0 \\
0 & T_{x \mathrm{Q}}
\end{array}\right]\left[\begin{array}{cc}
T_{x \mathrm{P}}+\Gamma S_{x Q}^{-1} \Gamma^{\prime} & -\Gamma S_{x Q}^{-1} \\
-S_{x Q}^{-1} \Gamma^{\prime} & S_{x Q}^{-1}
\end{array}\right]\left[\begin{array}{cc}
S_{x \mathrm{P}}^{-1} & 0 \\
0 & I
\end{array}\right]\right\} \\
& =\lambda\left\{\left[\begin{array}{cc}
S_{x \mathrm{P}}^{-1} & 0 \\
0 & I
\end{array}\right]\left[\begin{array}{cc}
I & 0 \\
0 & T_{x Q}
\end{array}\right]\left[\begin{array}{cc}
T_{x \mathrm{P}}+\Gamma S_{x Q}^{-1} \Gamma^{\prime} & -\Gamma S_{x Q}^{-1} \\
-S_{x Q}^{-1} \Gamma^{\prime} & S_{x Q}^{-1}
\end{array}\right]\right\} \\
& =\lambda\left\{\left[\begin{array}{cc}
S_{x \mathrm{P}}^{-1} & 0 \\
0 & T_{x \mathrm{Q}}
\end{array}\right]\left[\begin{array}{cc}
T_{x \mathrm{P}}+\Gamma S_{x Q}^{-1} \Gamma^{\prime} & -\Gamma S_{x \mathrm{Q}}^{-1} \\
-S_{x Q}^{-1} \Gamma^{\prime} & S_{x Q}^{-1}
\end{array}\right]\right\} . \tag{5.2.15}
\end{align*}
$$

Now, it is trivial to verify that both sub-matrices in (5.2.15) are symmetric and positive semi-definite. Then, using the result of Wielandt [110] (i.e., Theorem 3 ), it is simple to show that the eigenvalues of $M$ are real and nonnegative. 因

## Proposition 5.2.5.

1. $N(\gamma)$ has real eigenvalues for all $\gamma \in\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$.
2. $\lambda_{\max }\{N(\gamma)\}=f(\gamma)-\gamma^{2}$ is a continuous and strictly decreasing function of $\gamma$ in $\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$.

Proof. Note that both $\left(S_{x \mathrm{P}}-\gamma^{-2} T_{x \mathrm{P}}\right)^{-1}$ and $\left(S_{x \mathrm{Q}}-\gamma^{-2} T_{x \mathrm{Q}}\right)^{-1}$ are symmetric and positive definite for all $\gamma \in\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$. Hence, all the eigenvalues of $N(\gamma)$ are real for $\gamma \in\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$. The second item follows from Proposition 5.2.1. 因

Proposition 5.2.6. The roots of $\operatorname{det}[N(\gamma)]=0$ are real. Moreover, the largest root of $\operatorname{det}[N(\gamma)]=0$ in the interval $\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$ is equal to $\left\{\lambda_{\max }(M)\right\}^{\frac{1}{2}}$.

Proof. Using the definition of $N(\gamma)$ in (5.2.14), we have

$$
\begin{align*}
& \operatorname{det}[N(\gamma)]=(-1)^{n_{x \mathrm{P}}} \cdot \operatorname{det}\left[\gamma^{2} I-\left(S_{x \mathrm{P}}-\gamma^{-2} T_{x \mathrm{P}}\right)^{-1} \Gamma\left(S_{x \mathrm{Q}}-\gamma^{-2} T_{x \mathrm{Q}}\right)^{-1} \Gamma^{\prime}\right] \\
& \quad=\frac{(-1)^{n_{x \mathrm{P}}}}{\operatorname{det}\left[S_{x \mathrm{P}}-\gamma^{-2} T_{x \mathrm{P}}\right]} \cdot \operatorname{det}\left[\gamma^{2} S_{x \mathrm{P}}-T_{x \mathrm{P}}-\gamma^{2} \Gamma\left(\gamma^{2} S_{x \mathrm{Q}}-T_{x \mathrm{Q}}\right)^{-1} \Gamma^{\prime}\right] \\
& \quad=\frac{(-1)^{n_{x} \mathrm{P}}}{\operatorname{det}\left[S_{x \mathrm{P}}-\gamma^{-2} T_{x \mathrm{P}}\right] \cdot \operatorname{det}\left[\gamma^{2} S_{x \mathrm{Q}}-T_{x \mathrm{Q}}\right]} \cdot \operatorname{det}\left[\begin{array}{c}
\gamma^{2} S_{x \mathrm{P}}-T_{x \mathrm{P}} \\
\gamma^{2} \Gamma^{\prime} \\
\gamma^{2} S_{x \mathrm{Q}}-T_{x \mathrm{Q}}
\end{array}\right] \\
& \quad=\frac{(-1)^{n_{x \mathrm{P}} \cdot \operatorname{det}\left[S_{x \mathrm{P}}\right] \cdot \operatorname{det}\left[S_{x \mathrm{Q}}\right]}}{\operatorname{det}\left[S_{x \mathrm{P}}-\gamma^{-2} T_{x \mathrm{P}}\right] \cdot \operatorname{det}\left[\gamma^{2} S_{x \mathrm{Q}}-T_{x \mathrm{Q}}\right]} \cdot \operatorname{det}\left[\gamma^{2} I-M\right] . \tag{5.2.16}
\end{align*}
$$

Now it is simple to see that the roots of $\operatorname{det}[N(\gamma)]=0$ are real since all the roots of $\operatorname{det}\left[\gamma^{2} S_{x \mathrm{P}}-T_{x \mathrm{P}}\right]=0, \operatorname{det}\left[\gamma^{2} S_{x \mathrm{Q}}-T_{x \mathrm{Q}}\right]=0$ and $\operatorname{det}\left[\gamma^{2} I-M\right]=0$ are real. Clearly, $\operatorname{det}\left[S_{x \mathrm{P}}-\gamma^{-2} T_{x \mathrm{P}}\right] \neq 0$ and $\operatorname{det}\left[\gamma^{2} S_{x \mathrm{Q}}-T_{x \mathrm{Q}}\right] \neq 0$ for all $\gamma \in\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$. Hence the largest root of $\operatorname{det}[N(\gamma)]=0$ in $\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$ is equal to the largest root of $\operatorname{det}\left[\gamma^{2} I-M\right]=0$, which is equal to $\left\{\lambda_{\max }(M)\right\}^{\frac{1}{2}}$.

Finally, we are ready to prove our algorithm for computing the infimum $\gamma^{*}$ for measurement feedback systems. We have the following theorem.

Theorem 5.2.1. Consider the measurement feedback system (5.2.1), which satisfies Assumptions 5.M.1 to 5.M.7. Then

$$
\begin{equation*}
\gamma^{*}=\sqrt{\lambda_{\max }(M)} \tag{5.2.17}
\end{equation*}
$$

where $M$ as defined in (5.2.5), is indeed its infimum.
Proof. First, we will show that $\gamma^{*}$ is equal to the largest root of $\operatorname{det}[N(\gamma)]=0$ when $f(\gamma)=\gamma^{2}$ has a unique solution in $\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$. It is simple to observe that $\operatorname{det}\left[N\left(\gamma^{*}\right)\right]=0$ since $\lambda_{\max }\left[N\left(\gamma^{*}\right)\right]=f\left(\gamma^{*}\right)-\left(\gamma^{*}\right)^{2}=0$. Now suppose that there exists a $\gamma_{1}$ such that $\operatorname{det}\left[N\left(\gamma_{1}\right)\right]=0$ and $\gamma_{1}>\gamma^{*}$. This implies that there exists an eigenvalue of $N\left(\gamma_{1}\right)$, say $\lambda_{i}\left[N\left(\gamma_{1}\right)\right]$, such that $\lambda_{i}\left[N\left(\gamma_{1}\right)\right] \neq \lambda_{\max }\left[N\left(\gamma_{1}\right)\right]$ and $\lambda_{i}\left[N\left(\gamma_{1}\right)\right]=0$. Thus, we have

$$
\begin{equation*}
\lambda_{\max }\left[N\left(\gamma_{1}\right)\right]>\lambda_{i}\left[N\left(\gamma_{1}\right)\right]=0=\lambda_{\max }\left[N\left(\gamma^{*}\right)\right] \tag{5.2.18}
\end{equation*}
$$

contradicting the findings in Proposition 5.2.5 that $\lambda_{\max }[N(\gamma)]$ must be a nonincreasing function. Hence, $\gamma^{*}$ is the largest root of $\operatorname{det}[N(\gamma)]=0$ and it is equal to $\left\{\lambda_{\max }(M)\right\}^{\frac{1}{2}}$ as shown in Proposition 5.2.6.

Now we consider the situation when $f(\gamma)=\gamma^{2}$ has no solution in the interval $\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$. In this case, clearly we have $\gamma^{*}=\gamma_{\mathrm{PQ}}^{*}$ and $0 \leq f\left(\gamma_{\mathrm{PQ}}^{*}\right) \leq\left(\gamma_{\mathrm{PQ}}^{*}\right)^{2}$. The last inequality and the definition of $N(\gamma)$ in (5.2.14) imply that

$$
\begin{equation*}
-\left(\gamma_{\mathrm{PQ}}^{*}\right)^{2} \leq \lambda_{i}\left[N\left(\gamma_{\mathrm{PQ}}^{*}\right)\right] \leq 0 \tag{5.2.19}
\end{equation*}
$$

Thus, the determinant of $N\left(\gamma_{\mathrm{PQ}}^{*}\right)$ is bounded. Evaluating equation (5.2.16) at $\gamma=\gamma_{\mathrm{PQ}}^{*}$, we have

$$
\begin{align*}
& \operatorname{det}\left[N\left(\gamma_{\mathrm{PQ}}^{*}\right)\right] \cdot \operatorname{det}\left[S_{x \mathrm{P}}-\left(\gamma_{\mathrm{PQ}}^{*}\right)^{-2} T_{x \mathrm{P}}\right] \cdot \operatorname{det}\left[\left(\gamma_{\mathrm{PQ}}^{*}\right)^{2} S_{x \mathrm{Q}}-T_{x \mathrm{Q}}\right] \\
&=(-1)^{n_{x \mathrm{P}}} \cdot \operatorname{det}\left[S_{x \mathrm{P}}\right] \cdot \operatorname{det}\left[S_{x \mathrm{Q}}\right] \cdot \operatorname{det}\left[\left(\gamma_{\mathrm{PQ}}^{*}\right)^{2} I-M\right] \tag{5.2.20}
\end{align*}
$$

Note that from (5.2.7) and the definition of $\gamma_{\mathrm{PQ}}^{*}$, we have

$$
\begin{equation*}
\operatorname{det}\left[S_{x \mathrm{P}}-\left(\gamma_{\mathrm{PQ}}^{*}\right)^{-2} T_{x \mathrm{P}}\right] \cdot \operatorname{det}\left[\left(\gamma_{\mathrm{PQ}}^{*}\right)^{2} S_{x \mathrm{Q}}-T_{x \mathrm{Q}}\right]=0, \tag{5.2.21}
\end{equation*}
$$

and since $\operatorname{det}\left[N\left(\gamma_{\mathrm{PQ}}^{*}\right)\right]$ is bounded, it follows from (5.2.20) that

$$
\begin{equation*}
\operatorname{det}\left[\left(\gamma_{\mathrm{PQ}}^{*}\right)^{2} I-M\right]=0 \tag{5.2.22}
\end{equation*}
$$

or $\left(\gamma_{\mathrm{PQ}}^{*}\right)^{2}$ is an eigenvalue of $M$. Furthermore since $\operatorname{det}[N(\gamma)]=0$ and similarly $\operatorname{det}\left[\gamma^{2} I-M\right]=0$ do not have a root in $\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right)$, hence $\gamma_{\mathrm{PQ}}^{*}=\left\{\lambda_{\max }(M)\right\}^{\frac{1}{2}}$. This completes the proof of Theorem 5.2.1.

The following remarks are in order.
Remark 5.2.2. If Assumptions 5.M. 3 and 5.M.6, i.e., the geometric conditions, are not satisfied, then an iterative scheme might be used to determine the infimum. This can be done by finding the smallest scalar, say $\tilde{\gamma}^{*}$, such that the Riccati equation

$$
\begin{equation*}
\tilde{P}_{x} A_{x \mathrm{P}}+A_{x \mathrm{P}}^{\prime} \tilde{P}_{x}+\tilde{P}_{x} E_{x \mathrm{P}} E_{x \mathrm{P}}^{\prime} \tilde{P}_{x} /\left(\tilde{\gamma}^{*}\right)^{2}-\tilde{P}_{x} B_{x \mathrm{P}} B_{x \mathrm{P}}^{\prime} \tilde{P}_{x}+C_{x \mathrm{P}}^{\prime} C_{x \mathrm{P}}=0 \tag{5.2.23}
\end{equation*}
$$

has a positive definite solution $\tilde{P}_{x}>0$, the Riccati equation

$$
\begin{equation*}
\tilde{Q}_{x} A_{x Q}+A_{x Q}^{\prime} \tilde{Q}_{x}+\tilde{Q}_{x} E_{x Q} E_{x Q}^{\prime} \tilde{Q}_{x} /\left(\tilde{\gamma}^{*}\right)^{2}-\tilde{Q}_{x} B_{x Q} B_{x Q}^{\prime} \tilde{Q}_{x}+C_{x Q}^{\prime} C_{x Q}=0 \tag{5.2.24}
\end{equation*}
$$

has a positive definite solution $\tilde{Q}_{x}>0$, and

$$
\begin{equation*}
\lambda_{\max }\left\{\tilde{P}_{x} \Gamma \tilde{Q}_{x} \Gamma^{\prime}\right\}<\left(\tilde{\gamma}^{*}\right)^{2} \tag{5.2.25}
\end{equation*}
$$

Here $\Gamma$ is as defined in (5.2.4). Also, all sub-matrices with subscript ' $P$ ' are related to the special coordinate basis decomposition of $\Sigma_{P}$ and the system (5.2.2), and all sub-matrices with subscript ' $Q$ ' are related to the special coordinate basis decomposition of $\Sigma_{\mathrm{Q}}^{\star}$ and the system (5.2.3).

回
Remark 5.2.3. If $\Sigma_{P}$ is right invertible and $\Sigma_{Q}$ is left invertible, then Assumptions 5.M.3 and 5.M.6, i.e., the geometric conditions, are automatically satisfied. Moreover, Assumption 5.M.7, $D_{22}=0$, is no longer necessary and
the infimum $\gamma^{*}$ for the measurement feedback system (5.2.1) can be obtained as follows:

$$
\gamma^{*}=\left(\lambda_{\max }\left\{\left[\begin{array}{cccc}
D_{22,1 \mathrm{P}}^{\prime} D_{22,1 \mathrm{P}} & 0 & 0 & 0 \\
0 & \tilde{T}_{x \mathrm{P}} \tilde{S}_{x \mathrm{P}}^{-1}+\Gamma \tilde{S}_{x \mathrm{Q}}^{-1} \Gamma^{\prime} \tilde{S}_{x \mathrm{P}}^{-1} & -\Gamma \tilde{S}_{x \mathrm{Q}}^{-1} & 0 \\
0 & -\tilde{T}_{x \mathrm{Q}} \tilde{S}_{x \mathrm{Q}}^{-1} \Gamma^{\prime} \tilde{S}_{x \mathrm{P}}^{-1} & \tilde{T}_{x \mathrm{Q}} \tilde{S}_{x \mathrm{Q}}^{-1} & 0 \\
0 & 0 & 0 & D_{22,1 \mathrm{Q}}^{\prime} D_{22,1 \mathrm{Q}}
\end{array}\right]\right\}\right)
$$

where $\Gamma$ is as defined in (5.2.4), $\tilde{T}_{x \mathrm{P}}$ and $\tilde{S}_{x \mathrm{P}}$ are the positive semi-definite and positive definite solutions of the following Lyapunov equations,

$$
\begin{align*}
A_{a a \mathrm{P}}^{+} \tilde{T}_{x \mathrm{P}}+\tilde{T}_{x \mathrm{P}}\left(A_{a a \mathrm{P}}^{+}\right)^{\prime} & =\left(E_{a \mathrm{P}}^{+}-B_{0 a \mathrm{P}}^{+} D_{22,0 \mathrm{P}}-L_{a d \mathrm{P}}^{+} \Gamma_{o r \mathrm{P}}^{-1} D_{22,1 \mathrm{P}}\right) \\
& \times\left(E_{a \mathrm{P}}^{+}-B_{0 a \mathrm{P}}^{+} D_{22,0 \mathrm{P}}-L_{a d \mathrm{P}}^{+} \Gamma_{o r \mathrm{P}}^{-1} D_{22,1 \mathrm{P}}\right)^{\prime},  \tag{5.2.26}\\
A_{a a \mathrm{P}}^{+} \tilde{S}_{x \mathrm{P}}+\tilde{S}_{x \mathrm{P}}\left(A_{a a \mathrm{P}}^{+}\right)^{\prime} & =B_{0 a \mathrm{P}}^{+}\left(B_{0 a \mathrm{P}}^{+}\right)^{\prime}+L_{a d \mathrm{P}}^{+} \Gamma_{o r \mathrm{P}}^{-1}\left(L_{a d \mathrm{P}}^{+} \Gamma_{o r \mathrm{P}}^{-1}\right)^{\prime} \tag{5.2.27}
\end{align*}
$$

and $\tilde{T}_{x \mathrm{Q}}$ and $\tilde{S}_{x \mathrm{Q}}$ are the positive semi-definite and positive definite solutions of the following Lyapunov equations,

$$
\begin{align*}
A_{a a Q}^{+} \tilde{T}_{x Q}+\tilde{T}_{x Q}\left(A_{a a Q}^{+}\right)^{\prime} & =\left(E_{a Q}^{+}-B_{0 a Q}^{+} D_{22,0 Q}-L_{a d Q}^{+} \Gamma_{o r Q}^{-1} D_{22,1 Q}\right) \\
& \times\left(E_{a Q}^{+}-B_{0 a Q}^{+} D_{22,0 Q}-L_{a d Q}^{+} \Gamma_{o r Q}^{-1} D_{22,1 Q}\right)^{\prime}  \tag{5.2.28}\\
A_{a a Q}^{+} \tilde{S}_{x Q}+\tilde{S}_{x Q}\left(A_{a a Q}^{+}\right)^{\prime} & =B_{0 a Q}^{+}\left(B_{0 a Q}^{+}\right)^{\prime}+L_{a d Q}^{+} \Gamma_{o r Q}^{-1}\left(L_{a d Q}^{+} \Gamma_{o r Q}^{-1}\right)^{\prime} \tag{5.2.29}
\end{align*}
$$

Here again all sub-matrices with subscript ' P ' are related to the special coordinate basis decomposition of $\Sigma_{\mathrm{P}}$ and the system (5.2.2), while all sub-matrices with subscript ' $Q$ ' are related to the special coordinate basis decomposition of $\Sigma_{\mathrm{Q}}^{\star}$ and the system (5.2.3). The detailed proof of the above claim is similar to that of Theorem 5.2.1. It can be found in Chen [14].

We illustrate our results in the following examples.

Example 5.2.1. We consider a measurement feedback system (5.2.1) with $A$, $B, E, C_{2}, D_{2}, D_{22}$ being given as in Example 5.1.1 of Section 5.1 and

$$
C_{1}=\left[\begin{array}{rrrrr}
0 & -2 & -3 & -2 & -1  \tag{5.2.30}\\
1 & 2 & 3 & 2 & 1
\end{array}\right], \quad D_{1}=\left[\begin{array}{ll}
1 & 0 \\
0 & 0
\end{array}\right]
$$

Step 5.M.1. It was computed in Example 5.1.1 that $\Gamma_{s \mathrm{P}}=I_{5}, n_{x \mathrm{P}}=3$ and

$$
S_{x \mathrm{P}}=\left[\begin{array}{rrr}
0.556281 & 0.185427 & -0.305593 \\
0.185427 & 0.395142 & 0.231469 \\
-0.305593 & 0.231469 & 1.217984
\end{array}\right], \quad T_{x \mathrm{P}}=\left[\begin{array}{ccc}
13 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right] .
$$

Step 5.M.2. The subsystem $\left(A, E, C_{1}, D_{1}\right)$ is invertible and of nonminimum phase with invariant zeros at $\{-1.630662,-3.593415,0.521129 \pm j 0.363042\}$. Following our algorithm, we obtain

$$
\begin{gathered}
\Gamma_{s Q}=\left[\begin{array}{rrrrr}
-0.011218 & -0.106028 & -0.906482 & -0.212184 & 0.090909 \\
0.185213 & -0.745725 & 0.194520 & -0.119195 & 0.181818 \\
-0.919232 & 0.096732 & 0.326906 & -0.603079 & 0.272727 \\
0.279141 & 0.532936 & 0.087364 & -0.581308 & 0.181818 \\
-0.206551 & -0.373195 & 0.161098 & 0.489027 & 0.090909
\end{array}\right], \\
\Gamma_{\text {orQ }}=1, \quad A_{Q}=A_{a a Q}^{+}=\left[\begin{array}{rrr}
0.433179 & -0.253237 \\
0.551005 & 0.609080
\end{array}\right], \quad n_{x Q}=2 \\
B_{Q} B_{Q}^{\prime}=\left[\begin{array}{rrr}
0.033508 & -0.018630 \\
-0.018630 & 0.030289
\end{array}\right], \quad C_{Q}^{\prime} C_{Q}=\left[\begin{array}{ll}
0 & 0 \\
0 & 0
\end{array}\right] \\
E_{a Q}^{+}=\left[\begin{array}{rrrr}
-0.769496 & 0.010023 & 0.448951 & -0.769496 \\
-0.090061 & 0.655677 & -1.044466 & -0.090061
\end{array}\right]
\end{gathered}
$$

and

$$
S_{x Q}=\left[\begin{array}{rr}
0.026333 & -0.021114 \\
-0.021114 & 0.043965
\end{array}\right], \quad T_{x \mathrm{Q}}=\left[\begin{array}{rr}
1.274771 & -0.555799 \\
-0.555799 & 1.764580
\end{array}\right]
$$

Step 5.M.3. The $n_{x \mathrm{P}} \times n_{x \mathrm{Q}}$ matrix $\Gamma$ is then given by

$$
\Gamma=\left[\begin{array}{rr}
-0.011218 & -0.106028 \\
0.185213 & -0.745725 \\
-0.919232 & 0.096732
\end{array}\right]
$$

and

$$
M=10^{2} \times\left[\begin{array}{rrrrr}
0.500695 & -0.334250 & 0.245016 & 0.082332 & 0.052125 \\
-0.442374 & 0.992368 & -0.260321 & 0.032515 & 0.253182 \\
0.616882 & -0.513348 & 0.588766 & 0.501907 & 0.261525 \\
1.074941 & -1.295698 & 0.921909 & 0.622391 & 0.172484 \\
-0.583103 & 1.526365 & -0.286520 & 0.180099 & 0.487850
\end{array}\right]
$$

Step 5.M.4. Finally, the infimum for the measurement feedback system is given by

$$
\gamma^{*}=13.638725
$$

Example 5.2.2. We consider a measurement feedback system (5.2.1) with $A$, $B, E, C_{2}, D_{2}, D_{22}$ being given as in Example 5.1.2 of Section 5.1 and

$$
C_{1}=\left[\begin{array}{llll}
1 & -2 & 3 & -4 \tag{5.2.31}
\end{array}\right], \quad D_{1}=0
$$

It is again simple to verify that the subsystem $\left(A, E, C_{1}, D_{1}\right)$, i.e., $\Sigma_{\mathrm{Q}}$, is observable and invertible with two unstable invariant zeros at $0.5 \pm j 0.5916$ and one infinite zero of order two. Hence, all assumptions are satisfied. Following Remark 5.2.3, we obtain

$$
\begin{gathered}
n_{x \mathrm{P}}=1, \quad \tilde{S}_{x \mathrm{P}}=0.5, \quad \tilde{T}_{x \mathrm{P}}=0.25, \\
\Gamma_{o r Q}=1, \quad n_{x Q}=2, \\
E_{a Q}^{+}=\left[\begin{array}{rrr}
-1.2230247 & -0.5241535 \\
1.1679942 & 0.9408842
\end{array}\right], \quad L_{a d \mathrm{Q}}^{+}=\left[\begin{array}{r}
-0.6289841 \\
1.3756377
\end{array}\right], \\
A_{a a Q}^{+}=\left[\begin{array}{rrrr}
0.8842105 & -0.5101735 \\
0.9753892 & 0.1157895
\end{array}\right], \\
B_{0 a \mathrm{Q}}^{+}=\emptyset, \\
D_{22,0 \mathrm{Q}}=\emptyset, \\
\tilde{S}_{x \mathrm{Q}}=\left[\begin{array}{rrrr}
0.5274947 & 0.5264991 \\
0.5264991 & 3.7365053
\end{array}\right], \quad \tilde{T}_{x \mathrm{Q}}=\left[\begin{array}{ll}
2.5810175 & 0.9950273 \\
0.9950273 & 3.2589825
\end{array}\right], \\
\Gamma=\left[\begin{array}{rrrr}
-1.2230247 & 1.1679942
\end{array}\right], \\
M=\left[\begin{array}{rrrrr}
1 & 0 & 0 & 0 & 0 \\
0 & 9.7252904 & 3.0610640 & -0.7439148 & 0 \\
0 & 2.0766328 & 0.9724337 & 0.1292764 & 0 \\
0 & 1.2428740 & 1.1820112 & 0.7056473 & 0 \\
0 & 0 & 0 & 0 & 5
\end{array}\right],
\end{gathered}
$$

and finally the infimum for the given system,

$$
\gamma^{*}=3.2088448
$$

### 5.3. Plants with Imaginary Axis Zeros

We present in this section a non-iterative computational algorithm for the measurement feedback system (5.2.1) whose subsystems $\Sigma_{P}$ and/or $\Sigma_{Q}$ have invariant zeros on the imaginary axis. The procedure is similar to the algorithm of the previous section, although it is slightly more complicated. It involves finding eigenspaces for the imaginary axis invariant zeros of $\Sigma_{P}$ and $\Sigma_{Q}$ and finding solutions to two extra Sylvester equations. We consider the system (5.2.1) which satisfies the following assumptions:

Assumption 5.Z.1: $(A, B)$ is stabilizable;
Assumption 5.Z.2: $\operatorname{Im}(E) \subset \mathcal{V}^{-}\left(\Sigma_{\mathrm{P}}\right)+\mathcal{S}^{-}\left(\Sigma_{\mathrm{P}}\right)$;

Assumption 5.Z.3: $\left(A, C_{1}\right)$ is detectable;
Assumption 5.Z.4: $\operatorname{Ker}\left(C_{2}\right) \supset \mathcal{V}^{-}\left(\Sigma_{Q}\right) \cap \mathcal{S}^{-}\left(\Sigma_{Q}\right) ;$ and
Assumption 5.Z.5: $D_{22}=0$.
(
We have the following step-by-step algorithm for computing $\gamma^{*}$. We note that it has some overlaps with that in the previous section. However, this is merely for completeness and to properly define matrices required in the computation of the infimum $\gamma^{*}$.

Step 5.Z.1. Transform the subsystem system $\Sigma_{\mathrm{P}}$, i.e., $\left(A, B, C_{2}, D_{2}\right)$ into the special coordinate basis described in Theorem 2.3.1. To all sub-matrices and transformations in the special coordinate basis of $\Sigma_{\mathrm{P}}$, we append the subscript ' P ' to signify their relation to the system $\Sigma_{\mathrm{P}}$. We also introduce an additional permutation matrix to the original state transformation such that the transformed state variables are arranged as

$$
\tilde{x}_{\mathrm{P}}=\left(\begin{array}{c}
x_{a \mathrm{P}}^{+}  \tag{5.3.1}\\
x_{b \mathrm{P}} \\
x_{a \mathrm{P}}^{0} \\
x_{a \mathrm{P}}^{-} \\
x_{c \mathrm{P}} \\
x_{d \mathrm{P}}
\end{array}\right) .
$$

Next, we compute

$$
\Gamma_{s \mathrm{P}}^{-1} E=\left[\begin{array}{c}
E_{a \mathrm{P}}^{+}  \tag{5.3.2}\\
E_{b \mathrm{P}} \\
E_{a \mathrm{P}}^{0} \\
E_{a \mathrm{P}}^{-} \\
E_{c \mathrm{P}} \\
E_{d \mathrm{P}}
\end{array}\right] .
$$

Note that Assumption 5.Z.2 implies $E_{b \mathrm{P}}=0$. Then define the following matrices:

$$
\begin{gather*}
A_{\mathrm{P}}:=\left[\begin{array}{ccc}
A_{a a \mathrm{P}}^{+} & L_{a b \mathrm{P}}^{+} C_{b \mathrm{P}} & 0 \\
0 & A_{b b \mathrm{P}} & 0 \\
0 & L_{a b \mathrm{P}}^{0} C_{b \mathrm{P}} & A_{a a \mathrm{P}}^{0}
\end{array}\right], \quad B_{\mathrm{P}}:=\left[\begin{array}{cc}
B_{0 a \mathrm{P}}^{+} & L_{a d \mathrm{P}}^{+} \\
B_{0 b \mathrm{P}} & L_{b d \mathrm{P}} \\
B_{0 a \mathrm{P}}^{0} & L_{a d \mathrm{P}}^{0}
\end{array}\right],  \tag{5.3.3}\\
E_{\mathrm{P}}:=\left[\begin{array}{c}
E_{a \mathrm{P}}^{+} \\
E_{b \mathrm{P}} \\
E_{a \mathrm{P}}^{0}
\end{array}\right], \tag{5.3.4}
\end{gather*}
$$

and

$$
C_{\mathbf{P}}:=\Gamma_{o \mathrm{P}}\left[\begin{array}{ccc}
0 & 0 & 0  \tag{5.3.5}\\
0 & 0 & 0 \\
0 & C_{b \mathrm{P}} & 0
\end{array}\right], \quad D_{\mathrm{P}}:=\Gamma_{o \mathrm{P}}\left[\begin{array}{cc}
I_{m_{0 \mathrm{P}}} & 0 \\
0 & C_{d \mathrm{P}} C_{d \mathrm{P}}^{\prime} \\
0 & 0
\end{array}\right]
$$

By some simple algebra, it is straightforward to show that

$$
C_{\mathrm{P}}^{\prime}\left[I-D_{\mathrm{P}}\left(D_{\mathrm{P}}^{\prime} D_{\mathrm{P}}\right)^{-1} D_{\mathrm{P}}^{\prime}\right] C_{\mathrm{P}}=\left[\begin{array}{ccc}
0 & \tilde{C}^{\prime} & 0  \tag{5.3.6}\\
0 & \tilde{C}_{b \mathrm{P}} & 0 \\
0 & 0 & 0
\end{array}\right],
$$

for some full row rank $\tilde{C}_{b \mathrm{P}}$,

$$
A_{\mathrm{P}}-B_{\mathrm{P}}\left(D_{\mathrm{P}}^{\prime} D_{\mathrm{P}}\right)^{-1} D_{\mathrm{P}}^{\prime} C_{\mathrm{P}}=\left[\begin{array}{ccc}
A_{a a \mathrm{P}}^{+} & \tilde{L}_{a b \mathrm{P}}^{+} \tilde{C}_{b \mathrm{P}} & 0  \tag{5.3.7}\\
0 & \tilde{A}_{b b \mathrm{P}} & 0 \\
0 & \tilde{L}_{a b \mathrm{P}}^{0} \tilde{C}_{b \mathrm{P}} & A_{a a \mathrm{P}}^{0}
\end{array}\right],
$$

and

$$
B_{\mathrm{P}}\left(D_{\mathrm{P}}^{\prime} D_{\mathrm{P}}\right)^{-1} B_{\mathrm{P}}^{\prime}=\left[\begin{array}{cc}
B_{0 a \mathrm{P}}^{+} & \tilde{L}_{a d \mathrm{P}}^{+}  \tag{5.3.8}\\
B_{0 b \mathrm{P}} & \tilde{L}_{b d \mathrm{P}} \\
B_{0 a \mathrm{P}}^{0} & \tilde{L}_{a d \mathrm{P}}^{0}
\end{array}\right] \cdot\left[\begin{array}{cc}
B_{0 a \mathrm{P}}^{+} & \tilde{L}_{a d \mathrm{P}}^{+} \\
B_{0 b \mathrm{P}} & \tilde{L}_{b d \mathrm{P}} \\
B_{0 a \mathrm{P}}^{0} & \tilde{L}_{a d \mathrm{P}}^{0}
\end{array}\right]^{\prime}
$$

for some appropriate $\tilde{L}_{a b \mathrm{P}}, \tilde{L}_{a b \mathrm{P}}^{0}, \tilde{L}_{a d \mathrm{P}}^{+}, \tilde{L}_{b d \mathrm{P}}$ and $\tilde{L}_{a d \mathrm{P}}^{0}$. Here we note that it can easily be verified that the pair $\left(\tilde{A}_{b b \mathrm{P}}, \tilde{C}_{b \mathrm{P}}\right)$ is observable provided that $\left(A_{b b \mathrm{P}}, C_{b \mathrm{P}}\right)$ is observable.

Step 5.Z.2. Define

$$
A_{x \mathrm{P}}:=\left[\begin{array}{cc}
A_{a a \mathrm{P}}^{+} & \tilde{L}_{a b \mathrm{P}}^{+} \tilde{C}_{b \mathrm{P}}  \tag{5.3.9}\\
0 & \tilde{A}_{b b \mathrm{P}}
\end{array}\right], \quad B_{x \mathrm{P}}:=\left[\begin{array}{cc}
B_{0 a \mathrm{P}}^{+} & \tilde{L}_{a d \mathrm{P}}^{+} \\
B_{0 b \mathrm{P}} & \tilde{L}_{b d \mathrm{P}}
\end{array}\right],
$$

and

$$
C_{x \mathrm{P}}:=\left[\begin{array}{ll}
0 & \tilde{C}_{b \mathrm{P}}
\end{array}\right], \quad E_{x \mathrm{P}}:=\left[\begin{array}{c}
E_{a \mathrm{P}}^{+}  \tag{5.3.10}\\
E_{b \mathrm{P}}
\end{array}\right] .
$$

Then we solve for the unique positive definite solution $S_{x P}$ of the Riccati equation,

$$
\begin{equation*}
A_{x \mathrm{P}} S_{x \mathrm{P}}+S_{x \mathrm{P}} A_{x \mathrm{P}}^{\prime}-B_{x \mathrm{P}} B_{x \mathrm{P}}^{\prime}+S_{x \mathrm{P}} C_{x \mathrm{P}}^{\prime} C_{x \mathrm{P}} S_{x \mathrm{P}}=0, \tag{5.3.11}
\end{equation*}
$$

together with the matrix $T_{x \mathrm{P}}$ defined by

$$
T_{x \mathrm{P}}:=\left[\begin{array}{cc}
T_{a x \mathrm{P}} & 0 \\
0 & 0
\end{array}\right]
$$

where $T_{a x \mathrm{P}}$ is the unique solution to the Lyapunov equation,

$$
\begin{equation*}
A_{a a \mathrm{P}}^{+} T_{a x \mathrm{P}}+T_{a x \mathrm{P}}\left(A_{a a \mathrm{P}}^{+}\right)^{\prime}=E_{a \mathrm{P}}^{+}\left(E_{a \mathrm{P}}^{+}\right)^{\prime} \tag{5.3.12}
\end{equation*}
$$

Next, solve the unique solution $Y_{x \mathrm{P}}$ of the following Sylvester equation,

$$
\begin{align*}
\left(A_{x \mathrm{P}}+S_{x \mathrm{P}} C_{x \mathrm{P}}^{\prime} C_{x \mathrm{P}}\right) Y_{x \mathrm{P}}+Y_{x \mathrm{P}}\left(A_{a a \mathrm{P}}^{0}\right)^{\prime} & +S_{x \mathrm{P}} C_{x \mathrm{P}}^{\prime}\left(\tilde{L}_{a b \mathrm{P}}^{0}\right)^{\prime} \\
& -B_{x \mathrm{P}}\left[\begin{array}{ll}
B_{0 a \mathrm{P}}^{0} & \tilde{L}_{a d \mathrm{P}}^{0}
\end{array}\right]^{\prime}=0 . \tag{5.3.13}
\end{align*}
$$

Let us denote the set of eigenvalues of $A_{a a \mathrm{P}}^{0}$ with a non-negative imaginary part as $\left\{j \omega_{\mathrm{P} 1}, \cdots, j \omega_{\mathrm{P} k_{\mathrm{P}}}\right\}$ and for $i=1, \cdots, k_{\mathrm{P}}$, choose complex matrices $V_{i \mathrm{P}}$, whose columns form a basis for the eigenspace,

$$
\begin{equation*}
\left\{x \in \mathbb{C}^{n_{a \mathrm{P}}^{0}} \mid x^{\mathrm{H}}\left(j \omega_{\mathrm{P} i} I-A_{a a \mathrm{P}}^{0}\right)=0\right\} \tag{5.3.14}
\end{equation*}
$$

where $n_{a \mathrm{P}}^{0}$ is the dimension of $A_{a a \mathrm{P}}^{0}$. Then define

$$
\begin{align*}
& F_{i \mathrm{P}}:=V_{i \mathrm{P}}^{\mathrm{H}}\left(\left[\begin{array}{ll}
B_{0 a \mathrm{P}}^{0} & \tilde{L}_{a d \mathrm{P}}^{0}
\end{array}\right]\left[\begin{array}{ll}
B_{0 a \mathrm{P}}^{0} & \tilde{L}_{a d \mathrm{P}}^{0}
\end{array}\right]^{\prime}+\tilde{L}_{a b \mathrm{P}}^{0}\left(\tilde{L}_{a b \mathrm{P}}^{0}\right)^{\prime}\right. \\
&\left.-\left[\left(\tilde{L}_{a b \mathrm{P}}^{0}\right)^{\prime}+C_{\mathrm{P}} Y_{\mathrm{P}}\right]^{\prime}\left[\left(\tilde{L}_{a b \mathrm{P}}^{0}\right)^{\prime}+C_{\mathrm{P}} Y_{\mathrm{P}}\right]\right) V_{i \mathrm{P}} \tag{5.3.15}
\end{align*}
$$

for $i=1, \cdots, k_{\mathrm{P}}$, and

$$
\begin{equation*}
F_{\mathrm{P}}:=\operatorname{blkdiag}\left\{F_{1 \mathrm{P}}, \cdots, F_{k_{\mathrm{P}}}\right\} . \tag{5.3.16}
\end{equation*}
$$

It is shown in [95] that $F_{\mathrm{P}}>0$. Also, define

$$
\begin{equation*}
G_{\mathrm{P}}:=\mathrm{blkdiag}\left\{V_{1 \mathrm{P}}^{\mathrm{H}} E_{a \mathrm{P}}^{0}\left(E_{a \mathrm{P}}^{0}\right)^{\prime} V_{1 \mathrm{P}}, \cdots, V_{k_{\mathrm{P}} \mathrm{P}}^{\mathrm{H}} E_{a \mathrm{P}}^{0}\left(E_{a \mathrm{P}}^{0}\right)^{\prime} V_{k_{\mathrm{PP}}}\right\} \tag{5.3.17}
\end{equation*}
$$

Step 5.Z.3. Transform the subsystem $\Sigma_{\mathrm{Q}}^{\star}$, i.e., $\left(A^{\prime}, C_{1}^{\prime}, E^{\prime}, D_{1}^{\prime}\right)$, into the special coordinate basis described in Theorem 2.3.1. Again we add here the subscript ' $Q$ ' to all sub-matrices and transformations in the special coordinate basis of the system $\Sigma_{\mathrm{Q}}^{\star}$ and re-arrange the transformed state variables as

$$
\tilde{x}_{\mathrm{Q}}=\left(\begin{array}{c}
x_{a \mathrm{Q}}^{+}  \tag{5.3.18}\\
x_{b \mathrm{Q}} \\
x_{a \mathrm{Q}}^{0} \\
x_{a \mathrm{Q}}^{-} \\
x_{c \mathrm{Q}} \\
x_{d \mathrm{Q}}
\end{array}\right) .
$$

Next, we compute

$$
\Gamma_{s Q}^{-1} C_{2}^{\prime}=\left[\begin{array}{c}
E_{a Q}^{+}  \tag{5.3.19}\\
E_{b Q} \\
E_{a Q}^{0} \\
E_{a Q}^{-} \\
E_{c Q} \\
E_{d Q}
\end{array}\right] .
$$

Note that Assumption 5.Z.4 implies $E_{b Q}=0$. Then define the following matrices:

$$
\begin{gather*}
A_{\mathrm{Q}}:=\left[\begin{array}{ccc}
A_{a a Q}^{+} & L_{a b Q}^{+} C_{b \mathrm{Q}} & 0 \\
0 & A_{b b Q} & 0 \\
0 & L_{a b Q}^{0} C_{b Q} & A_{a a Q}^{0}
\end{array}\right], \quad B_{Q}:=\left[\begin{array}{cc}
B_{0 a Q}^{+} & L_{a d Q}^{+} \\
B_{0 b Q} & L_{b d Q} \\
B_{0 a Q}^{0} & L_{a d Q}^{0}
\end{array}\right]  \tag{5.3.20}\\
E_{Q}:=\left[\begin{array}{c}
E_{a Q}^{+} \\
E_{b Q} \\
E_{a Q}^{0}
\end{array}\right] \tag{5.3.21}
\end{gather*}
$$

and

$$
C_{\mathrm{Q}}:=\Gamma_{o Q}\left[\begin{array}{ccc}
0 & 0 & 0  \tag{5.3.22}\\
0 & 0 & 0 \\
0 & C_{b Q} & 0
\end{array}\right], \quad D_{\mathrm{Q}}:=\Gamma_{o \mathrm{Q}}\left[\begin{array}{cc}
I_{m_{0 Q}} & 0 \\
0 & C_{d \mathrm{Q}} C_{d_{\mathrm{Q}}}^{\prime} \\
0 & 0
\end{array}\right]
$$

By some simple algebra, it is straightforward to show that

$$
C_{\mathrm{Q}}^{\prime}\left[I-D_{\mathrm{Q}}\left(D_{\mathrm{Q}}^{\prime} D_{\mathrm{Q}}\right)^{-1} D_{\mathrm{Q}}^{\prime}\right] C_{\mathrm{Q}}=\left[\begin{array}{ccc}
0 & \tilde{C}_{b q}^{\prime} \tilde{C}_{b Q} & 0  \tag{5.3.23}\\
0 & 0 & 0
\end{array}\right],
$$

for some full row rank $\tilde{C}_{b \mathrm{Q}}$, and

$$
A_{Q}-B_{Q}\left(D_{Q}^{\prime} D_{Q}\right)^{-1} D_{Q}^{\prime} C_{Q}=\left[\begin{array}{ccc}
A_{a a Q}^{+} & \tilde{L}_{a b Q}^{+} \tilde{C}_{b Q} & 0  \tag{5.3.24}\\
0 & \tilde{A}_{b b Q} & 0 \\
0 & \tilde{L}_{a b Q}^{0} \tilde{C}_{b Q} & A_{a a Q}^{0}
\end{array}\right]
$$

and

$$
B_{\mathrm{Q}}\left(D_{\mathrm{Q}}^{\prime} D_{\mathrm{Q}}\right)^{-1} B_{\mathrm{Q}}^{\prime}=\left[\begin{array}{cc}
B_{0 a \mathrm{Q}}^{+} & \tilde{L}_{a d \mathrm{Q}}^{+}  \tag{5.3.25}\\
B_{0 b Q} & \tilde{L}_{b d \mathrm{Q}} \\
B_{0 a \mathrm{Q}}^{0} & \tilde{L}_{a d \mathrm{Q}}^{0}
\end{array}\right] \cdot\left[\begin{array}{cc}
B_{0 a \mathrm{Q}}^{+} & \tilde{L}_{a d \mathrm{Q}}^{+} \\
B_{0 b \mathrm{Q}} & \tilde{L}_{b d \mathrm{Q}} \\
B_{0 a Q}^{0} & \tilde{L}_{a d \mathrm{Q}}^{0}
\end{array}\right]^{\prime}
$$

for some appropriate $\tilde{L}_{a b Q}, \tilde{L}_{a b Q}^{0}, \tilde{L}_{a d Q}^{+}, \tilde{L}_{b d Q}$ and $\tilde{L}_{a d Q}^{0}$. Here we note that it can easily be verified that the pair $\left(\tilde{A}_{b b Q}, \tilde{C}_{b Q}\right)$ is observable provided that $\left(A_{b b Q}, C_{b Q}\right)$ is observable.

Step 5.Z.4. Define

$$
A_{x Q}:=\left[\begin{array}{cc}
A_{a a Q}^{+} & \tilde{L}_{a b Q}^{+} \tilde{C}_{b Q}  \tag{5.3.26}\\
0 & \tilde{A}_{b b Q}
\end{array}\right], \quad B_{x Q}:=\left[\begin{array}{cc}
B_{0 a Q}^{+} & \tilde{L}_{a d Q}^{+} \\
B_{0 b Q} & \tilde{L}_{b d Q}
\end{array}\right]
$$

and

$$
C_{x Q}:=\left[\begin{array}{ll}
0 & \tilde{C}_{b Q}
\end{array}\right], \quad E_{x Q}=\left[\begin{array}{c}
E_{a Q}^{+}  \tag{5.3.27}\\
E_{b Q}
\end{array}\right]
$$

Then we solve for the unique positive definite solution $S_{x Q}$ of the Riccati equation,

$$
\begin{equation*}
A_{x Q} S_{x Q}+S_{x Q} A_{x Q}^{\prime}-B_{x Q} B_{x Q}^{\prime}+S_{x Q} C_{x Q}^{\prime} C_{x Q} S_{x Q}=0 \tag{5.3.28}
\end{equation*}
$$

together with the matrix $T_{x Q}$ defined by

$$
T_{x \mathrm{Q}}:=\left[\begin{array}{cc}
T_{a x \mathrm{Q}} & 0 \\
0 & 0
\end{array}\right]
$$

where $T_{a x Q}$ is the unique solution to the Lyapunov equation,

$$
\begin{equation*}
A_{a a Q}^{+} T_{a x Q}+T_{a x Q}\left(A_{a a Q}^{+}\right)^{\prime}=E_{a Q}^{+}\left(E_{a Q}^{+}\right)^{\prime} \tag{5.3.29}
\end{equation*}
$$

Next, solve the unique solution $Y_{x Q}$ of the following Sylvester equation,

$$
\begin{align*}
\left(A_{x Q}+S_{x Q} C_{x Q}^{\prime} C_{x Q}\right) Y_{x Q}+Y_{x Q}\left(A_{a a Q}^{0}\right)^{\prime} & +S_{x Q} C_{x Q}^{\prime}\left(\tilde{L}_{a b Q}^{0}\right)^{\prime} \\
& -B_{x Q}\left[\begin{array}{ll}
B_{0 a Q}^{0} & \tilde{L}_{a d Q}^{0}
\end{array}\right]^{\prime}=0 \tag{5.3.30}
\end{align*}
$$

Let us denote the set of eigenvalues of $A_{a a Q}^{0}$ with a non-negative imaginary part as $\left\{j \omega_{\mathrm{Q} 1}, \cdots, j \omega_{\mathrm{Q} k_{\mathrm{Q}}}\right\}$ and for $i=1, \cdots, k_{\mathrm{Q}}$, choose complex matrices $V_{i \mathrm{Q}}$, whose columns form a basis for the eigenspace,

$$
\begin{equation*}
\left\{x \in \mathbb{C}^{n_{a Q}^{0}} \mid x^{\mathrm{H}}\left(j \omega_{\mathrm{Q} i} I-A_{a a Q}^{0}\right)=0\right\} \tag{5.3.31}
\end{equation*}
$$

where $n_{a Q}^{0}$ is the dimension of $A_{a a Q}^{0}$. Then define

$$
\begin{align*}
& F_{i Q}:=V_{i Q}^{\mathrm{H}}\left(\left[\begin{array}{ll}
B_{0 a Q}^{0} & \tilde{L}_{a d Q}^{0}
\end{array}\right]\left[\begin{array}{ll}
B_{0 a Q}^{0} & \tilde{L}_{a d Q}^{0}
\end{array}\right]^{\prime}+\tilde{L}_{a b Q}^{0}\left(\tilde{L}_{a b Q}^{0}\right)^{\prime}\right. \\
&\left.-\left[\left(\tilde{L}_{a b Q}^{0}\right)^{\prime}+C_{Q} Y_{Q}\right]^{\prime}\left[\left(\tilde{L}_{a b Q}^{0}\right)^{\prime}+C_{Q} Y_{Q}\right]\right) V_{i Q} \tag{5.3.32}
\end{align*}
$$

for $i=1, \cdots, k_{\mathrm{Q}}$, and

$$
\begin{equation*}
F_{\mathrm{Q}}:=\operatorname{blkdiag}\left\{F_{1 \mathrm{Q}}, \cdots, F_{k_{\mathrm{Q} \mathrm{Q}}}\right\} \tag{5.3.33}
\end{equation*}
$$

Again, it can be shown that $F_{\mathrm{Q}}>0$. Also, define

$$
\begin{equation*}
G_{\mathrm{Q}}:=\operatorname{blkdiag}\left\{V_{1 \mathrm{Q}}^{\mathrm{H}} E_{a Q}^{0}\left(E_{a Q}^{0}\right)^{\prime} V_{1 Q}, \cdots, V_{k_{Q} \mathrm{Q}}^{\mathrm{H}} E_{a \mathrm{Q}}^{0}\left(E_{a \mathrm{Q}}^{0}\right)^{\prime} V_{k_{Q Q}}\right\} \tag{5.3.34}
\end{equation*}
$$

Step 5.Z.5. Define

$$
\begin{equation*}
n_{x \mathrm{P}}:=\operatorname{dim}\left\{\mathbb{R}^{n} / \mathcal{S}^{+}\left(\Sigma_{\mathrm{P}}\right)\right\}-n_{a \mathrm{P}}^{0}, \tag{5.3.35}
\end{equation*}
$$

and

$$
\begin{equation*}
n_{x Q}:=\operatorname{dim}\left\{\mathbb{R}^{n} / \mathcal{S}^{+}\left(\Sigma_{Q}^{\star}\right)\right\}-n_{a Q}^{0}, \tag{5.3.36}
\end{equation*}
$$

and partition

$$
\Gamma_{s \mathrm{P}}^{-1}\left(\Gamma_{s \mathrm{Q}}^{-1}\right)^{\prime}=\left[\begin{array}{cc}
\Gamma & \star  \tag{5.3.37}\\
\star & \star
\end{array}\right],
$$

where $\Gamma$ is of dimension $n_{x \mathrm{P}} \times n_{x \mathrm{Q}}$. Finally, define a constant matrix

$$
M:=\left[\begin{array}{cccc}
G_{\mathrm{P}} F_{\mathrm{P}}^{-1} & 0 & 0 & 0  \tag{5.3.38}\\
0 & T_{x \mathrm{P}} S_{x \mathrm{P}}^{-1}+\Gamma S_{x \mathrm{Q}}^{-1} \Gamma^{\prime} S_{x \mathrm{P}}^{-1} & -\Gamma S_{x \mathrm{Q}}^{-1} & 0 \\
0 & -T_{x \mathrm{Q}} S_{x \mathrm{Q}}^{-1} \Gamma^{\prime} S_{x \mathrm{P}}^{-1} & T_{x \mathrm{Q}} S_{x \mathrm{Q}}^{-1} & 0 \\
0 & 0 & 0 & G_{\mathrm{Q}} F_{\mathrm{Q}}^{-1}
\end{array}\right] .
$$

Step 5.Z.6. The infimum $\gamma^{*}$ is then given by

$$
\begin{equation*}
\gamma^{*}=\sqrt{\lambda_{\max }(M)} \tag{5.3.39}
\end{equation*}
$$

This will be justified in Theorem 5.3.1 below.
We have the following main theorem.
Theorem 5.3.1. Consider the given measurement feedback system (5.2.1). Then under Assumptions 5.Z.1 to 5.Z.5, its infimum is given by (5.3.39). T

Proof. Following the results of Scherer [96], it can be show that

$$
\begin{equation*}
\gamma>\gamma_{\mathrm{P}}^{*}:=\max \left\{\sqrt{\lambda_{\max }\left(T_{x \mathrm{P}} S_{x \mathrm{P}}^{-1}\right)}, \sqrt{\lambda_{\max }\left(G_{\mathrm{P}} F_{\mathrm{P}}^{-1}\right)}\right\} \tag{5.3.40}
\end{equation*}
$$

if and only if the following algebraic Riccati inequality,

$$
\begin{aligned}
& {\left[A_{\mathrm{P}}-B_{\mathrm{P}}\left(D_{\mathrm{P}}^{\prime} D_{\mathrm{P}}\right)^{-1} D_{\mathrm{P}} C_{\mathrm{P}}\right] X+X\left[A_{\mathrm{P}}-B_{\mathrm{P}}\left(D_{\mathrm{P}}^{\prime} D_{\mathrm{P}}\right)^{-1} D_{\mathrm{P}} C_{\mathrm{P}}\right]^{\prime}} \\
& \quad+\gamma^{-2} E_{\mathrm{P}} E_{\mathrm{P}}^{\prime}+X C_{\mathrm{P}}\left[I-D_{\mathrm{P}}\left(D_{\mathrm{P}}^{\prime} D_{\mathrm{P}}\right)^{-1} D_{\mathrm{P}}^{\prime}\right] C_{\mathrm{P}} X-B_{\mathrm{P}}\left(D_{\mathrm{P}}^{\prime} D_{\mathrm{P}}\right)^{-1} B_{\mathrm{P}}^{\prime}<0
\end{aligned}
$$

has a positive definite solution. Then it follows from the results of [95] and [96] (see also Theorem 3.1.2) and some simple algebraic manipulations that for $\gamma>\gamma_{\mathrm{P}}^{*}$, the positive semi-definite matrix $P(\gamma)$ given by

$$
P(\gamma)=\left(\Gamma_{s \mathrm{P}}^{-1}\right)^{\prime}\left[\begin{array}{cc}
\left(S_{x \mathrm{P}}-\gamma^{-2} T_{x \mathrm{P}}\right)^{-1} & 0  \tag{5.3.41}\\
0 & 0
\end{array}\right] \Gamma_{s \mathrm{P}}^{-1}
$$

is the lower limit point of the set

$$
\begin{aligned}
\left\{P>0 \mid \exists F:(A+B F)^{\prime} P+P(A+B F)\right. & +\gamma^{-2} P E E^{\prime} P \\
& \left.+\left(C_{2}+D_{2} F\right)^{\prime}\left(C_{2}+D_{2} F\right)<0\right\}
\end{aligned}
$$

Moreover, such a $P(\gamma)$ does not exist when $\gamma<\gamma_{\mathrm{P}}^{*}$. By dual reasoning, one can show that

$$
\begin{equation*}
\gamma>\gamma_{Q}^{*}:=\max \left\{\sqrt{\lambda_{\max }\left(T_{x Q} S_{x Q}^{-1}\right)}, \sqrt{\lambda_{\max }\left(G_{Q} F_{Q}^{-1}\right)}\right\} \tag{5.3.42}
\end{equation*}
$$

if and only if the following algebraic Riccati inequality,

$$
\begin{aligned}
& {\left[A_{\mathrm{Q}}-B_{\mathrm{Q}}\left(D_{\mathrm{Q}}^{\prime} D_{\mathrm{Q}}\right)^{-1} D_{\mathrm{Q}} C_{\mathrm{Q}}\right] Z+Z\left[A_{\mathrm{Q}}-B_{\mathrm{Q}}\left(D_{\mathrm{Q}}^{\prime} D_{\mathrm{Q}}\right)^{-1} D_{\mathrm{Q}} C_{\mathrm{Q}}\right]^{\prime}} \\
& \quad+\gamma^{-2} E_{\mathrm{Q}} E_{\mathrm{Q}}^{\prime}+Z C_{\mathrm{Q}}\left[I-D_{\mathrm{Q}}\left(D_{\mathrm{Q}}^{\prime} D_{\mathrm{Q}}\right)^{-1} D_{\mathrm{Q}}^{\prime}\right] C_{\mathrm{Q}} Z-B_{\mathrm{Q}}\left(D_{\mathrm{Q}}^{\prime} D_{\mathrm{Q}}\right)^{-1} B_{\mathrm{Q}}^{\prime}<0
\end{aligned}
$$

has a positive definite solution. For $\gamma>\gamma_{Q}^{*}$, the positive semi-definite matrix $Q(\gamma)$ given by

$$
Q(\gamma)=\left(\Gamma_{s Q}^{-1}\right)^{\prime}\left[\begin{array}{cc}
\left(S_{x Q}-\gamma^{-2} T_{x Q}\right)^{-1} & 0  \tag{5.3.43}\\
0 & 0
\end{array}\right] \Gamma_{s Q}^{-1}
$$

is the lower limit point of the set

$$
\begin{aligned}
\left\{Q>0 \mid \exists K:\left(A+K C_{1}\right) Q+Q\left(A+K C_{1}\right)^{\prime}\right. & +\gamma^{-2} Q C_{2}^{\prime} C_{2} Q \\
& \left.+\left(E+K D_{1}\right)\left(E+K D_{1}\right)^{\prime}<0\right\}
\end{aligned}
$$

Again, such a $Q(\gamma)$ does not exist when $\gamma<\gamma_{\mathbf{Q}}^{*}$. Now, let us define

$$
\begin{equation*}
\gamma_{\mathrm{PQ}}^{*}:=\max \left\{\sqrt{\lambda_{\max }\left(T_{x \mathrm{P}} S_{x \mathrm{P}}^{-1}\right)}, \sqrt{\lambda_{\max }\left(T_{x \mathrm{Q}} S_{x \mathrm{Q}}^{-1}\right)}\right\} \tag{5.3.44}
\end{equation*}
$$

and

$$
\begin{equation*}
\gamma_{\text {coup }}^{*}:=\sup \left\{\gamma \in\left(\gamma_{\mathrm{PQ}}^{*}, \infty\right) \mid \rho[P(\gamma) Q(\gamma)]<\gamma^{2}\right\} \tag{5.3.45}
\end{equation*}
$$

where $P(\gamma)$ and $Q(\gamma)$ are as given in (5.3.41) and (5.3.43), respectively. Then following the results of Scherer [96], it can easily be shown that

$$
\begin{equation*}
\gamma^{*}=\max \left\{\gamma_{\text {coup }}^{*}, \sqrt{\lambda_{\max }\left(G_{\mathrm{P}} F_{\mathrm{P}}^{-1}\right)}, \sqrt{\lambda_{\max }\left(G_{\mathrm{Q}} F_{\mathrm{Q}}^{-1}\right)}\right\} . \tag{5.3.46}
\end{equation*}
$$

Also, it follows from Theorem 5.2.1 that

$$
\gamma_{c o u p}^{*}=\left\{\lambda_{\max }\left[\begin{array}{cc}
T_{x \mathrm{P}} S_{x \mathrm{P}}^{-1}+\Gamma S_{x \mathrm{Q}}^{-1} \Gamma^{\prime} S_{x \mathrm{P}}^{-1} & -\Gamma S_{x \mathrm{Q}}^{-1}  \tag{5.3.47}\\
-T_{x \mathrm{Q}} S_{x \mathrm{Q}}^{-1} \Gamma^{\prime} S_{x \mathrm{P}}^{-1} & T_{x \mathrm{Q}} S_{x \mathrm{Q}}^{-1}
\end{array}\right]\right\}^{\frac{1}{2}}
$$

Hence, the result of Theorem 5.3.1 follows.

We illustrate our main result in the following example.

Example 5.3.1. Consider a given system characterized by

$$
\begin{align*}
A & =\left[\begin{array}{rrrrr}
0 & 1 & 1 & -1 & 1 \\
0 & 0 & 0 & 0 & 1 \\
0 & 1 & 0 & 0 & 1 \\
1 & 1 & 1 & 0 & 1 \\
1 & 1 & 1 & 1 & 0
\end{array}\right], \quad B=\left[\begin{array}{ll}
0 & 0 \\
0 & 0 \\
1 & 0 \\
0 & 0 \\
0 & 1
\end{array}\right], \quad E=\left[\begin{array}{ll}
1 & 1 \\
0 & 0 \\
0 & 0 \\
2 & 1 \\
1 & 2
\end{array}\right],  \tag{5.3.48}\\
C_{1} & =\left[\begin{array}{rrrrr}
-1 & 11 & -21.876238 & -4.2239 & -2.425699 \\
1 & 2 & 3 & 2 & 1
\end{array}\right], \quad D_{1}=\left[\begin{array}{ll}
1 & 0 \\
0 & 0
\end{array}\right], \tag{5.3.49}
\end{align*}
$$

and

$$
C_{2}=\left[\begin{array}{lllll}
0 & 0 & 1 & 0 & 0  \tag{5.3.50}\\
0 & 0 & 0 & 0 & 1 \\
0 & 1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0
\end{array}\right], \quad D_{2}=\left[\begin{array}{ll}
1 & 0 \\
0 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right], \quad D_{22}=0
$$

First, it is simple to verify that the subsystem $\Sigma_{\mathrm{P}}$ is left invertible with two invariant zeros at $\pm j$ and Assumption 5.Z.2 is satisfied. Applying the special coordinate basis transformation to $\Sigma_{\mathrm{P}}$, we have

$$
\begin{gathered}
\Gamma_{s \mathrm{P}}=\left[\begin{array}{rrrrr}
0 & 0 & 0 & -1 & 0 \\
1.3660254 & 0.3660254 & 0 & 0 & 0 \\
0.1988066 & 1.9900945 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0 \\
0 & 0 & 0 & 0 & 1
\end{array}\right], \\
A_{x \mathrm{P}}=\left[\begin{array}{rr}
-0.1614784 & 0.2246812 \\
0.6026457 & -0.8385216
\end{array}\right], \quad B_{x \mathrm{P}}=\left[\begin{array}{rr}
0.6040578 & -0.1762197 \\
0.4723969 & 0.4878984
\end{array}\right], \\
C_{x \mathrm{P}}=\left[\begin{array}{rr}
1.3544397 & 0.2665382 \\
0.2665382 & 2.0058434
\end{array}\right], \quad E_{b \mathrm{P}}=\left[\begin{array}{ll}
0 & 0 \\
0 & 0
\end{array}\right], \\
A_{a a \mathrm{P}}^{0}=\left[\begin{array}{rr}
0 & -1 \\
1 & 0
\end{array}\right], \quad \tilde{L}_{a b \mathrm{P}}^{0}=\left[\begin{array}{rr}
0.9489977 & 1.0485243 \\
-0.9489977 & -1.0485243
\end{array}\right],
\end{gathered}
$$

and

$$
\left[\begin{array}{ll}
B_{0 a \mathrm{P}}^{0} & \tilde{L}_{a d \mathrm{P}}^{0}
\end{array}\right]=\left[\begin{array}{rr}
0 & 1 \\
0 & -1
\end{array}\right], \quad E_{a \mathrm{P}}^{0}=\left[\begin{array}{rr}
2 & 1 \\
-1 & -1
\end{array}\right] .
$$

Following Step 5.Z.2, we obtain

$$
\begin{gathered}
S_{x \mathrm{P}}=\left[\begin{array}{rr}
0.6180716 & -0.2516670 \\
-0.2516670 & 0.7339429
\end{array}\right], \quad T_{x \mathrm{P}}=\left[\begin{array}{ll}
0 & 0 \\
0 & 0
\end{array}\right] \\
Y_{x \mathrm{P}}=\left[\begin{array}{rr}
-0.6928337 & -0.0822109 \\
-0.3161228 & 0.3068152
\end{array}\right]
\end{gathered}
$$

and

$$
F_{\mathrm{P}}=2.3885733, \quad G_{\mathrm{P}}=3.5
$$

Next, the subsystem $\Sigma_{Q}$ is invertible and of nonminimum phase with invariant zeros at $\{0.078944, \pm j 2.302011,-4.095803\}$. Hence, Assumption 5.Z.4 is automatically satisfied. Applying the special coordinate basis transformation to $\Sigma_{\mathrm{Q}}^{\star}$, we obtain

$$
\begin{gathered}
\Gamma_{s Q}=\left[\begin{array}{rrrrr}
0.2148444 & 0.0018481 & 0.2169145 & 0.0698280 & 0.2 \\
0.5503097 & 0.6645646 & -0.6352193 & 0.8023543 & 0.4 \\
-0.7990597 & -0.7456317 & -0.5938518 & -0.5805731 & 0.6 \\
-0.0941402 & -0.0440333 & 0.3437855 & 0.0892284 & 0.4 \\
-0.0603521 & 0.0210926 & -0.2803500 & -0.0795282 & 0.2
\end{array}\right] \\
A_{x Q}=A_{a a Q}^{+}=0.0789442, \quad B_{x Q}=\left[\begin{array}{llll}
2.3596219 & -0.1725085
\end{array}\right], \quad C_{x Q}=0, \\
E_{a Q}^{+}=\left[\begin{array}{llll}
0.1593412 & 0.0009204 & 0.0116587 & 0.1593412
\end{array}\right]
\end{gathered}
$$

and

$$
\begin{gathered}
A_{a a Q}^{0}=\left[\begin{array}{l}
0.8733954 \\
0.4222493 \\
-14.3566212 \\
-0.8733953
\end{array}\right] \\
{\left[\begin{array}{ll}
B_{0 a Q}^{0} & \tilde{L}_{a d Q}^{0}
\end{array}\right]=\left[\begin{array}{rrr}
13.8502316 & -10.8089077 \\
0.3251762 & -1.3752299
\end{array}\right],} \\
E_{a Q}^{0}=\left[\begin{array}{llll}
-1.9958628 & 6.3511003 & -0.7973732 & -1.9958628 \\
-0.5082606 & 0.0920508 & -0.4908900 & -0.5082606
\end{array}\right] .
\end{gathered}
$$

Following Step 5.Z.4, we have

$$
S_{x Q}=35.4527292, \quad T_{x Q}=0.3224810, \quad Y_{x Q}=\left[\begin{array}{ll}
-5.2529064 & 93.6614674
\end{array}\right]
$$

and

$$
F_{\mathrm{Q}}=8.4694885, \quad G_{\mathrm{Q}}=35.4527292
$$

Finally, evaluate

$$
M=\left[\begin{array}{rrrrr}
1.4653098 & 0 & 0 & 0 & 0 \\
0 & -0.0000103 & -0.0000451 & 0.0003744 & 0 \\
0 & 0.0000632 & 0.0002763 & -0.0022958 & 0 \\
0 & -0.0002503 & -0.0010946 & 0.0090961 & 0 \\
0 & 0 & 0 & 0 & 0.2110284
\end{array}\right]
$$

We obtain

$$
\gamma^{*}=\sqrt{\lambda_{\max }(M)}=1.2104998
$$

E

## Chapter 6

## Solutions to Continuoustime $H_{\infty}$ Problem

The main contribution of this chapter is to provide closed-form solutions to the $H_{\infty}$ suboptimal control problem for continuous-time systems. Here by closed-form solutions we mean solutions which are explicitly parameterized in terms of $\gamma$ and are obtained without explicitly requiring a value for $\gamma$. Hence one can easily tune the parameter $\gamma$ to obtain the desired level of disturbance attenuation. Such a design can be called a 'one-shot' design. We provide these closed-form solutions for a class of singular $H_{\infty}$ suboptimal control problems for which the subsystem from the control input to the controlled output and the subsystem from the disturbance to the measurement output satisfy certain geometric conditions and some other minor assumptions, namely, Assumptions 5.M. 1 to 5.M. 7 of Chapter 5. Moreover, for this class of systems we also provide conditions under which the $H_{\infty}$ optimal control problem via state feedback has a solution. Explicit expressions for the solutions will also be given. Finally the issue of pole-zero cancellations in the closed-loop system resulting from the $H_{\infty}$ optimal or suboptimal state or output feedback control laws are examined.

Some significant attributes of our method of generating the closed-form solutions in the $H_{\infty}$ suboptimal control problem are as follows:

1. No $H_{\infty}$-CARE's are solved in generating the closed-form solutions. As a result, all the numerical difficulties associated with the $H_{\infty}$-CARE's are alleviated.
2. The value for $\gamma$ can be adjusted on line when the closed-form solution to the $H_{\infty}$ suboptimal control problem is implemented using either software or hardware. Since the effect of such a 'knob' on the performance and the
robustness of the closed-loop system is straightforward, it should be very appealing from a practical point of view.
3. Having closed-form solutions to the $H_{\infty}$ suboptimal control problem enables us to understand the behavior of the controller (i.e., high-gain, bandwidth, etc.) as the parameter $\gamma$ approaches the infimum value of the $H_{\infty}$ norm of $T_{h w}$ over all stabilizing controllers.

The above mentioned results were reported in Saberi, Chen and Lin [86]. In the case when Assumptions 5.M.1 and 5.M.7 are not satisfied, a similar method will also be adapted to compute $\gamma$-suboptimal solutions. It is, however, no longer a closed-form one. The outline of this chapter is as follows: Section 6.1 gives a closed-form solution to the $H_{\infty}$ suboptimal state feedback control problem, while Section 6.2 provides a closed-form solution (full order controller) to the $H_{\infty}$ suboptimal measurement feedback control problem. A reduced order $\gamma$-suboptimal controller design method is introduced in Section 6.3. Finally, all main results are to be proved in Section 6.4.

### 6.1. Full State Feedback

We consider in this section the $H_{\infty}$ optimization problem for the following full state feedback systems characterized by

$$
\left\{\begin{array}{l}
\dot{x}=A x+B u+E w  \tag{6.1.1}\\
y=x \\
h=C_{2} x+\iota_{2} u+D_{22} w
\end{array}\right.
$$

where $x \in \mathbb{R}^{n}$ is the state, $u \in \mathbb{R}^{m}$ is the control input, $w \in \mathbb{R}^{q}$ is the external disturbance input, and $h \in \mathbb{R}^{\ell}$ is the controlled output of $\Sigma$. Again, we let $\Sigma_{\mathrm{P}}$ be the subsystem characterized by the matrix quadruple $\left(A, B, C_{2}, D_{2}\right)$. As in Section 5.1 of Chapter 5 , we first make the following assumptions:

Assumption 6.F.1: $(A, B)$ is stabilizable;
Assumption 6.F.2: $\Sigma_{\mathrm{P}}$ has no invariant zero on the imaginary axis;
Assumption 6.F.3: $\operatorname{Im}(E) \subset \mathcal{V}^{-}\left(\Sigma_{\mathrm{P}}\right)+\mathcal{S}^{-}\left(\Sigma_{\mathrm{P}}\right)$; and
Assumption 6.F.4: $D_{22}=0$.
We introduce a procedure for obtaining the closed-form solutions for the $H_{\infty}$ suboptimal state feedback control problem utilizing an asymptotic timescale and eigenstructure assignment (ATEA). The concept of the ATEA design
procedure was proposed originally in Saberi and Sannuti [90] and its complete time-scale properties and Lyapunov stability analysis were done in Chen [10]. It uses the special coordinate basis of the given system (See Theorem 2.3.1). We also give conditions under which the $H_{\infty}$ optimal control problem has a solution. Furthermore, explicit expressions for these optimal solutions will be given. The following is a step-by-step algorithm to construct the closed-form of the $\gamma$-suboptimal state feedback laws, which are explicitly parameterized by $\gamma>\gamma^{*}$ and a tuning parameter $\varepsilon$.

Step 6.F.1: Transform the system $\Sigma_{\mathbf{P}}$ into the special coordinate basis as given by Theorem 2.3.1 in Chapter 2. To all sub-matrices and transformations in the special coordinate basis of $\Sigma_{\mathrm{P}}$, we append a subscript $\mathrm{r}_{\mathrm{r}}$ to signify their relation to the system $\Sigma_{\mathbf{P}}$. We also choose the output transformation $\Gamma_{o P}$ to have the following form:

$$
\Gamma_{o \mathrm{P}}=\left[\begin{array}{cc}
I_{m_{\mathrm{OP}}} & 0  \tag{6.1.2}\\
0 & \Gamma_{o r \mathrm{P}}
\end{array}\right]
$$

where $m_{0 \mathrm{P}}=\operatorname{rank}\left(D_{2}\right)$. Next, we compute

$$
\bar{E}=\Gamma_{s \mathrm{P}}^{-1} E=\left[\begin{array}{c}
E_{a \mathrm{P}}^{+}  \tag{6.1.3}\\
E_{b \mathrm{P}} \\
E_{a \mathrm{P}}^{-} \\
E_{c \mathrm{P}} \\
E_{d \mathrm{P}}
\end{array}\right]
$$

Note that Assumption 6.F. 3 implies $E_{b \mathrm{P}} \equiv 0$. Also, for economy of notation, we denote $n_{x \mathrm{P}}$ the dimension of $\mathbb{R}^{n} / \mathcal{S}^{+}\left(\Sigma_{\mathrm{P}}\right)$. Note that $n_{x \mathrm{P}}=0$ if and only if the system $\Sigma_{\mathrm{P}}$ is right invertible and is of minimum phase. Next, define

$$
\begin{gathered}
A_{11 \mathrm{P}}=\left[\begin{array}{cc}
A_{a a \mathrm{P}}^{+} & L_{a b \mathrm{P}}^{+} C_{b \mathrm{P}} \\
0 & A_{b b \mathrm{P}}
\end{array}\right], \quad B_{11 \mathrm{P}}=\left[\begin{array}{c}
B_{0 a \mathrm{P}}^{+} \\
B_{0 b \mathrm{P}}
\end{array}\right], \quad A_{13 \mathrm{P}}=\left[\begin{array}{c}
L_{a d \mathrm{P}}^{+} \\
L_{b d \mathrm{P}}
\end{array}\right], \\
C_{21 \mathrm{P}}=\Gamma_{o r \mathrm{P}}\left[\begin{array}{cc}
0 & 0 \\
0 & C_{b \mathrm{P}}
\end{array}\right], \quad C_{23 \mathrm{P}}=\Gamma_{o r \mathrm{P}}\left[\begin{array}{c}
C_{d \mathrm{P}} C_{d \mathrm{P}}^{\prime} \\
0
\end{array}\right], \quad E_{x \mathrm{P}}=\left[\begin{array}{c}
E_{a \mathrm{P}}^{+} \\
E_{b \mathrm{P}}
\end{array}\right],
\end{gathered}
$$

and

$$
\begin{gathered}
A_{x \mathrm{P}}=A_{11 \mathrm{P}}-A_{13 \mathrm{P}}\left(C_{23 \mathrm{P}}^{\prime} C_{23 \mathrm{P}}\right)^{-1} C_{23 \mathrm{P}}^{\prime} C_{21 \mathrm{P}}, \\
B_{x \mathrm{P}} B_{x \mathrm{P}}^{\prime}=B_{11 \mathrm{P}} B_{11 \mathrm{P}}^{\prime}+A_{13 \mathrm{P}}\left(C_{23 \mathrm{P}}^{\prime} C_{23 \mathrm{P}}\right)^{-1} A_{13 \mathrm{P}}^{\prime}, \\
C_{x \mathrm{P}}^{\prime} C_{x \mathrm{P}}=C_{21 \mathrm{P}}^{\prime} C_{21 \mathrm{P}}-C_{21 \mathrm{P}}^{\prime} C_{21 \mathrm{P}}\left(C_{23 \mathrm{P}}^{\prime} C_{23 \mathrm{P}}\right)^{-1} C_{23 \mathrm{P}}^{\prime} C_{21 \mathrm{P}}
\end{gathered}
$$

Step 6.F.2: Solve for the unique positive definite solution $S_{x \mathrm{P}}$ of the algebraic matrix Riccati equation,

$$
\begin{equation*}
A_{x \mathrm{P}} S_{x \mathrm{P}}+S_{x \mathrm{P}} A_{x \mathrm{P}}^{\prime}-B_{x \mathrm{P}} B_{x \mathrm{P}}^{\prime}+S_{x \mathrm{P}} C_{x \mathrm{P}}^{\prime} C_{x \mathrm{P}} S_{x \mathrm{P}}=0 \tag{6.1.4}
\end{equation*}
$$

together with the matrix $T_{x \mathrm{P}}$ defined by

$$
T_{x \mathrm{P}}=\left[\begin{array}{cc}
T_{a a \mathrm{P}} & 0  \tag{6.1.5}\\
0 & 0
\end{array}\right]
$$

where $T_{a a \mathrm{P}}$ is the unique semi-positive solution of the algebraic matrix Lyapunov equation,

$$
\begin{equation*}
A_{a a \mathrm{P}}^{+} T_{a a \mathrm{P}}+T_{a a \mathrm{P}}\left(A_{a a \mathrm{P}}^{+}\right)^{\prime}=E_{a \mathrm{P}}^{+}\left(E_{a \mathrm{P}}^{+}\right)^{\prime} \tag{6.1.6}
\end{equation*}
$$

Then it was shown in Section 5.1 of Chapter 5 that the infimum for the given system (6.1.1) is given by

$$
\begin{equation*}
\gamma^{*}=\sqrt{\lambda_{\max }\left(T_{x \mathrm{P}} S_{x \mathrm{P}}^{-1}\right)} \tag{6.1.7}
\end{equation*}
$$

Then, for any $\gamma>\gamma^{*}$, we define

$$
F_{11}(\gamma):=\left[\begin{array}{cc}
F_{a 0}^{+}(\gamma) & F_{b 0}(\gamma)  \tag{6.1.8}\\
F_{a 1}^{+}(\gamma) & F_{b 1}(\gamma)
\end{array}\right]=\left[\begin{array}{c}
B_{11 \mathrm{P}}^{\prime} P_{x} \\
\left(C_{23 \mathrm{P}}^{\prime} C_{23 \mathrm{P}}\right)^{-1}\left[A_{13 \mathrm{P}}^{\prime} P_{x}+C_{23 \mathrm{P}}^{\prime} C_{21 \mathrm{P}}\right]
\end{array}\right],
$$

where

$$
\begin{equation*}
P_{x}:=\left(S_{x \mathrm{P}}-\gamma^{-2} T_{x \mathrm{P}}\right)^{-1}, \tag{6.1.9}
\end{equation*}
$$

and define

$$
A_{11 \mathrm{P}}^{c}:=A_{11 \mathrm{P}}-\left[\begin{array}{ll}
B_{11 \mathrm{P}} & A_{13 \mathrm{P}}
\end{array}\right] F_{11}(\gamma) .
$$

We will show later on that the eigenvalues of $A_{11 \mathrm{p}}^{c}$ are in $\mathbb{C}^{-}$. Let us partition $\left[F_{a 1}^{+}(\gamma) \quad F_{b 1}(\gamma)\right]$ as,

$$
\left[F_{a 1}^{+}(\gamma) \quad F_{b 1}(\gamma)\right]=\left[\begin{array}{cc}
F_{a 11}^{+}(\gamma) & F_{b 11}(\gamma)  \tag{6.1.10}\\
F_{a 12}^{+}(\gamma) & F_{b 12}(\gamma) \\
\vdots & \vdots \\
F_{a 1 m_{d \mathrm{P}}}^{+}(\gamma) & F_{b 1 m_{d \mathrm{P}}}(\gamma)
\end{array}\right]
$$

where $F_{a 1 i}^{+}(\gamma)$ and $F_{b 1 i}(\gamma)$ are of dimensions $1 \times n_{a \mathrm{P}}^{+}$and $1 \times n_{b \mathrm{P}}$, respectively.

Step 6.F.3: Let $\Delta_{c \mathrm{P}}$ be any arbitrary $m_{c \mathrm{P}} \times n_{c \mathrm{P}}$ matrix subject to the constraint that

$$
\begin{equation*}
A_{c c \mathrm{P}}^{c}=A_{c c \mathrm{P}}-B_{c \mathrm{P}} \Delta_{c \mathrm{P}}, \tag{6.1.11}
\end{equation*}
$$

is a stable matrix. Note that the existence of such a $\Delta_{c P}$ is guaranteed by the property that ( $A_{c c \mathrm{P}}, B_{c \mathrm{P}}$ ) is controllable.

Step 6.F.4: This step makes use of subsystems, $i=1$ to $m_{d \mathbf{p}}$, represented by (2.3.14) of Chapter 2. Let $\Lambda_{i}=\left\{\lambda_{i 1}, \lambda_{i 2}, \cdots, \lambda_{i q_{i}}\right\}, i=1$ to $m_{d \mathrm{p}}$, be the sets of $\boldsymbol{q}_{i}$ elements all in $\mathbf{C}^{-}$, which are closed under complex conjugation, where $q_{i}$ and $m_{d \mathrm{P}}$ are as defined in Theorem 2.3.1 but associated with the special coordinate basis of $\Sigma_{\mathbf{p}}$. Let $\Lambda_{d \mathrm{p}}:=\Lambda_{1} \cup \Lambda_{2} \cup \cdots \cup \Lambda_{m_{d \mathrm{P}}}$. For $i=1$ to $m_{d \mathrm{P}}$, we define

$$
\begin{equation*}
p_{i}(s):=\prod_{j=1}^{q_{i}}\left(s-\lambda_{i j}\right)=s^{q_{i}}+F_{i 1} s^{q_{i}-1}+\cdots+F_{i q_{i}-1} s+F_{i q_{i}} \tag{6.1.12}
\end{equation*}
$$

and

$$
\begin{equation*}
\tilde{F}_{i}\left(\varepsilon, \Lambda_{i}\right):=\frac{1}{\varepsilon^{q_{i}}}\left[F_{i q_{i}}, \varepsilon F_{i q_{i}-1}, \cdots, \varepsilon^{q_{i}-1} F_{i 1}\right] . \tag{6.1.13}
\end{equation*}
$$

Step 6.F.5: In this step, various gains calculated in Steps 6.F. 2 to 6.F. 4 are put together to form a composite state feedback gain for the given system $\Sigma_{\mathrm{p}}$. Let

$$
\tilde{F}_{a 1}^{+}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}\right):=\left[\begin{array}{c}
F_{a 11}^{+}(\gamma) F_{1 q_{1}} / \varepsilon^{q_{1}} \\
F_{a 12}^{+}(\gamma) F_{2 q_{2}} / \varepsilon^{q_{2}} \\
\vdots \\
F_{a 1 m_{d \mathrm{P}}}^{+}(\gamma) F_{m_{d \mathrm{P} q_{m_{d \mathrm{P}}}} / \varepsilon^{q_{m_{d \mathrm{P}}}}}
\end{array}\right]
$$

and

$$
\tilde{F}_{b 1}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}\right):=\left[\begin{array}{c}
F_{b 11}(\gamma) F_{1 q_{1}} / \varepsilon^{q_{1}} \\
F_{b 12}(\gamma) F_{2 q_{2}} / \varepsilon^{q_{2}} \\
\vdots \\
F_{b 1 m_{d \mathrm{P}}}(\gamma) F_{m_{d \mathrm{P}} q_{m_{d \mathrm{P}}}} / \varepsilon^{q_{m_{d \mathrm{P}}}}
\end{array}\right]
$$

Then define the state feedback gain $F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right)$ as

$$
\begin{equation*}
F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right)=-\Gamma_{i \mathrm{P}}\left(\tilde{F}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right)+\tilde{F}_{0}\right) \Gamma_{s \mathrm{P}}^{-1} \tag{6.1.14}
\end{equation*}
$$

where $\tilde{F}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right)$ is given by

$$
\begin{gather*}
{\left[\begin{array}{ccccc}
F_{a 0}^{+}(\gamma) & F_{b 0}(\gamma) & 0 & 0 & 0 \\
\tilde{F}_{a 1}^{+}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}\right) & \tilde{F}_{b 1}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}\right) & 0 & 0 & \tilde{F}_{d}\left(\varepsilon, \Lambda_{d \mathrm{P}}\right) \\
0 & 0 & 0 & \Delta_{c \mathrm{P}} & 0
\end{array}\right],}  \tag{6.1.15}\\
\tilde{F}_{0}=\left[\begin{array}{ccccc}
C_{0 a \mathrm{P}}^{+} & C_{0 b \mathrm{P}} & C_{0 a \mathrm{P}}^{-} & C_{0 c \mathrm{P}} & C_{0 d \mathrm{P}} \\
E_{d a \mathrm{P}}^{+} & E_{d b \mathrm{P}} & E_{d a \mathrm{P}}^{-} & E_{d c \mathrm{P}} & E_{d \mathrm{P}} \\
E_{c a \mathrm{P}}^{+} & E_{c b \mathrm{P}} & E_{c a \mathrm{P}}^{-} & 0 & 0
\end{array}\right] \tag{6.1.16}
\end{gather*}
$$

and where

$$
E_{d \mathrm{P}}=\left[\begin{array}{ccc}
E_{11} & \cdots & E_{1 m_{d \mathrm{P}}}  \tag{6.1.17}\\
\vdots & \ddots & \vdots \\
E_{m_{d \mathrm{P} 1}} & \cdots & E_{m_{d \mathrm{P}} m_{d \mathrm{P}}}
\end{array}\right]
$$

and

$$
\begin{equation*}
\tilde{F}_{d}\left(\varepsilon, \Lambda_{d \mathrm{P}}\right)=\operatorname{diag}\left[\tilde{F}_{1}\left(\varepsilon, \Lambda_{1}\right), \tilde{F}_{2}\left(\varepsilon, \Lambda_{2}\right), \cdots, \tilde{F}_{m_{d \mathrm{P}}}\left(\varepsilon, \Lambda_{\left.m_{d \mathrm{P}}\right)}\right)\right] \tag{6.1.18}
\end{equation*}
$$

This completes the algorithm.
We have the following theorem.
Theorem 6.1.1. Consider the full state feedback system (6.1.1) which satisfies Assumptions 6.F.1 to 6.F.4. Then with state feedback gain given by (6.1.14), we have the following properties:

1. For any $\gamma>\gamma^{*}$, for any $\Lambda_{d \mathrm{P}} \subset \mathbb{C}^{-}$which is closed under complex conjugation and for any $\Delta_{c P}$ subject to the constraints that $A_{c c P}^{c}$ is stable, there exists an $\varepsilon^{*}>0$ such that for all $0<\varepsilon \leq \varepsilon^{*}$, the state feedback control law,

$$
\begin{equation*}
u=F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right) x \tag{6.1.19}
\end{equation*}
$$

with $F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right)$ being given as in (6.1.14) is a $\gamma$-suboptimal control law for the given system (6.1.1). Namely, the closed-loop system comprising $\Sigma_{\mathrm{P}}$ and the state feedback law (6.1.19) is internally stable and the $H_{\infty}$-norm of the closed-loop transfer function from the disturbance $w$ to the controlled output $h$ is less than $\gamma$, i.e., $\left\|T_{h w}\right\|_{\infty}<\gamma$.
2. Moreover as $\varepsilon \rightarrow 0$, the poles of the closed-loop system, i.e., the eigenvalues of $A+B F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right)$, are given by

$$
\lambda\left(A_{a a \mathrm{P}}^{-}\right), \quad \lambda\left(A_{c c \mathrm{P}}^{c}\right), \quad \lambda\left(A_{11 \mathrm{P}}^{c}\right)+0(\varepsilon) \quad \text { and } \quad \frac{\Lambda_{d \mathrm{P}}}{\varepsilon}+0(1)
$$

Clearly, there are at least $n_{d \mathrm{p}}$ poles of the closed-loop system have infinite negative real parts as $\varepsilon \rightarrow 0$.

Proof. See Subsection 6.4.A.
因

The following remarks are in order.
Remark 6.1.1. (Interpretations of $\varepsilon, \Lambda_{d \mathrm{P}}$ and $\Delta_{c \mathrm{P}}$ ). Theorem 6.1.1 shows that the closed-loop system under $H_{\infty}$ suboptimal state feedback laws, i.e., $T_{h w}$, has fast eigenvalues $\Lambda_{d \mathrm{P}} / \varepsilon$. So the set of parameters $\Lambda_{d \mathrm{P}}$ in the $H_{\infty}$
suboptimal gain $F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{\mathrm{cP}}\right)$ of (6.1.14) represents the asymptotes of these fast eigenvalues while $\varepsilon$ represents their time-scale. The closed-loop system also has $\lambda\left(A_{\text {ccr }}^{c}\right)$ as slow eigenvalues. These eigenvalues can be assigned to any desired locations in $\mathbf{C}^{-}$by choosing an appropriate $\Delta_{c r}$. Hence, the set of parameters $\Delta_{c p}$ in the $H_{\infty}$ suboptimal state feedback gain prescribes the locations of these slow eigenvalues.

Remark 6.1.2. (Regular Case). If $D_{2}$ is injective, it is obvious from our algorithm that $F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{r}}\right)=F(\gamma)$ does not depend on $\varepsilon, \Lambda_{d \mathrm{p}}$ and $\Delta_{\mathrm{cr}}$, and is given by

$$
F(\gamma)=-\Gamma_{i \mathrm{P}}\left[C_{0 a \mathrm{P}}^{+}+F_{a 0}^{+}(\gamma) \quad C_{0 b \mathrm{P}}+F_{b 0}(\gamma) \quad C_{0 a \mathrm{P}}^{-}\right] \Gamma_{s \mathrm{P}}^{-1}
$$

This corresponds to the regular case, and is the central controller given in Doyle et al [39]. Moreover, if $\gamma=\infty$, the result reduces to the solution of the well-known LQG problem.

回
Remark 6.1.3. Finally, we would like to note that if Assumption 6.F.3, i.e., the geometric condition, is not satisfied, one can use the iterative procedure in Chapter 5 to find an approximation of the infimum, say $\tilde{\gamma}^{*}$. Moreover, the algorithm for finding the $\gamma$-suboptimal state feedback laws can be slightly modified to handle this situation. To be more specific, one only needs to modify Step 6.F. 2 slightly as follows:

Step 6.F.2m: For any $\gamma>\tilde{\gamma}^{*}$, we define

$$
F_{11}(\gamma):=\left[\begin{array}{cc}
F_{a 0}^{+}(\gamma) & F_{b 0}(\gamma) \\
F_{a 1}^{+}(\gamma) & F_{b 1}(\gamma)
\end{array}\right]=\left[\begin{array}{c}
B_{11 \mathrm{p}}^{\prime} P_{x} \\
\left(C_{23 \mathrm{P}}^{\prime} C_{23 \mathrm{P}}\right)^{-1}\left[A_{13 \mathrm{p}}^{\prime} P_{x}+C_{23 \mathrm{P}}^{\prime} C_{21 \mathrm{P}}\right]
\end{array}\right],
$$

where $P_{x}$ is the positive definite solution of the Riccati equation,

$$
P_{x} A_{x \mathrm{P}}+A_{x \mathrm{P}}^{\prime} P_{x}+P_{x} E_{x \mathrm{P}} E_{x \mathrm{P}}^{\prime} P_{x} / \gamma^{2}-P_{x} B_{x \mathrm{P}} B_{x \mathrm{P}}^{\prime} P_{x}+C_{x \mathrm{P}}^{\prime} C_{x \mathrm{P}}=0,
$$

and define

$$
A_{11 \mathrm{P}}^{c}:=A_{11 \mathrm{P}}-\left[\begin{array}{ll}
B_{11 \mathrm{P}} & A_{13 \mathrm{P}}
\end{array}\right] F_{11}(\gamma) .
$$

Let's partition $\left[F_{a 1}^{+}(\gamma) F_{b 1}(\gamma)\right]$ as,

$$
\left[F_{a 1}^{+}(\gamma) \quad F_{b 1}(\gamma)\right]=\left[\begin{array}{cc}
F_{a 11}^{+}(\gamma) & F_{b 11}(\gamma) \\
F_{a 12}^{+}(\gamma) & F_{b 12}(\gamma) \\
\vdots & \vdots \\
F_{a 1 m_{d \mathrm{P}}}^{+}(\gamma) & F_{b 1 m_{d \mathrm{P}}}(\gamma)
\end{array}\right]
$$

where $F_{a 1 i}^{+}(\gamma)$ and $F_{b 1 i}(\gamma)$ are of dimensions $1 \times n_{a \mathrm{P}}^{+}$and $1 \times n_{b \mathrm{P}}$.

The rest steps of the algorithm, i.e., Steps 6.F.1, 6.F. 3 to 6.F.5, remain unchanged. All results in Theorem 6.1.1 are valid for this situation as well. The only difference is that the control law is no longer of closed-form.

The following theorem deals with pole-zero cancellations in the closed-loop system $T_{h w}$ under the state feedback law $u=F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right) x$.

Theorem 6.1.2. (Pole-zero Cancellations). $\lambda\left(A_{a a \mathrm{P}}^{-}\right)$, the stable invariant zeros of the system $\Sigma_{\mathrm{P}}$, and $\lambda\left(A_{c \mathrm{cP}}^{c}\right)$ are the output decoupling zeros of the closed-loop transfer matrix $T_{h w}$. Hence, they cancel with the poles of $T_{h w}$. T

Proof. See Subsection 6.4.B.
因

We illustrate our algorithm in the following example.

Example 6.1.1. Re-consider the system in Example 5.1.1, i.e., a full state feedback system characterized by

$$
A=\left[\begin{array}{lllll}
1 & 1 & 1 & 0 & 1  \tag{6.1.20}\\
0 & 1 & 0 & 0 & 1 \\
0 & 1 & 1 & 0 & 1 \\
1 & 1 & 1 & 1 & 1 \\
1 & 1 & 1 & 1 & 0
\end{array}\right], \quad B=\left[\begin{array}{lll}
0 & 0 & 0 \\
0 & 0 & 0 \\
1 & 0 & 0 \\
0 & 0 & 1 \\
0 & 1 & 0
\end{array}\right], \quad E=\left[\begin{array}{ll}
5 & 1 \\
0 & 0 \\
0 & 0 \\
2 & 3 \\
1 & 4
\end{array}\right],
$$

and

$$
C_{2}=\left[\begin{array}{lllll}
0 & 0 & 1 & 0 & 0  \tag{6.1.21}\\
0 & 0 & 0 & 0 & 1 \\
0 & 1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0
\end{array}\right], \quad D_{2}=\left[\begin{array}{lll}
1 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right], \quad D_{22}=\left[\begin{array}{ll}
0 & 0 \\
0 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right]
$$

It is easy to verify that $(A, B)$ is stabilizable, and the system $\Sigma_{\mathbf{P}}$ is neither right nor left invertible and is of nonminimum phase with an invariant zero at $s=1$. Moreover, it is already in the form of the special coordinate basis with $n_{a \mathrm{P}}^{+}=1, n_{a \mathrm{P}}^{-}=n_{a \mathrm{P}}^{0}=0, n_{b \mathrm{P}}=2$ and $n_{c \mathrm{P}}=n_{d \mathrm{P}}=1$. Also, it is simple to see that $\operatorname{Im}(E) \subseteq \mathcal{V}^{-}\left(\Sigma_{\mathrm{P}}\right) \cup \mathcal{S}^{-}\left(\Sigma_{\mathrm{P}}\right)$ since $E_{b \mathrm{P}}=0$. Hence, all Assumptions 6.F. 1 to 6.F. 4 are satisfied. Moreover, it was obtained in Example 5.1.1 that the infimum is given by

$$
\gamma^{*}=6.4679044
$$



Figure 6.1.1: Maximum singular values of $T_{h w}$ (state feedback case).

Following the algorithm in this section, we obtain the closed-form solution of the $\gamma$-suboptimal state feedback gains, $F\left(\gamma, \varepsilon, \lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right)$, which is given by

$$
\left[\begin{array}{ccc}
\frac{-0.163673 \gamma^{2}}{0.132909 \gamma^{2}-5.560084} & -1+\frac{0.294790 \gamma^{2} \lambda_{d \mathrm{P}}}{\left(0.132909 \gamma^{2}-5.560084\right) \varepsilon} & -1  \tag{6.1.22}\\
\frac{0.185427 \gamma^{2}-3.009097}{0.132909 \gamma^{2}-5.560084} & -1+\frac{\left(0.102145 \gamma^{2}-12.824695\right) \lambda_{d \mathrm{p}}}{\left(0.132909 \gamma^{2}-5.560084\right) \varepsilon} & -1 \\
\frac{-0.318336 \gamma^{2}+10.696930}{0.132909 \gamma^{2}-5.560084} & -1+\frac{\left(0.163673 \gamma^{2}-2.127749\right) \lambda_{d \mathrm{P}}}{\left(0.132909 \gamma^{2}-5.560084\right) \varepsilon} & -1 \\
0 & -1 & -\Delta_{\mathrm{cP}} \\
0 & \frac{\lambda_{d \mathrm{P}}}{\varepsilon} & 0
\end{array}\right]^{\prime}
$$

where the scalars $\lambda_{d \mathrm{P}}<0$ and $\Delta_{c \mathrm{P}}>1$ (note that $\Delta_{c \mathrm{P}}$ must be greater than one in order to have stable $A_{c c \mathrm{P}}^{c}$ ). We demonstrate our results in Figure 6.1.1 by the plots of maximum singular values of the closed-loop transfer function matrix for several values of $\gamma$ and $\varepsilon$. Note that in Figure 6.1.1, we choose parameters $\lambda_{d \mathrm{P}}=-1$ and $\Delta_{c \mathrm{P}}=3$.

### 6.2. Full Order Output Feedback

This section deals with $H_{\infty}$ suboptimal and optimal design using full order measurement output feedback laws, i.e., the dynamical order of these control laws will be exactly the same as that of the given system. To be more specific, we consider the following measurement feedback system

$$
\left\{\begin{array}{l}
\dot{x}=A x+B u+E w  \tag{6.2.1}\\
y=C_{1} x+D_{1} w \\
h=C_{2} x+D_{2} u+D_{22} w
\end{array}\right.
$$

where $x \in \mathbb{R}^{n}$ is the state, $u \in \mathbb{R}^{m}$ is the control input, $w \in \mathbb{R}^{q}$ is the external disturbance input, $y \in \mathbb{R}^{p}$ is the measurement output, and $h \in \mathbb{R}^{\ell}$ is the controlled output of $\Sigma$. Again, we let $\Sigma_{P}$ be the subsystem characterized by the matrix quadruple $\left(A, B, C_{2}, D_{2}\right)$ and $\Sigma_{Q}$ be the subsystem characterized by the matrix quadruple ( $A, E, C_{1}, D_{1}$ ). The following assumptions are made first:

Assumption 6.M.1: $(A, B)$ is stabilizable;
Assumption 6.M.2: $\Sigma_{\mathrm{P}}$ has no invariant zero on the imaginary axis;
Assumption 6.M.3: $\operatorname{Im}(E) \subset \mathcal{V}^{-}\left(\Sigma_{\mathrm{P}}\right)+\mathcal{S}^{-}\left(\Sigma_{\mathrm{P}}\right)$;
Assumption 6.M.4: $\left(A, C_{1}\right)$ is detectable;
Assumption 6.M.5: $\Sigma_{\mathbf{Q}}$ has no invariant zero on the imaginary axis;
Assumption 6.M.6: $\operatorname{Ker}\left(C_{2}\right) \supset \mathcal{V}^{-}\left(\Sigma_{Q}\right) \cap \mathcal{S}^{-}\left(\Sigma_{Q}\right)$; and
Assumption 6.M.7: $D_{22}=0$.
The class of output feedback controllers that we consider in this section are basically observer based control laws and can be regarded as an extension of the central output feedback controller that was proposed in Doyle et al [39] for the regular case. We have modified the central output feedback controller of the regular case to deal with the singular case. This modification will be discussed later on. We assume that the infimum $\gamma^{*}$ has been obtained using methods given in Section 5.2 of Chapter 5. The procedure for obtaining the closed-form of the $H_{\infty}$ suboptimal output feedback laws for any $\gamma>\gamma^{*}$ proceeds as follows.

Step 6.M.1: Define an auxiliary full state feedback system

$$
\left\{\begin{array}{l}
\dot{x}=A x+B u+E w \\
y=x \\
h=C_{2} x+D_{2} u+D_{22} w
\end{array}\right.
$$

and proceed to perform Steps 6.F.1 to 6.F.5 of Section 6.1 to obtain the gain matrix $F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right)$. Also, define

$$
P(\gamma):=\left(\Gamma_{s \mathrm{P}}^{-1}\right)^{\prime}\left[\begin{array}{cc}
\left(S_{x \mathrm{~F}}-\gamma^{-2} T_{x \mathrm{P}}\right)^{-1} & 0  \tag{6.2.2}\\
0 & 0
\end{array}\right] \Gamma_{s \mathrm{~F}}^{-1}
$$

Step 6.M.2: Define another auxiliary full state feedback system as follows,

$$
\Sigma_{Q}:\left\{\begin{array}{l}
\dot{x}=A^{\prime} x+C_{1}^{\prime} u+C_{2}^{\prime} w  \tag{6.2.3}\\
y=x \\
h=E^{\prime} x+D_{1}^{\prime} u+D_{22}^{\prime} w
\end{array}\right.
$$

and proceed to perform Steps 6.F.1 to 6.F.5 of Section 6.1 but for this auxiliary system to obtain a gain $F\left(\gamma, \varepsilon, \Lambda_{d Q}, \Delta_{c Q}\right)$. Let $K\left(\gamma, \varepsilon, \Lambda_{d Q}, \Delta_{c Q}\right):=$ $F\left(\gamma, \varepsilon, \Lambda_{d Q}, \Delta_{c Q}\right)^{\prime}$. Also, define

$$
Q(\gamma):=\left(\Gamma_{s Q}^{-1}\right)^{\prime}\left[\begin{array}{cc}
\left(S_{x Q}-\gamma^{-2} T_{x Q}\right)^{-1} & 0  \tag{6.2.4}\\
0 & 0
\end{array}\right] \Gamma_{s Q}^{-1}
$$

Step 6.M.3: Construct the following full order observer based controller,

$$
\Sigma_{\mathrm{cmp}}:\left\{\begin{array}{l}
\dot{v}=A_{\mathrm{cmp}} v+B_{\mathrm{cmp}} y  \tag{6.2.5}\\
u=C_{\mathrm{cmp}} v+0 y
\end{array}\right.
$$

where

$$
\begin{align*}
A_{\mathrm{cmp}}= & A+\gamma^{-2} E E^{\prime} P(\gamma)+B F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right) \\
& +\left[I-\gamma^{-2} Q(\gamma) P(\gamma)\right]^{-1}\left\{K\left(\gamma, \varepsilon, \Lambda_{d \mathrm{Q}}, \Delta_{c \mathrm{Q}}\right)\left[C_{1}+\gamma^{-2} D_{1} E^{\prime} P(\gamma)\right]\right. \\
& +\gamma^{-2} Q(\gamma)\left[A^{\prime} P(\gamma)+P(\gamma) A+C_{2}^{\prime} C_{2}+\gamma^{-2} P(\gamma) E E^{\prime} P(\gamma)\right] \\
& \left.+\gamma^{-2} Q(\gamma)\left[P(\gamma) B+C_{2}^{\prime} D_{2}\right] F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right)\right\},  \tag{6.2.6}\\
B_{\mathrm{cmp}}= & -\left[I-\gamma^{-2} Q(\gamma) P(\gamma)\right]^{-1} K\left(\gamma, \varepsilon, \Lambda_{d \mathrm{Q}}, \Delta_{c \mathrm{Q}}\right),  \tag{6.2.7}\\
C_{\mathrm{cmp}}= & F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right) . \tag{6.2.8}
\end{align*}
$$

It is to be shown that $\Sigma_{\text {cmp }}$ is indeed a $\gamma$-suboptimal controller. Clearly, it has a dynamical order of $n$, i.e., it is a full order output feedback controller.

We have the following theorem.
Theorem 6.2.1. Consider the given measurement feedback system (6.2.1) satisfying Assumptions 6.M.1 to 6.M.7. Then for any $\gamma>\gamma^{*}$, for any $\Lambda_{d \mathrm{P}} \subset \mathbb{C}^{-}$ and $\Lambda_{d Q} \subset \mathbb{C}^{-}$which are closed under complex conjugation, and for any $\Delta_{c P}$
and $\Delta_{c Q}$ subject to the constraints that $A_{c c P}^{c}$ and $A_{c c Q}^{c}$ are stable matrices, there exists an $\varepsilon^{*}>0$ such that for all $0<\varepsilon \leq \varepsilon^{*}$, the control law $\Sigma_{\text {cmp }}$ as given in (6.2.5) is $\gamma$-suboptimal controller, namely, the closed-loop system comprising $\Sigma$ and the output feedback controller $\Sigma_{\text {cmp }}$, is internally stable and the $H_{\infty}$-norm of the closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$ is less than $\gamma$, i.e., $\left\|T_{h w}\right\|_{\infty}<\gamma$.

Proof. See Subsection 6.4.C.
因
The following theorem deals with the issue of pole-zero cancellations and the closed-loop eigenvalues in the $\gamma$-suboptimal output feedback control.

Theorem 6.2.2. Consider the given measurement feedback system (6.2.1) satisfying Assumptions 6.M.1 to 6.M. 7 with the $\gamma$-suboptimal control $\Sigma_{\mathrm{cmp}}$ as given in (6.2.5). Then the following properties hold:

1. $\lambda\left(A_{a a \mathrm{P}}^{-}\right)$, the stable invariant zeros of the system $\left(A, B, C_{2}, D_{2}\right)$, and $\lambda\left(A_{c c \mathrm{P}}^{c}\right)$ are the output decoupling zeros of the closed-loop system $T_{h w}$. Hence they cancel with the poles of $T_{h w}$.
2. $\lambda\left(A_{a a \mathrm{Q}}^{-}\right)$, the stable invariant zeros of the system $\left(A, E, C_{1}, D_{1}\right)$, and $\lambda\left(A_{c c Q}^{c}\right)$ are the input decoupling zeros of the closed-loop system $T_{h w}$. Hence they cancel with the poles of $T_{h w}$.
3. As $\varepsilon \rightarrow 0$, the fast eigenvalues of the closed-loop system are asymptotically given by $\Lambda_{d \mathrm{P}} / \varepsilon+0(1)$ and $\Lambda_{d \mathrm{Q}} / \varepsilon+0(1)$.

Proof. See Subsection 6.4.D.
The following remarks are in order.
Remark 6.2.1. (Interpretations of $\varepsilon, \Lambda_{d \mathrm{P}}, \Lambda_{d Q}, \Delta_{c \mathrm{P}}$ and $\Delta_{c Q}$ ). Again, as in Remark 6.1.1, the set of parameters $\Lambda_{d \mathrm{P}}$ and $\Lambda_{d Q}$ represent the asymptotes of the fast eigenvalues of the closed-loop system while $\varepsilon$ represents their timescale. The set of parameters $\Delta_{c \mathrm{P}}$ and $\Delta_{c Q}$ prescribe the locations of the slow eigenvalues of the closed-loop system corresponding to $\lambda\left(A_{c c \mathrm{P}}^{c}\right)$ and $\lambda\left(A_{c c Q}^{c}\right)$. The eigenvalues can be assigned to any desired locations in $\mathbb{C}^{-}$by choosing appropriate $\Delta_{C P}$ and $\Delta_{C Q}$.

Remark 6.2.2. (Regular Case). If $D_{1}$ is surjective and $D_{2}$ is injective, it is simple to verify that $F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right)=F(\gamma)$ and $K\left(\gamma, \varepsilon, \Lambda_{d \mathrm{Q}}, \Delta_{c \mathrm{Q}}\right)=K(\gamma)$ depend only on $\gamma$. Moreover, we have

$$
\left[P(\gamma) B+C_{2}^{\prime} D_{2}\right] F(\gamma)+\left[A^{\prime} P(\gamma)+P(\gamma) A+C_{2}^{\prime} C_{2}+\gamma^{-2} P(\gamma) E E^{\prime} P(\gamma)\right]=0
$$

Hence, $\boldsymbol{\Sigma}_{\mathrm{cmp}}$ reduces to

$$
\Sigma_{\mathrm{cmp}}:\left\{\begin{array}{l}
\dot{v}=A_{\mathrm{cmp}} v+B_{\mathrm{cmp}} y \\
u=C_{\mathrm{cmp}} v+0 y
\end{array}\right.
$$

where

$$
\begin{aligned}
A_{\mathrm{cmp}}= & A+\gamma^{-2} E E^{\prime} P(\gamma)+B F(\gamma) \\
& +\left[I-\gamma^{-2} Q(\gamma) P(\gamma)\right]^{-1} K(\gamma)\left[C_{1}+\gamma^{-2} D_{1} E^{\prime} P(\gamma)\right] \\
B_{\mathrm{cmp}}= & -\left[I-\gamma^{-2} Q(\gamma) P(\gamma)\right]^{-1} K(\gamma) \\
C_{\mathrm{cmp}}= & F(\gamma)
\end{aligned}
$$

This corresponds to the regular case, and is the central controller given in Doyle et al [39].

Remark 6.2.3. It is known that for a mixed sensitivity problem (see for example, Kwakernaak [52] and Postlewaite et al [80]),

1. the $H_{\infty}$ design results in pole-zero cancellation between plant and controller at all of the stable poles of the uncompensated plant;
2. the closed-loop poles include the mirror image positions of all unstable poles of the plant.

We would like to point out that none of these behaviors arise in the class of problem that we have considered. It is obvious that the class of mixed sensitivity problem and our class of problem are disjoint since a mixed sensitivity problem always involves a feedthrough term from the disturbance to the controlled output.

Remark 6.2.4. Finally, we would like to note that if Assumptions 6.M. 3 and 6.M.6, i.e., the geometric conditions, are not satisfied, one can use the iterative procedure in Chapter 5 to find an approximation of the infimum, say $\tilde{\gamma}^{*}$. Moreover, the algorithm for finding the $\gamma$-suboptimal output feedback laws can also be modified to handle this situation. To be more specific, one only needs to modify Steps 6.M.1 and 7.M. 2 slightly as follows:

Step 6.M.1m: Define an auxiliary full state feedback system

$$
\left\{\begin{array}{l}
\dot{x}=A x+B u+E w \\
y=x \\
h=C_{2} x+D_{2} u+D_{22} w
\end{array}\right.
$$

and proceed to perform Steps 6.F.1, 6.F.2m, and 6.F. 3 to $6 . F .5$ of Section 6.1 to obtain the gain matrix $F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{CP}}\right)$ and $P_{x}$. Let $P_{x \mathrm{P}}:=P_{x}$. Also, define

$$
P(\gamma):=\left(\Gamma_{s \mathrm{P}}^{-1}\right)^{\prime}\left[\begin{array}{cc}
P_{x \mathrm{P}} & 0  \tag{6.2.9}\\
0 & 0
\end{array}\right] \Gamma_{s \mathrm{P}}^{-1}
$$

Step 6.M.2m: Define another auxiliary full state feedback system as follows,

$$
\Sigma_{\mathbf{Q}}:\left\{\begin{array}{l}
\dot{x}=A^{\prime} x+C_{1}^{\prime} u+C_{2}^{\prime} w \\
y=x \\
h=E^{\prime} x+D_{1}^{\prime} u+D_{22}^{\prime} w
\end{array}\right.
$$

and proceed to perform Steps 6.F.1, 6.F.2m, and 6.F. 3 to 6.F. 5 of Section 6.1 but for this auxiliary system to obtain $F\left(\gamma, \varepsilon, \Lambda_{d P}, \Delta_{c \mathrm{P}}\right)$ and $P_{x}$. Let $K\left(\gamma, \varepsilon, \Lambda_{d Q}, \Delta_{c Q}\right):=F\left(\gamma, \varepsilon, \Lambda_{d Q}, \Delta_{c Q}\right)^{\prime}$ and $Q_{x Q}:=P_{x}$. Also, define

$$
Q(\gamma):=\left(\Gamma_{s Q}^{-1}\right)^{\prime}\left[\begin{array}{cc}
Q_{x Q} & 0  \tag{6.2.10}\\
0 & 0
\end{array}\right] \Gamma_{s Q}^{-1}
$$

The last step of the algorithm, i.e., Step 6.M.3, remains unchanged. All results in Theorems 6.2.1 and 6.2.2 are valid for this situation as well. However, the output feedback control law is not of closed-form any more.

园

Again, we illustrate our results in the following example.
Example 6.2.1. Consider a given measurement feedback system characterized by matrices $A, B, E, C_{2}, D_{2}$ and $D_{22}$ as given in Example 6.1.1 of the previous section and

$$
C_{1}=\left[\begin{array}{rrrrr}
0 & -2 & -3 & -2 & -1  \tag{6.2.11}\\
1 & 2 & 3 & 2 & 1
\end{array}\right], \quad D_{1}=\left[\begin{array}{ll}
1 & 0 \\
0 & 0
\end{array}\right]
$$

We first note that the pair $\left(A, C_{1}\right)$ is detectable, and the system $\left(A, E, C_{1}, D_{1}\right)$ is invertible (hence Assumption 6.M.6 is satisfied) and of nonminimum phase with invariant zeros at $\{-1.630662,-3.593415,0.521129 \pm j 0.363042\}$. It was obtained in Example 5.2.1 that

$$
\gamma^{*}=13.638725
$$

The closed-form to the output feedback suboptimal controllers as in (6.2.5) to (6.2.8) with $F\left(\gamma, \varepsilon, \lambda_{d \mathrm{P}}, \Delta_{\mathrm{cP}}\right)$ given by (6.1.22),

$$
K\left(\gamma, \varepsilon, \lambda_{d Q}, \Delta_{c Q}\right)=\left[\begin{array}{ll}
K_{0} & K_{1} \tag{6.2.12}
\end{array}\right]
$$

where

$$
K_{0}=\left[\begin{array}{c}
\frac{-43.91 \gamma^{4}+4257.86 \gamma^{2}-97026.13}{7.12 \gamma^{4}-790.42 \gamma^{2}+19405.23} \\
\frac{-12.45 \gamma^{4}+372.65 \gamma^{2}-0.02}{7.12 \gamma^{4}-790.42 \gamma^{2}+19405.23} \\
\frac{-48.44 \gamma^{4}+1803.08 \gamma^{2}+0.02}{7.12 \gamma^{4}-790.42 \gamma^{2}+19405.23} \\
\frac{62.57 \gamma^{4}-1212.58 \gamma^{2}-38810.46}{7.12 \gamma^{4}-790.42 \gamma^{2}+19405.23} \\
\frac{17.80 \gamma^{4}-83.04 \gamma^{2}-19405.21}{7.12 \gamma^{4}-790.42 \gamma^{2}+19405.23}
\end{array}\right],
$$

and

$$
K_{1}=\left[\begin{array}{c}
-5+0.090909 \frac{\lambda_{d Q}}{\varepsilon}-\frac{\left(0.24 \gamma^{4}-10.14 \gamma^{2}\right) \lambda_{d Q}}{\left(7.12 \gamma^{4}-790.42 \gamma^{2}+19405.23\right) \varepsilon} \\
-0.363636-\frac{\left(-2.39 \gamma^{4}+190.91 \gamma^{2}\right) \lambda_{d Q}}{\left(7.12 \gamma^{4}-790.42 \gamma^{2}+19405.23\right) \varepsilon} \\
-0.382726-\frac{\left(2.04 \gamma^{4}-108.95 \gamma^{2}\right) \lambda_{d Q}}{\left(7.12 \gamma^{4}-790.42 \gamma^{2}+19405.23\right) \varepsilon} \\
-2.545451+0.272727 \frac{\lambda_{d Q}}{\varepsilon}-\frac{\left(-1.13 \gamma^{4}+14.86 \gamma^{2}\right) \lambda_{d Q}}{\left(7.12 \gamma^{4}-790.42 \gamma^{2}+19405.23\right) \varepsilon} \\
-1.272726+0.363636 \frac{\lambda_{d Q}}{\varepsilon}-\frac{\left(0.69 \gamma^{4}-74.56 \gamma^{2}\right) \lambda_{d Q}}{\left(7.12 \gamma^{4}-790.42 \gamma^{2}+19405.23\right) \varepsilon}
\end{array}\right]
$$

with $\lambda_{d Q}<0$, and

$$
\begin{aligned}
& P(\gamma)= \frac{1}{0.132909 \gamma^{2}-5.560084} \times \\
& {\left[\begin{array}{rrrrr}
0.42770 \gamma^{2} & -0.29658 \gamma^{2} & 0.16367 \gamma^{2} & 0 & 0 \\
-0.29658 \gamma^{2} & -15.8338+0.58415 \gamma^{2} & 3.0091-0.18543 \gamma^{2} & 0 & 0 \\
0.16367 \gamma^{2} & 3.0091-0.18543 \gamma^{2} & -5.1368+0.18543 \gamma^{2} & 0 & 0 \\
0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{array}\right], } \\
& Q(\gamma)=\frac{\gamma^{2}}{0.071193 \gamma^{4}-7.904171 \gamma^{2}+194.052288}\left(\gamma^{2} Q_{1}+Q_{0}\right),
\end{aligned}
$$

where

$$
Q_{1}=\left[\begin{array}{rrrrr}
0.083104 & 0.124442 & 0.484459 & -0.768087 & -0.249208 \\
0.124423 & 1.778706 & 0.340500 & -1.759522 & -1.184163 \\
0.484459 & 0.340500 & 2.917279 & -4.330299 & -1.256601 \\
-0.768087 & -1.759522 & -4.330299 & 7.332315 & 2.613520 \\
-0.249208 & -1.184163 & -1.256601 & 2.613520 & 1.160281
\end{array}\right]
$$



Figure 6.2.1: Maximum singular values of $T_{h w}$ (output feedback case).
and
$Q_{0}=\left[\begin{array}{rrrrr}-3.0576430 & -3.7265760 & -18.030782 & 27.934279 & 8.7345960 \\ -3.7265760 & -122.50790 & 6.5460280 & 79.188507 & 70.727376 \\ -18.030781 & 6.5460280 & -113.22255 & 153.81266 & 36.981101 \\ 27.934279 & 79.188509 & 153.81266 & -272.47959 & -102.79025 \\ 8.7345960 & 70.727376 & 36.981101 & -102.79025 & -55.552230\end{array}\right]$.
As in the previous example, we demonstrate our results in Figure 6.2 .1 by the plots of maximum singular values of the closed-loop transfer function matrix for several values of $\gamma$ and $\varepsilon$. Note that in Figure 6.2.1, we choose $\lambda_{d \mathrm{P}}=-1$, $\Delta_{c \mathbf{P}}=3$ and $\lambda_{d \mathbf{Q}}=-1$. Note that since $\Sigma_{\mathbf{Q}}$ for this example is left invertible, the gain $K\left(\gamma, \varepsilon, \lambda_{d Q}, \Delta_{c Q}\right)$ depends only on $\gamma, \varepsilon$ and $\lambda_{d Q}$.

### 6.3. Reduced Order Output Feedback

In this section, the $H_{\infty}$ control problem with reduced order measurement output feedback is investigated. For the case that some entries of the measurement vector are not noise-corrupted, we show that one can find dynamic compensators
of a lower dynamical order. More specifically, we will show that there exists a time-invariant, finite-dimensional dynamic compensator $\Sigma_{\mathbf{c m p}}$ of the form

$$
\Sigma_{\mathrm{cmp}}:\left\{\begin{array}{l}
\dot{v}=A_{\mathrm{cmp}} v+B_{\mathrm{cmp}} y  \tag{6.3.1}\\
u=C_{\mathrm{cmp}} v+D_{\mathrm{cmp}} y
\end{array}\right.
$$

and with a McMillan degree $n-\operatorname{rank}\left[C_{1}, D_{1}\right]+\operatorname{rank}\left(D_{1}\right) \leq n$ for $\Sigma$ of (6.2.1) such that the resulting closed loop system is internally stable and the closed loop transfer function from $w$ to $h$ has an $H_{\infty}$ norm less than $\gamma>\gamma^{*}$. Moreover, we give an explicit construction of such a reduced order compensator. The result of this section was previously reported in [103] while the original idea for how to construct a reduced order observer for a general system was given by Chen et al [22].

Let $\gamma^{*}$ be the infimum for the given system $\Sigma$ of (6.2.1) and let $\gamma>\gamma^{*}$ be given. Using the result of the previous section, one can easily find two positive semi-definite matrices $P$ and $Q$ which satisfy

$$
F_{\gamma}(P):=\left[\begin{array}{cc}
A^{\prime} P+P A+C_{2}^{\prime} C_{2}+P E E^{\prime} P / \gamma^{2} & P B+C_{2}^{\prime} D_{2} \\
B^{\prime} P+D_{2}^{\prime} C_{2} & D_{2}^{\prime} D_{2}
\end{array}\right] \geq 0
$$

and

$$
G_{\gamma}(Q):=\left[\begin{array}{cc}
A Q+Q A^{\prime}+E E^{\prime}+Q C_{2}^{\prime} C_{2} Q / \gamma^{2} & Q C_{1}^{\prime}+E D_{1}^{\prime} \\
C_{1} Q+D_{1} E^{\prime} & D_{1} D_{1}^{\prime}
\end{array}\right] \geq 0
$$

respectively, i.e., $P$ and $Q$ are the solutions of the quadratic matrix inequalities $F_{\gamma}(P) \geq 0$ and $G_{\gamma}(Q) \geq 0$. Next, we define an auxiliary system,

$$
\Sigma_{\mathrm{PQ}}:\left\{\begin{align*}
& \dot{x}_{\mathrm{PQ}}=A_{\mathrm{PQ}} x_{\mathrm{PQ}}+B_{\mathrm{PQ}} u+E_{\mathrm{PQ}} w_{\mathrm{PQ}}  \tag{6.3.2}\\
& y=C_{1 \mathrm{P}} x_{\mathrm{PQ}} \\
& h_{\mathrm{PQ}}=C_{2 \mathrm{P}} x_{\mathrm{PQ}}+D_{2 \mathrm{P}} u,
\end{align*}\right.
$$

where

$$
\left[\begin{array}{l}
C_{2 \mathrm{P}}^{\prime} \\
D_{2 \mathrm{P}}^{\prime}
\end{array}\right]\left[\begin{array}{ll}
C_{2 \mathrm{P}} & D_{2 \mathrm{P}}
\end{array}\right]:=F_{\gamma}(P), \quad\left[\begin{array}{c}
E_{\mathrm{Q}} \\
D_{1 \mathrm{PQ}}
\end{array}\right]\left[\begin{array}{ll}
E_{\mathrm{Q}}^{\prime} & D_{1 \mathrm{PQ}}^{\prime}
\end{array}\right]:=G_{\gamma}(Q)
$$

and

$$
\left.\begin{array}{l}
A_{\mathrm{PQ}}:=A+E E^{\prime} P / \gamma^{2}+\left(\gamma^{2} I-Q P\right)^{-1} Q C_{2 \mathrm{P}}^{\prime} C_{2 \mathrm{P}},  \tag{6.3.3}\\
B_{\mathrm{PQ}}:=B+\left(\gamma^{2} I-Q P\right)^{-1} Q C_{2 \mathrm{P}}^{\prime} D_{2 \mathrm{P}}, \\
E_{\mathrm{PQ}}:=\left(I-Q P / \gamma^{2}\right)^{-1} E_{\mathrm{Q}}, \\
C_{1 \mathrm{P}}:=C_{1}+D_{1} E^{\prime} P / \gamma^{2} .
\end{array}\right\}
$$

It can be shown (see e.g., [100]) that i) $\left(A_{\mathrm{PQ}}, B_{\mathrm{PQ}}, C_{2 \mathrm{P}}, D_{2 \mathrm{P}}\right)$ is right invertible and of minimum phase; and ii) $\left(A_{\mathrm{PQ}}, E_{\mathrm{PQ}}, C_{1 \mathrm{P}}, D_{1 \mathrm{PQ}}\right)$ is left invertible and of minimum phase.

We will build the reduced order compensator upon the above auxiliary system and show later that it works for the original system $\Sigma$ of (6.2.1) as well. Let us first eliminate states which can be directly observed and concentrate on those states which still need to be observed. In order to do this, we need to choose a suitable basis. Without loss of generality, but for simplicity of presentation, we assume that the matrices $C_{1 \mathrm{P}}$ and $D_{2 \mathrm{PQ}}$ are transformed in the following form:

$$
C_{1 \mathrm{P}}=\left[\begin{array}{cc}
0 & C_{1,02}  \tag{6.3.4}\\
I_{k} & 0
\end{array}\right] \quad \text { and } \quad D_{2 \mathrm{PQ}}=\left[\begin{array}{c}
D_{1,0} \\
0
\end{array}\right]
$$

Thus, the system $\Sigma_{\text {PQ }}$ as in (6.3.2) can be partitioned as follows,

$$
\left\{\begin{array}{l}
\binom{\dot{x}_{1}}{\dot{x}_{2}}=\left[\begin{array}{ll}
A_{11} & A_{12} \\
A_{21} & A_{22}
\end{array}\right]\binom{x_{1}}{x_{2}}+\left[\begin{array}{l}
B_{1} \\
B_{2}
\end{array}\right] u+\left[\begin{array}{l}
E_{1} \\
E_{2}
\end{array}\right] w_{\mathrm{PQ}}  \tag{6.3.5}\\
\binom{y_{0}}{y_{1}}=\left[\begin{array}{cc}
0 & C_{1,02} \\
I_{k} & 0
\end{array}\right]\binom{x_{1}}{x_{2}} \\
h_{1,0} \\
h_{\mathrm{PQ}}=C_{2 \mathrm{P}}
\end{array} x_{\mathrm{PQ}}, ~ x_{\mathrm{PQ}}+D_{2 \mathrm{P}} u, ~ \$\right.
$$

where $\left(x_{1}^{\prime}, x_{2}^{\prime}\right)^{\prime}=x_{\mathrm{PQ}}$ and $\left(y_{0}^{\prime}, y_{1}^{\prime}\right)^{\prime}=y$. We observe that $y_{1}=x_{1}$ is already available and need not be estimated. Thus we need to estimate only the state variable $x_{2}$. We first rewrite the state equation for $x_{1}$ in terms of the output $y_{1}$ and state $x_{2}$ as follows,

$$
\begin{equation*}
\dot{y}_{1}=A_{11} y_{1}+A_{12} x_{2}+B_{1} u+E_{1} w_{\mathrm{PQ}} \tag{6.3.6}
\end{equation*}
$$

where $y_{1}$ and $u$ are known signals. Equation (6.3.6) can be rewritten as

$$
\begin{equation*}
\tilde{y}=A_{12} x_{2}+E_{1} w_{\mathrm{PQ}}=\dot{y}_{1}-A_{11} x_{1}-B_{1} u . \tag{6.3.7}
\end{equation*}
$$

Thus, observation of $x_{2}$ is made via (6.3.7) as well as by

$$
y_{0}=C_{1,02} x_{2}+D_{1,0} w_{\mathrm{PQ}}
$$

Now, a reduced order system suitable for estimating the state $x_{2}$ is given by

$$
\left\{\begin{align*}
& \dot{x}_{2}=A_{22}  \tag{6.3.8}\\
& x_{2}+\left[\begin{array}{ll}
A_{21} & B_{2}
\end{array}\right]\binom{y_{1}}{u}+\begin{array}{cc}
E_{2} & w_{\mathrm{PQ}} \\
\binom{y_{0}}{\tilde{y}} & =\left[\begin{array}{c}
C_{1,02} \\
A_{12}
\end{array}\right] x_{2}
\end{array}+\left[\begin{array}{c}
D_{1,0} \\
E_{1}
\end{array}\right] w_{\mathrm{PQ}}
\end{align*}\right.
$$

Before we proceed to construct the reduced order observer, we present in the following a key lemma which plays an important role in our design.

Lemma 6.3.1. Let $\Sigma_{\mathrm{R}}$ denote the subsystem characterized by

$$
\left(A_{\mathrm{R}}, B_{\mathrm{R}}, C_{\mathrm{R}}, D_{\mathrm{R}}\right):=\left(A_{22}, E_{2},\left[\begin{array}{c}
C_{1,02} \\
A_{12}
\end{array}\right],\left[\begin{array}{c}
D_{1,0} \\
E_{1}
\end{array}\right]\right)
$$

Then we have

1. $\Sigma_{\mathrm{R}}$ is (non-)minimum phase if and only if $\left(A_{\mathrm{PQ}}, E_{\mathrm{PQ}}, C_{1_{\mathrm{P}}}, D_{1_{\mathrm{PQ}}}\right)$ is (non-) minimum phase.
2. $\Sigma_{\mathrm{R}}$ is detectable if and only if $\left(A_{\mathrm{PQ}}, E_{\mathrm{PQ}}, C_{1 \mathrm{P}}, D_{1 \mathrm{PQ}}\right)$ is detectable.
3. $\Sigma_{\mathrm{R}}$ is left invertible if and only if $\left(A_{\mathrm{PQ}}, E_{\mathrm{PQ}}, C_{1 \mathrm{P}}, D_{1 \mathrm{PQ}}\right)$ is left invertible.
4. Invariant zeros of $\Sigma_{\mathrm{R}}$ are the same as those of $\left(A_{\mathrm{PQ}}, E_{\mathrm{PQ}}, C_{1 \mathrm{P}}, D_{1 \mathrm{PQ}}\right)$.
5. Orders of infinite zeros of the reduced order system, $\Sigma_{R}$, are reduced by one from those of ( $\left.A_{\mathrm{PQ}}, E_{\mathrm{PQ}}, C_{1 \mathrm{P}}, D_{1 \mathrm{PQ}}\right)$.

Proof. It follows from Proposition 2.2.1 of Chen [10].
因

Now, based on equation (6.3.8), we can construct a reduced order observer of $x_{2}$ as,

$$
\dot{\hat{x}}_{2}=A_{22} \hat{x}_{2}+A_{21} y_{1}+B_{2} u+K_{\mathrm{R}}\left(\left[\begin{array}{c}
y_{0} \\
\tilde{y}
\end{array}\right]-\left[\begin{array}{c}
C_{1,02} \\
A_{12}
\end{array}\right] \hat{x}_{2}\right)
$$

and

$$
\hat{x}_{\mathrm{PQ}}=\left[\begin{array}{c}
0 \\
I_{n-k}
\end{array}\right] \hat{x}_{2}+\left[\begin{array}{c}
I_{k} \\
0
\end{array}\right] y_{1},
$$

where $K_{\mathrm{R}}$ is the observer gain matrix for the reduced order system and is chosen such that

$$
A_{22}-K_{\mathrm{R}}\left[\begin{array}{c}
C_{1,02} \\
A_{12}
\end{array}\right]
$$

is asymptotically stable. In order to move the dependency on $\dot{y}_{1}$, let us partition $K_{\mathrm{R}}=\left[K_{\mathrm{R} 0}, K_{\mathrm{R} 1}\right]$ to be compatible with the dimensions of the output $\left(y_{0}^{\prime}, y_{1}^{\prime}\right)^{\prime}$. Then (see e.g. [53]), one can define a new variable $v:=\hat{x}_{2}-K_{\mathrm{R} 1} y_{1}$ and obtain a new dynamic equation,

$$
\begin{align*}
\dot{v}= & \left(A_{22}-K_{\mathrm{R} 0} C_{1,02}-K_{\mathrm{R} 1} A_{12}\right) v+\left(B_{2}-K_{\mathrm{R} 1} B_{1}\right) u \\
& +\left[K_{\mathrm{R} 0}, A_{21}-K_{\mathrm{R} 1} A_{11}+\left(A_{22}-K_{\mathrm{R} 0} C_{1,02}-K_{\mathrm{R} 1} A_{12}\right) K_{\mathrm{R} 1}\right]\binom{y_{0}}{y_{1}} . \tag{6.3.9}
\end{align*}
$$

Thus by implementing (6.3.9), $\hat{x}_{2}$ can be obtained without generating $\dot{y}_{1}$.

Theorem 6.3.1. Let $\Sigma_{\mathrm{PQ}}$ be given by (6.3.2). Then there exist for every $\varepsilon>0$, a state feedback gain $F$ and a reduced order observer gain matrix $K_{\mathrm{R}}$ such that the following reduced order observer based controller,

$$
\Sigma_{\mathrm{cmp}}:\left\{\begin{align*}
\dot{v}= & \left(A_{22}-K_{\mathrm{R} 0} C_{1,02}-K_{\mathrm{R} 1} A_{12}\right) v+\left(B_{2}-K_{\mathrm{R} 1} B_{1}\right) u  \tag{6.3.10}\\
& +\left[K_{\mathrm{R} 0}, A_{21}-K_{\mathrm{R} 1} A_{11}+\left(A_{22}-K_{\mathrm{R} 0} C_{1,02}-K_{\mathrm{R} 1} A_{12}\right) K_{\mathrm{R} 1}\right] y \\
u= & -F \hat{x}_{\mathrm{PQ}}=-F\left[\begin{array}{c}
0 \\
I_{n-k}
\end{array}\right] v-F\left[\begin{array}{cc}
0 & I_{k} \\
0 & K_{\mathrm{R} 1}
\end{array}\right] y
\end{align*}\right.
$$

when applied to $\Sigma_{\mathrm{PQ}}$ is internally stabilizing and yields an $H_{\infty}$ norm of the closed-loop transfer matrix from $w_{\mathrm{PQ}}$ to $h_{\mathrm{PQ}}$ strictly less than $\varepsilon$. Moreover, if $\Sigma_{\mathrm{cmp}}$ is applied to the original system $\Sigma$ of (6.2.1), then the resulting closedloop system comprising $\Sigma$ and $\Sigma_{\mathrm{cmp}}$ is internally stable and the $H_{\infty}$ norm of the closed-loop transfer matrix from $w$ to $h$ is less than $\gamma$.

Proof. See Subsection 6.4.E.
因

Remark 6.3.1. The gain matrix $F$ and $K_{\mathrm{R}}$ can be found using a systematic procedure given in Chapter 7.

Remark 6.3.2. In the case that the given system $\Sigma$ of (6.2.1) is regular, then the controller (6.3.10) reduces to the well-known full order observer based control design for the regular $H_{\infty}$-optimization as given in [39].

We illustrate the above result with a numerical example.

Example 6.3.1. We again consider a given measurement feedback system characterized by matrices $A, B, E, C_{2}, D_{2}$ as in Example 6.1.1 and $C_{1}, D_{1}$ as in Example 6.2.1. The infimum for this problem is $\gamma^{*}=13.638725$. In what follows, we will construct a reduced order measurement output feedback control law that makes the $H_{\infty}$ norm of the resulting closed-loop transfer matrix from $w$ to $h$ strictly less that $\gamma=14$. Following the procedure, we obtain an auxiliary system $\Sigma_{P Q}$ of the form (6.3.2) with

$$
A_{\mathrm{PQ}}=\left[\begin{array}{rrrrr}
4.2254 & -0.7415 & 4.1946 & 0 & 1.4335 \\
-11.8293 & 7.6804 & -13.7917 & 0 & -0.7102 \\
19.4695 & -9.0672 & 22.8277 & 0 & 4.0975 \\
-17.4591 & 10.0905 & -19.5135 & 1 & -2.1038 \\
1.2144 & 0.5197 & 1.4176 & 1 & -0.0983
\end{array}\right]
$$

$$
\begin{gathered}
B_{\mathrm{PQ}}=\left[\begin{array}{rrr}
0.9327 & 0 & 0 \\
-4.4755 & 0 & 0 \\
7.8569 & 0 & 0 \\
-6.3735 & 0 & 1 \\
0.1940 & 1 & 0
\end{array}\right], \quad E_{\mathrm{PQ}}=\left[\begin{array}{rrr}
18.5391 & 0.8299 \\
-62.8474 & -29.3560 \\
102.9481 & 28.5462 \\
-97.9601 & -22.3008 \\
-0.0958 & 3.1029
\end{array}\right], \\
C_{1 \mathrm{P}}=\left[\begin{array}{rrrrr}
0.1044 & -2.0724 & -2.9601 & -2 & -1 \\
1 & 2 & 3 & 2 & 1
\end{array}\right] \\
C_{2 \mathrm{P}}=\left[\begin{array}{rrrrr}
3.0616 & -0.9592 & 2.8464 & 0 & 0.6772 \\
-1.0146 & -1.3601 & 0.6330 & 0 & -0.7358
\end{array}\right]
\end{gathered}
$$

and

$$
D_{1 \mathrm{P}}=\left[\begin{array}{rr}
0.9409 & -0.3383 \\
0 & 0
\end{array}\right], \quad D_{2 \mathrm{PQ}}=\left[\begin{array}{rr}
0.9409 & -0.3383 \\
0 & 0
\end{array}\right] .
$$

It is simple to show that the transformation $T_{s}$ and $T_{o}$,

$$
T_{s}=\left[\begin{array}{rrrrr}
1 & -2 & -3 & -2 & -1 \\
0 & 1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1 & 0 \\
0 & 0 & 0 & 0 & 1
\end{array}\right], \quad T_{o}=\left[\begin{array}{rr}
1 & 0.1044 \\
0 & 1
\end{array}\right]
$$

will transform $C_{1}$ and $D_{1}$ to the following form,

$$
T_{o}^{-1} C_{1 \mathrm{p}} T_{s}=\left[\begin{array}{c|c}
0 & C_{1,02} \\
\hline I_{k} & 0
\end{array}\right]=\left[\begin{array}{c|cccc}
0 & -2.2811 & -3.2732 & -2.2087 & -1.1044 \\
\hline 1 & 0 & 0 & 0 & 0
\end{array}\right]
$$

and

$$
T_{o}^{-1} D_{1 \mathrm{PQ}}=\left[\frac{D_{1,0}}{0}\right]=\left[\begin{array}{cc}
0.9409 & -0.3383 \\
\hline 0 & 0
\end{array}\right]
$$

Moreover, we have

$$
\begin{aligned}
& T_{s}^{-1} A T_{s}= {\left[\begin{array}{r|r}
A_{11} & A_{12} \\
\hline A_{21} & A_{22}
\end{array}\right] } \\
&= {\left[\begin{array}{rrrrr}
5.2714 & -2.4247 & -8.3291 & -7.5428 & 2.7283 \\
\hline-11.8293 & 31.3390 & 21.6962 & 23.6586 & 11.1191 \\
19.4695 & -48.0062 & -35.5807 & -38.9390 & -15.3720 \\
-17.4591 & 45.0087 & 32.8639 & 35.9182 & 15.3553 \\
1.2144 & -1.9092 & -2.2257 & -1.4289 & -1.3127
\end{array}\right] } \\
& T_{s}^{-1} B=\left[\begin{array}{r}
B_{1} \\
\hline B_{2}
\end{array}\right]=\left[\begin{array}{rrr}
2.9993 & 1 & 2 \\
\hline-4.4755 & 0 & 0 \\
7.8569 & 0 & 0 \\
-6.3735 & 0 & 1 \\
0.1940 & 1 & 0
\end{array}\right]
\end{aligned}
$$

$$
T_{s}^{-1} E=\left[\frac{E_{1}}{E_{2}}\right]=\left[\begin{array}{rr}
5.6724 & -13.7425 \\
\hline-62.8474 & -29.3560 \\
102.9481 & 28.5462 \\
-97.9601 & -22.3008 \\
-0.0958 & 3.1029
\end{array}\right]
$$

and $A_{\mathrm{R}}=A_{22}, E_{\mathrm{R}}=E_{2}$,

$$
C_{\mathrm{R}}=\left[\begin{array}{llrr}
-2.2811 & -3.2732 & -2.2087 & -1.1044 \\
-2.4247 & -8.3291 & -7.5428 & 2.7283
\end{array}\right]
$$

and

$$
D_{\mathrm{R}}=\left[\begin{array}{rr}
0.9409 & -0.3383 \\
5.6724 & -13.7425
\end{array}\right] .
$$

Using the algorithm given in Chapter 7, we obtain a gain matrix $F$,

$$
F T_{s}=\left[\begin{array}{rrrrr}
-1.5656 & 4.7579 & 2.1737 & 3.1311 & 1.5656 \\
-299.4859 & 555.2644 & 742.6408 & 597.9718 & 189.8014 \\
7.4811 & -14.6842 & -19.0100 & -16.9623 & -5.3773
\end{array}\right]
$$

and

$$
K_{\mathrm{R}}=\left[K_{\mathrm{R} 0} \mid K_{\mathrm{R} 1}\right]=\left[\begin{array}{r|c}
93.5515 & -4.4388 \\
-143.1777 & 5.6013 \\
133.7360 & -4.9145 \\
-1.4788 & 0.2622
\end{array}\right]
$$

Finally, we obtain a reduced order output feedback controller of the form (6.3.1) with

$$
\begin{aligned}
A_{\mathrm{cmp}} & =10^{3} \cdot\left[\begin{array}{rrrr}
-2.5903 & -3.4139 & -2.7089 & -0.9269 \\
3.3280 & 4.3868 & 3.4717 & 1.1995 \\
-2.9478 & -3.8917 & -3.0775 & -1.0641 \\
0.6986 & 0.9299 & 0.7488 & 0.2393
\end{array}\right], \\
C_{\mathrm{cmp}} & =\left[\begin{array}{rrrr}
4.7579 & 2.1737 & 3.1311 & 1.5656 \\
555.2644 & 742.6408 & 597.9718 & 189.8014 \\
-14.6842 & -19.0100 & -16.9623 & -5.3773
\end{array}\right],
\end{aligned}
$$

and

$$
B_{\mathrm{cmp}}=10^{3} \cdot\left[\begin{array}{rr}
-0.0936 & -4.1798 \\
0.1432 & 5.3492 \\
-0.1337 & -4.7217 \\
0.0015 & 1.1362
\end{array}\right], \quad D_{\mathrm{cmp}}=\left[\begin{array}{rr}
0 & 22.3556 \\
0 & 894.3952 \\
0 & -33.1683
\end{array}\right]
$$

which yields the poles of the closed-loop system, when it is applied to the given system, at

$$
-97.337,-34.72,-3.591,-1.848,-1.632,-0.248,-1.346,-0.765,-1
$$

Obviously, they are in the stable region. The singular value plots of the resulting closed-loop transfer matrix $T_{h w}$ in Figure 6.3.1 also show that $\left\|T_{h w}\right\|_{\infty}$ is indeed less than 14 , the given $\gamma$.


Figure 6.3.1: Max. singular values of $T_{h w}$ under reduced order output feedback.

### 6.4. Proofs of Main Results

### 6.4.A. Proof of Theorem 6.1.1

We need to recall the following two lemmas in order to proceed with our proof of Theorem 6.1.1.

Lemma 6.4.1. Let an auxiliary system $\Sigma_{\text {aux }}$ be characterized by

$$
\Sigma_{\mathrm{aux}}:\left\{\begin{array}{l}
\dot{x}_{x}=A_{x} x_{x}+B_{x} u_{x}+E_{x} w_{x},  \tag{6.4.1}\\
h_{x}=C_{x} x_{x}+D_{x} u_{x}
\end{array}\right.
$$

where

$$
A_{x}=A_{11 \mathrm{P}}, \quad B_{x}=\left[\begin{array}{ll}
B_{11 \mathrm{P}} & A_{13 \mathrm{P}}
\end{array}\right], \quad E_{x}=\left[\begin{array}{c}
E_{a \mathrm{P}}^{+} \\
0
\end{array}\right]
$$

and

$$
C_{x}=\Gamma_{o \mathrm{P}}\left[\begin{array}{cc}
0 & 0 \\
0 & 0 \\
0 & C_{b \mathrm{P}}
\end{array}\right], \quad D_{x}=\Gamma_{o \mathrm{P}}\left[\begin{array}{cc}
I & 0 \\
0 & C_{d \mathrm{P}} C_{d \mathrm{P}}^{\prime} \\
0 & 0
\end{array}\right]
$$

Then $\Sigma_{\text {aux }}$ comprising the state feedback law $u_{x}=-F_{11}(\gamma) x_{x}$ is internally stable, i.e.,

$$
\begin{equation*}
\lambda\left(A_{11 \mathrm{P}}^{c}\right)=\lambda\left\{A_{11 \mathrm{P}}-\left[B_{11 \mathrm{P}}, A_{13 \mathrm{P}}\right] F_{11}(\gamma)\right\}=\lambda\left\{A_{x}-B_{x} F_{11}(\gamma)\right\} \subset \mathbb{C}^{-} \tag{6.4.2}
\end{equation*}
$$

and the resulting closed-loop transfer function from $w_{x}$ to $h_{x}$ has $H_{\infty}$ norm less than $\gamma$, i.e.,

$$
\left\|T_{h_{x} w_{x}}\right\|_{\infty}=\left\|\Gamma_{o \mathrm{P}}\left[\begin{array}{c}
-F_{11}(\gamma)  \tag{6.4.3}\\
0
\end{array} C_{b \mathrm{P}}\right]\left[s I-A_{11 \mathrm{P}}^{c}\right)^{-1}\left[\begin{array}{c}
E_{a \mathrm{P}}^{+} \\
0
\end{array}\right]\right\|_{\infty}<\gamma
$$

That is $u_{x}=-F_{11}(\gamma) x_{x}$ is a $\gamma$-suboptimal control law for $\Sigma_{\text {aux }}$.
回

Proof. We first note that $\Gamma_{o \mathrm{P}}$ is nonsingular and $C_{d \mathrm{P}} C_{d \mathrm{P}}^{\prime}=I$ which implies that $D_{x}$ is injective. Furthermore, it is simple to verify that the invariant zeros of $\left(A_{x}, B_{x}, C_{x}, D_{x}\right)$ are given by $\lambda\left(A_{a a \mathrm{P}}^{+}\right)$, and are not on the imaginary axis. Hence $\Sigma_{\text {aux }}$ satisfies the assumptions of the regular $H_{\infty}$ control problem. Moreover, it is straightforward to verify that for any $\gamma>\gamma^{*}$,

$$
P_{x}=\left(S_{x \mathrm{P}}-\gamma^{-2} T_{x \mathrm{P}}\right)^{-1}>0,
$$

is the solution of the following well-known $H_{\infty}$-CARE:

$$
\begin{align*}
P_{x} A_{x}+A_{x}^{\prime} P_{x} & +\gamma^{-2} P_{x} E_{x} E_{x}^{\prime} P_{x}+C_{x}^{\prime} C_{x} \\
& -\left[P_{x} B_{x}+C_{x}^{\prime} D_{x}\right]\left(D_{x}^{\prime} D_{x}\right)^{-1}\left[B_{x}^{\prime} P_{x}+D_{x}^{\prime} C_{x}\right]=0 \tag{6.4.4}
\end{align*}
$$

with

$$
\left.\lambda\left(A_{x x}^{c}\right):=\lambda\left\{A_{x}+\gamma^{-2} E_{x} E_{x}^{\prime} P_{x}-B_{x}\left(D_{x}^{\prime} D_{x}\right)^{-1}\left(B_{x}^{\prime} P x\right)+D_{x} C_{x}\right)\right\} \in \mathbb{C}^{-}
$$

Then the results of Lemma 6.4.1 follow.
区
Lemma 6.4.2. Let $(A, B, C)$, where $A \in \mathbb{R}^{n \times n}, B \in \mathbb{R}^{n \times m}$ and $C \in \mathbb{R}^{p \times n}$, be right invertible and of minimum phase. Let $F(\varepsilon) \in \mathbb{R}^{m \times n}$ be parameterized in terms of $\varepsilon$ and be of the form,

$$
\begin{equation*}
F(\varepsilon)=N(\varepsilon) \Gamma(\varepsilon) T(\varepsilon)+R(\varepsilon) \tag{6.4.5}
\end{equation*}
$$

where $N(\varepsilon) \in \mathbb{R}^{m \times p}, \Gamma(\varepsilon) \in \mathbb{R}^{p \times p}, T(\varepsilon) \in \mathbb{R}^{p \times n}$ and $R(\varepsilon) \in \mathbb{R}^{m \times n}$. Also, $\Gamma(\varepsilon)$ is nonsingular. Moreover, assume that the following conditions hold:

1. $A+B F(\varepsilon)$ is asymptotically stable for all $0<\varepsilon \leq \varepsilon^{*}$ where $\varepsilon^{*}>0$;
2. $T(\varepsilon) \rightarrow W C$ as $\varepsilon \rightarrow 0$ where $W$ is some $p \times p$ nonsingular matrix;
3. as $\varepsilon \rightarrow 0, N(\varepsilon)$ tends to some finite matrix $N$ such that $C(s I-A)^{-1} B N$ is invertible;
4. as $\varepsilon \rightarrow 0, R(\varepsilon)$ tends to some finite matrix $R$; and
5. $\Gamma^{-1}(\varepsilon) \rightarrow 0$ as $\varepsilon \rightarrow 0$.

Then as $\varepsilon \rightarrow 0$, we have $\left\|C[s I-A-B F(\varepsilon)]^{-1}\right\|_{\infty} \rightarrow 0$.
Proof. This is a dual version of Lemma 2.2 given by Saberi and Sannuti [91]. The proof of this lemma follows from similar arguments as in [91].

Now we are ready to proceed with the proof of Theorem 6.1.1. Note that $F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)$ is constructed under the standard ATEA procedure. It can be shown using the techniques of the well-known singular perturbation theory as in Chen [10] that as $\varepsilon \rightarrow 0$, the eigenvalues of $A+B F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)$ are given by $\lambda\left(A_{\text {aap }}^{-}\right) \in \mathbb{C}^{-}, \lambda\left(A_{c c \mathrm{P}}^{c}\right) \in \mathbb{C}^{-}, \Lambda_{d \mathrm{P}} / \varepsilon \in \mathbb{C}^{-}$and $\lambda\left(A_{11 \mathrm{P}}^{c}\right) \in \mathbb{C}^{-}$(see Lemma 6.4.1). Hence the closed-loop is internally stable. Moreover, following the results of Chen [10], it can be shown that for any $\lambda_{d} \in \Lambda_{d \mathrm{p}} / \varepsilon \in \mathbb{C}^{-}$, the corresponding right eigenvector, say $W(\varepsilon)$, satisfies

$$
\begin{equation*}
\lim _{\varepsilon \rightarrow 0} W(\varepsilon)=\bar{W} \in \mathcal{S}^{+}\left(\Sigma_{\mathrm{P}}\right) \tag{6.4.6}
\end{equation*}
$$

In fact, following the same arguments, one can show that as $\varepsilon \rightarrow 0$, the eigenvalues of $A+\gamma^{-2} E E^{\prime} P(\gamma)+B F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)$, where $P(\gamma)$ is as defined in (6.2.2), are given by $\lambda\left(A_{\text {aap }}^{-}\right) \in \mathbb{C}^{-}, \lambda\left(A_{c c \mathrm{P}}^{c}\right) \in \mathbb{C}^{-}, \Lambda_{d \mathrm{P}} / \varepsilon \in \mathbb{C}^{-}$and $\lambda\left(A_{x x}^{c}\right) \in \mathbb{C}^{-}$. We will use these properties later on in our proofs of other theorems. This proves the second part of Theorem 6.1.1.

Next, we show that the state feedback law $u=F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right) x$ yields
$\left\|T_{h w}\right\|_{\infty}=\left\|\left[C_{2}+D_{2} F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)\right]\left[s I-A-B F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)\right]^{-1} E\right\|_{\infty}<\gamma$.
Without loss of generality but for simplicity of presentation, we assume that the nonsingular transformations $\Gamma_{s \mathrm{P}}=I$ and $\Gamma_{i \mathrm{P}}=I$, i.e., we assume that the system $\left(A, B, \Gamma_{o \mathrm{P}}^{-1} C_{2}, \Gamma_{o \mathrm{P}}^{-1} D_{2}\right)$ is in the form of the special coordinate basis. In view of (6.1.14), let us partition $F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)$ as,

$$
F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)=\bar{F}_{0}(\gamma)+\left[\begin{array}{c}
0 \\
\bar{F}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)
\end{array}\right],
$$

where

$$
\bar{F}_{0}(\gamma)=-\left[\begin{array}{ccccc}
C_{0 a \mathrm{P}}^{+}+F_{a 0}^{+}(\gamma) & C_{0 b \mathrm{P}}+F_{b 0}(\gamma) & C_{0 a \mathrm{P}}^{-} & C_{0 c \mathrm{P}} & C_{0 d \mathrm{P}} \\
0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0
\end{array}\right]
$$

and

$$
\bar{F}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)=-\left[\begin{array}{ccc}
E_{d a \mathrm{P}}^{+}+\tilde{F}_{a 1}^{+}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}\right) & E_{d b \mathrm{P}}+\tilde{F}_{b 1}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}\right) \\
E_{c a \mathrm{P}}^{+} & & E_{c b \mathrm{P}}  \tag{6.4.7}\\
E_{d a \mathrm{P}}^{-} & E_{d c \mathrm{P}} & \tilde{F}_{f}\left(\varepsilon, \Lambda_{d \mathrm{P}}\right)+E_{d \mathrm{P}} \\
E_{c a \mathrm{P}}^{-} & \Delta_{c \mathrm{P}} & 0
\end{array}\right] .
$$

Then we have

$$
\bar{C}=C_{2}+D_{2} F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)=\Gamma_{o \mathrm{P}}\left[\begin{array}{ccccc}
-F_{a 0}^{+}(\gamma) & -F_{b 0}(\gamma) & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & C_{d \mathrm{P}} \\
0 & C_{b \mathrm{P}} & 0 & 0 & 0
\end{array}\right]
$$

and

$$
\bar{A}=A+B \bar{F}_{0}(\gamma), \quad \bar{B}=\left[\begin{array}{cc}
0 & 0  \tag{6.4.8}\\
0 & 0 \\
0 & 0 \\
0 & B_{c \mathrm{P}} \\
B_{d \mathrm{P}} & 0
\end{array}\right]
$$

With these definitions, we can write $T_{h w}$ as

$$
T_{h w}=\bar{C}\left[s I-\bar{A}-\bar{B} \bar{F}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)\right]^{-1} E .
$$

Then in view of (6.4.7), it can easily be seen that $\bar{F}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)$ has the form,

$$
\bar{F}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)=N \Gamma(\varepsilon) T(\varepsilon)+R,
$$

where

$$
\Gamma(\varepsilon)=\operatorname{diag}\left[\frac{1}{\varepsilon^{q_{1}}}, \frac{1}{\varepsilon^{q_{2}}}, \cdots, \frac{1}{\varepsilon^{q_{m_{d \mathrm{P}}}}}\right], \quad N=-\left[\begin{array}{c}
I_{m_{d \mathrm{P}}} \\
0
\end{array}\right]
$$

and

$$
R=-\left[\begin{array}{ccccc}
E_{d a \mathrm{P}}^{+} & E_{d b \mathrm{P}} & E_{d a \mathrm{P}}^{-} & E_{d c \mathrm{P}} & E_{d \mathrm{P}} \\
E_{c a \mathrm{P}}^{+} & E_{c b \mathrm{P}} & E_{c a \mathrm{P}}^{-} & \Delta_{c \mathrm{P}} & 0
\end{array}\right]
$$

while $T(\varepsilon)$ satisfies

$$
T(\varepsilon) \rightarrow T C_{m}
$$

as $\varepsilon \rightarrow 0$, where

$$
T=\operatorname{diag}\left[F_{1 q_{1}}, F_{1 q_{2}}, \cdots, F_{\left.m_{d \mathrm{P} q_{m_{d \mathrm{P}}}}\right], ~}\right.
$$

and

$$
C_{m}=\left[\begin{array}{lllll}
F_{a 1}^{+}(\gamma) & F_{b 1}(\gamma) & 0 & 0 & C_{d \mathrm{P}} \tag{6.4.9}
\end{array}\right] .
$$

Using the same arguments as in Chen et al [27], it is straightforward to show that the triple $\left(\bar{A}, \bar{B}, C_{m}\right)$ is right invertible and of minimum phase. Thus, it follows from Lemma 6.4.2 that

$$
\left\|C_{m}\left[s I-\bar{A}-\bar{B} \bar{F}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)\right]^{-1}\right\|_{\infty} \rightarrow 0
$$

as $\varepsilon \rightarrow 0$. We should also note that following the same line of reasoning, one can show that the triple $\left(\bar{A}+\gamma^{-2} E E^{\prime} P(\gamma), \bar{B}, C_{m}\right)$ is right invertible and of minimum phase, and moreover as $\varepsilon \rightarrow 0$,

$$
\begin{equation*}
\left\|C_{m}\left[s I-\bar{A}-\gamma^{-2} E E^{\prime} P(\gamma)-\bar{B} \bar{F}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)\right]^{-1}\right\|_{\infty} \rightarrow 0 \tag{6.4.10}
\end{equation*}
$$

Next, let

$$
\bar{C}=\Gamma_{o \mathbf{P}}\left[\begin{array}{c}
0 \\
C_{m} \\
0
\end{array}\right]+C_{e}
$$

where

$$
C_{e}=\Gamma_{o \mathrm{P}}\left[\begin{array}{ccccc}
-F_{a 0}^{+}(\gamma) & -F_{b 0}(\gamma) & 0 & 0 & 0 \\
-F_{a 1}^{+}(\gamma) & -F_{b 1}(\gamma) & 0 & 0 & 0 \\
0 & C_{b \mathrm{P}} & 0 & 0 & 0
\end{array}\right] .
$$

We have

$$
\left\|T_{h w}\right\|_{\infty} \rightarrow\left\|C_{e}\left[s I-\bar{A}-\bar{B} \bar{F}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)\right]^{-1} E\right\|_{\infty}
$$

as $\varepsilon \rightarrow 0$. Following the procedures of Chen [10] or Saberi, Chen and Sannuti [87], it can be shown that
$C_{e}\left[s I-\bar{A}-\bar{B} \bar{F}\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{d \mathrm{P}}\right)\right]^{-1} E \rightarrow \Gamma_{o \mathrm{P}}\left[\begin{array}{c}-F_{11}(\gamma) \\ {\left[\begin{array}{ll}0 & C_{b \mathrm{P}}\end{array}\right]}\end{array}\right]\left(s I-A_{11 \mathrm{P}}^{c}\right)^{-1}\left[\begin{array}{c}E_{a \mathrm{P}}^{+} \\ 0\end{array}\right]$,
pointwise in $s$ as $\varepsilon \rightarrow 0$. Hence, the results of Theorem 6.1.1 follow readily from Lemma 6.4.1.

### 6.4.B. Proof of Theorem 6.1.2

Without loss of generality but for simplicity of presentation, we assume that the nonsingular state and input transformations $\Gamma_{s \mathrm{P}}=I$ and $\Gamma_{i \mathrm{P}}=I$, i.e., the system $\left(A, B, \Gamma_{o \mathrm{P}}^{-1} C_{2}, \Gamma_{o \mathrm{P}}^{-1} D_{2}\right)$ is in the form of the special coordinate basis. Then it is trivial to show that

$$
A+B F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right)=\left[\begin{array}{cccc}
\star & 0 & 0 & \star \\
\star & A_{a a \mathrm{P}}^{-} & 0 & \star \\
\star & 0 & A_{c c \mathrm{P}}^{c} & \star \\
\star & 0 & 0 & \star
\end{array}\right],
$$

and

$$
C_{2}+D_{2} F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right)=\Gamma_{o \mathrm{P}}\left[\begin{array}{cccc}
\star & 0 & 0 & 0 \\
0 & 0 & 0 & \star \\
\star & 0 & 0 & 0
\end{array}\right]
$$

where $\star$ 's represent some sub-matrices which are of no interest to our proof. Hence, for any $\alpha \in \lambda\left(A_{a \mathrm{aP}}^{-}\right) \cup \lambda\left(A_{c c \mathrm{P}}^{c}\right)$, the corresponding right eigenvector is in the kernel of $C_{2}+D_{2} F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right)$. This proves that $\alpha$ is an output decoupling zero of $T_{h w}$.

### 6.4.C. Proof of Theorem 6.2.1

For the sake of simplicity in presentation, we drop in the following proof the arguments of $F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right)$ and $K\left(\gamma, \varepsilon, \Lambda_{d \mathrm{Q}}, \Delta_{c \mathrm{Q}}\right)$. Also, we assume without loss of generality that $\gamma=1$. Thus, we will drop the dependency of $\gamma$ in all the variables.

First, it is simple to verify that the positive semi-definite matrices $P$ of (6.2.2) and $Q$ of (6.2.4) satisfy

$$
F_{\gamma}(P):=\left[\begin{array}{cc}
A^{\prime} P+P A+C_{2}^{\prime} C_{2}+P E E^{\prime} P & P B+C_{2}^{\prime} D_{2} \\
B^{\prime} P+D_{2}^{\prime} C_{2} & D_{2}^{\prime} D_{2}
\end{array}\right] \geq 0
$$

and

$$
G_{\gamma}(Q):=\left[\begin{array}{cc}
A Q+Q A^{\prime}+E E^{\prime}+Q C_{2}^{\prime} C_{2} Q & Q C_{1}^{\prime}+E D_{1}^{\prime} \\
C_{1} Q+D_{1} E^{\prime} & D_{1} D_{1}^{\prime}
\end{array}\right] \geq 0
$$

respectively, i.e., $P$ and $Q$ are the solutions of the quadratic matrix inequalities $F_{\gamma}(P) \geq 0$ and $G_{\gamma}(Q) \geq 0$. Moreover, the following auxiliary system,

$$
\Sigma_{\mathrm{PQ}}:\left\{\begin{align*}
\dot{x}_{\mathrm{PQ}} & =A_{\mathrm{PQ}} x_{\mathrm{PQ}}+B_{\mathrm{PQ}} u+E_{\mathrm{PQ}} w_{\mathrm{PQ}}  \tag{6.4.11}\\
y & =C_{1 \mathrm{P}} x_{\mathrm{PQ}} \\
h_{\mathrm{PQ}} & =C_{2 \mathrm{P}} x_{\mathrm{PQ}}+D_{2 \mathrm{P}} u,
\end{align*}\right.
$$

where

$$
F_{\gamma}(P)=\left[\begin{array}{c}
C_{2 \mathrm{P}}^{\prime} \\
D_{2 \mathrm{P}}^{\prime}
\end{array}\right]\left[\begin{array}{ll}
C_{2 \mathrm{P}} & D_{2 \mathrm{P}}
\end{array}\right], \quad G_{\gamma}(Q)=\left[\begin{array}{c}
E_{\mathrm{Q}} \\
D_{1 \mathrm{PQ}}
\end{array}\right]\left[\begin{array}{ll}
E_{\mathrm{Q}}^{\prime} & D_{1 \mathrm{PQ}}^{\prime}
\end{array}\right]
$$

and

$$
\left.\begin{array}{l}
A_{\mathrm{PQ}}:=A+E E^{\prime} P+(I-Q P)^{-1} Q C_{2 \mathrm{P}}^{\prime} C_{2 \mathrm{P}},  \tag{6.4.12}\\
B_{\mathrm{PQ}}:=B+(I-Q P)^{-1} Q C_{2 \mathrm{P}}^{\prime} D_{2 \mathrm{P}}, \\
E_{\mathrm{PQ}}:=(I-Q P)^{-1} E_{\mathrm{Q}}, \\
C_{1 \mathrm{P}}:=C_{1}+D_{1} E^{\prime} P,
\end{array}\right\}
$$

has the following properties: 1) the subsystem $\left(A_{\mathrm{PQ}}, B_{\mathrm{PQ}}, C_{2 \mathrm{P}}, D_{2 \mathrm{P}}\right)$ is right invertible and of minimum phase; and 2) the subsystem ( $A_{\mathrm{PQ}}, E_{\mathrm{PQ}}, C_{1 \mathrm{P}}, D_{1 \mathrm{PQ}}$ ) is left invertible and of minimum phase.

The following lemma is due to Stoorvogel [100].
Lemma 6.4.3. For any given compensator $\Sigma_{\mathrm{cmp}}$ of the form

$$
\Sigma_{\mathrm{cmp}}:\left\{\begin{array}{l}
\dot{v}=A_{\mathrm{cmp}} v+B_{\mathrm{cmp}} y \\
u=C_{\mathrm{cmp}} v+D_{\mathrm{cmp}} y
\end{array}\right.
$$

The following two statements are equivalent:

1. $\Sigma_{\mathrm{cmp}}$ applied to the system $\Sigma$ defined by (6.2.1) is internally stabilizing and the resulting closed-loop transfer function from $w$ to $h$ has an $H_{\infty}$ norm less than 1, i.e., $\left\|T_{h w}\right\|_{\infty}<1$.
2. $\Sigma_{\mathrm{cmp}}$ applied to the new system $\Sigma_{\mathrm{PQ}}$ defined by (6.4.11) is internally stabilizing and the resulting closed loop transfer function from $w_{P Q}$ to $h_{\mathrm{PQ}}$ has an $H_{\infty}$ norm less than 1, i.e., $\left\|T_{h_{\mathrm{PQ}} w_{\mathrm{PQ}}}\right\|_{\infty}<1$.

Hence, it is sufficient to show Theorem 6.2.1 by showing that $\Sigma_{\text {cmp }}$ of (6.2.5) to (6.2.8) applied to $\Sigma_{\mathrm{PQ}}$ achieves almost disturbance decoupling with internal stability. Observing that

$$
C_{2 \mathrm{P}}^{\prime} C_{2 \mathrm{P}}=A^{\prime} P+P A+C_{2}^{\prime} C_{2}+P E E^{\prime} P \quad \text { and } \quad C_{2 \mathrm{P}}^{\prime} D_{2 \mathrm{P}}=P B+C_{2}^{\prime} D_{2}
$$

it is simple to rewrite $A_{\text {cmp }}$ of (6.2.6) as

$$
A_{\mathrm{cmp}}=A_{\mathrm{PQ}}+B_{\mathrm{PQ}} F+(I-Q P)^{-1} K C_{1 \mathrm{P}}
$$

Now it is trivial to see that $\Sigma_{\text {cmp }}$ of (6.2.5) is simply the well-known full order observer based controller for the system $\Sigma_{\mathrm{PQ}}$ with state feedback gain $F$ and observer gain $(I-Q P)^{-1} K$. Hence the well-known separation principle holds. Also, noting the facts that ( $A_{\mathrm{PQ}}, B_{\mathrm{PQ}}, C_{2 \mathrm{P}}, D_{2 \mathrm{P}}$ ) and ( $A_{\mathrm{PQ}}, E_{\mathrm{PQ}}, C_{1 \mathrm{P}}, D_{1 \mathrm{PQ}}$ ) are of minimum phase, and right invertible and left invertible, respectively, it is sufficient to prove Theorem 6.2 .1 by showing that as $\varepsilon \rightarrow 0$,

1. $A_{\mathrm{PQ}}+B_{\mathrm{PQ}} F$ is asymptotically stable;
2. $\left\|\left[C_{2 \mathrm{P}}+D_{2 \mathrm{P}} F\right]\left[s I-A_{\mathrm{PQ}}-B_{\mathrm{PQ}} F\right]^{-1}\right\|_{\infty} \rightarrow 0$;
3. $A_{\mathrm{PQ}}+(I-Q P)^{-1} K C_{1 \mathrm{P}}$ is asymptotically stable; and
4. $\left\|\left[s I-A_{\mathrm{PQ}}-(I-Q P)^{-1} K C_{1 \mathrm{P}}\right]^{-1}\left[E_{\mathrm{PQ}}+(I-Q P)^{-1} K D_{1 \mathrm{PQ}}\right]\right\|_{\infty} \rightarrow 0$.

We shall introduce the following lemma for further development.
Lemma 6.4.4. As $\varepsilon \rightarrow 0$, we have

1. $A+E E^{\prime} P+B F$ is asymptotically stable and

$$
\begin{equation*}
\left\|\left[C_{2 \mathrm{P}}+D_{2 \mathrm{P}} F\right]\left[s I-A-E E^{\prime} P-B F\right]^{-1}\right\|_{\infty} \rightarrow 0 \tag{6.4.13}
\end{equation*}
$$

2. $A+Q C_{2}^{\prime} C_{2}+K C_{1}$ is asymptotically stable and

$$
\begin{equation*}
\left\|\left[s I-A-Q C_{2}^{\prime} C_{2}-K C_{1}\right]^{-1}\left[E_{\mathrm{Q}}+K D_{1 \mathrm{PQ}}\right]\right\|_{\infty} \rightarrow 0 \tag{6.4.14}
\end{equation*}
$$

Proof. It is shown in the proof of Theorem 6.1.1 that for $\varepsilon \rightarrow 0$, the matrix $A+E E^{\prime} P+B F$ is asymptotically stable. In what follows, we will show (6.4.13). By some elementary algebra, it can be shown that

$$
C_{2 \mathrm{P}}=\Gamma_{o \mathrm{P}}\left[\begin{array}{ccccc}
C_{0 a \mathrm{P}}^{+}+F_{a 0}^{+} & C_{0 b \mathrm{P}}+F_{b 0} & C_{0 a \mathrm{P}}^{-} & C_{0 c \mathrm{P}} & C_{0 d \mathrm{P}} \\
F_{a 1}^{+} & F_{b 1} & 0 & 0 & C_{d \mathrm{P}} \\
0 & 0 & 0 & 0 & 0
\end{array}\right] \Gamma_{s \mathrm{P}}^{-1},
$$

and

$$
D_{2 \mathrm{P}}=D_{2}=\Gamma_{o \mathrm{P}}\left[\begin{array}{lll}
I & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right] \Gamma_{i \mathrm{P}}^{-1}
$$

Moreover,

$$
\left[C_{2 \mathrm{p}}+D_{2 \mathrm{p}} F\right]\left[s I-A-E E^{\prime} P-B F\right]^{-1}=\left[\begin{array}{c}
0 \\
C_{m} \\
0
\end{array}\right]\left[s I-\bar{A}-E E^{\prime} P-\bar{B} \bar{F}\right]^{-1}
$$

where $\bar{A}$ and $\bar{B}$ are as in (6.4.8), $\bar{F}$ is as in (6.4.7) and $C_{m}$ is given by (6.4.9). In view of (6.4.10), we have the result.

Item 2 of Lemma 6.4 .4 is the dual version of item 1. Hence, the results follow. This completes the proof of Lemma 6.4.4.

Next, we will first show that $A_{\mathrm{PQ}}+B_{\mathrm{PQ}} F$ is asymptotically stable for some sufficiently small $\varepsilon$ and

$$
\left\|\left[C_{2 \mathrm{P}}+D_{2 \mathrm{P}} F\right]\left[s I-A_{\mathrm{PQ}}-B_{\mathrm{PQ}} F\right]^{-1}\right\|_{\infty} \rightarrow 0
$$

as $\varepsilon \rightarrow 0$. In view of Lemma 6.4.4, we have

$$
\begin{aligned}
s I- & A_{\mathrm{PQ}}-B_{\mathrm{PQ}} F \\
& =s I-A-E E^{\prime} P-B F-(I-Q P)^{-1} Q C_{2 \mathrm{P}}^{\prime}\left[C_{2 \mathrm{P}}+D_{x \mathrm{P}} F\right] \\
& =\left\{I-(I-Q P)^{-1} Q C_{2 \mathrm{P}}^{\prime}\left[C_{2 \mathrm{P}}+D_{x \mathrm{P}} F\right]\left[s I-A-E E^{\prime} P-B F\right]^{-1}\right\} \\
& \cdot\left[s I-A-E E^{\prime} P-B F\right] \\
& \rightarrow s I-A-E E^{\prime} P-B F \text { pointwise in } s \text { as } \varepsilon \rightarrow 0 .
\end{aligned}
$$

This implies that $A_{\mathrm{PQ}}+B_{\mathrm{PQ}} F$ is asymptotically stable for sufficiently small $\varepsilon$, and

$$
\begin{align*}
{\left[C_{2 \mathrm{P}}+\right.} & \left.D_{2 \mathrm{P}} F\right]\left[s I-A_{\mathrm{PQ}}-B_{\mathrm{PQ}} F\right]^{-1} \\
= & {\left[C_{2 \mathrm{P}}+D_{2 \mathrm{P}} F\right]\left[s I-A-E E^{\prime} P-B F\right]^{-1} } \\
& \cdot\left\{I-(I-Q P)^{-1} Q C_{2 \mathrm{P}}^{\prime}\left[C_{2 \mathrm{P}}+D_{x \mathrm{P}} F\right]\left[s I-A-E E^{\prime} P-B F\right]^{-1}\right\}^{-1} \\
\rightarrow & 0, \quad \text { pointwise in } s \text { as } \varepsilon \rightarrow 0 . \tag{6.4.15}
\end{align*}
$$

Again, in view of Lemma 6.4.4 and

$$
\begin{gathered}
C_{2 \mathrm{p}}^{\prime} C_{2 \mathrm{P}}=A^{\prime} P+P A+C_{2}^{\prime} C_{2}+P E E^{\prime} P \\
E_{\mathrm{Q}} E_{\mathrm{Q}}^{\prime}=A Q+Q A^{\prime}+E E^{\prime}+Q C_{2}^{\prime} C_{2} Q
\end{gathered}
$$

we have the following induction:

$$
\begin{align*}
&(I-Q P)\left[s I-A_{\mathrm{PQ}}-(I-Q P)^{-1} L C_{1 \mathrm{P}}\right] \\
&=\left[(I-Q P)\left(s I-A-E E^{\prime} P\right)-Q C_{2 \mathrm{P}}^{\prime} C_{2 \mathrm{P}}-L C_{1}-L D_{1} E^{\prime} P\right] \\
&=\left[s I-A-E E^{\prime} P-Q C_{2 \mathrm{P}}^{\prime} C_{2 \mathrm{P}}-L C_{1}-L D_{1} E^{\prime} P-s Q P\right. \\
&\left.\quad+Q P A+Q P E E^{\prime} P\right] \\
&=\left[s I-A-E E^{\prime} P-Q\left(A^{\prime} P+P A+C_{2}^{\prime} C_{2}+P E E^{\prime} P\right)\right. \\
&\left.\quad-L C_{1}-L D_{1} E^{\prime} P-s Q P+Q P A+Q P E E^{\prime} P\right] \\
&=\left[s I-A-Q C_{2}^{\prime} C_{2}-L C_{1}-E E^{\prime} P-L D_{1} E^{\prime} P-Q A^{\prime} P-s Q P\right] \\
&=\left[s I-A-Q C_{2}^{\prime} C_{2}-L C_{1}-\left(E_{\mathrm{Q}} E_{\mathrm{Q}}^{\prime}-A Q-Q A^{\prime}-Q C_{2}^{\prime} C_{2} Q\right) P\right. \\
&=\left[s I-A-Q C_{2}^{\prime} C_{2}-L C_{1}-s Q P+A Q P+Q C_{2}^{\prime} C_{2} Q P-E_{\mathrm{Q}} E_{\mathrm{Q}}^{\prime} L D_{1} E^{\prime} P\right] \\
&=\left[\left(s I-A-Q C_{2}^{\prime} C_{2}-L C_{1}\right)(I-Q P)-\left(E_{\mathrm{Q}}+L D_{1 \mathrm{PQ}}\right) E_{\mathrm{Q}}^{\prime} P\right] \\
&=\left[s I-A-Q C_{2}^{\prime} C_{2}-L C_{1}\right] \\
& \quad\left[(I-Q P)-\left(s I-A-Q C_{2}^{\prime} C_{2}-L C_{1}\right)^{-1}\left(E_{\mathrm{Q}}+L D_{1 \mathrm{PQ}}\right) E_{\mathrm{Q}}^{\prime} P\right] \\
& \rightarrow\left[s I-A-Q C_{2}^{\prime} C_{2}-L C_{1}\right](I-Q P), \quad \text { pointwise in } s \text { as } \varepsilon \rightarrow 0 .
\end{align*}
$$

Hence, $A_{\mathrm{PQ}}+(I-Q P)^{-1} K C_{1 \mathrm{P}}$ is asymptotically stable for sufficiently small $\varepsilon$.
Now it follows from (6.4.16) that

$$
\begin{aligned}
{[s I} & \left.-A_{\mathrm{PQ}}-(I-Q P)^{-1} L C_{1 \mathrm{P}}\right]^{-1}\left[E_{\mathrm{PQ}}+(I-Q P)^{-1} L D_{1 \mathrm{PQ}}\right] \\
& \rightarrow(I-Q P)^{-1}\left[s I-A-Q C_{2}^{\prime} C_{2}-L C_{1}\right]^{-1}(I-Q P)\left[E_{\mathrm{PQ}}+(I-Q P)^{-1} L D_{1 \mathrm{PQ}}\right] \\
& =(I-Q P)^{-1}\left[s I-A-Q C_{2}^{\prime} C_{2}-L C_{1}\right]^{-1}\left[E_{\mathrm{Q}}+L D_{1 \mathrm{PQ}}\right] \\
& \rightarrow 0, \text { pointwise in } s \text { as } \varepsilon \rightarrow 0 .
\end{aligned}
$$

This completes the proof of Theorem 6.2.1.

### 6.4.D. Proof of Theorem 6.2.2

As in the previous proofs, for simplicity, we will assume that $\gamma=1$ and let $F=F\left(\gamma, \varepsilon, \Lambda_{d \mathrm{P}}, \Delta_{c \mathrm{P}}\right)$ and $K=K\left(\gamma, \varepsilon, \Lambda_{d \mathrm{Q}}, \Delta_{c Q}\right)$. Then the closed loop system $T_{h w}(s)$ is given by

$$
\left[\begin{array}{ll}
C_{2} & D_{2} F
\end{array}\right]\left(s I-\left[\begin{array}{cc}
A & B F \\
-(I-Q P)^{-1} K C_{1} & A_{\mathrm{cmp}}
\end{array}\right]\right)^{-1}\left[\begin{array}{c}
E \\
-(I-Q P)^{-1} K D_{1}
\end{array}\right]
$$

It follows from the proof of Theorem 6.1.2 that for any

$$
\alpha \in \lambda\left(A_{a a \mathrm{P}}^{-}\right) \cup \lambda\left(A_{c c \mathrm{P}}^{c}\right) \subseteq \lambda(A+B F)
$$

the corresponding right eigenvector, say $W$, i.e., $(A+B F) W=\alpha W$, satisfies $\left(C_{2}+D_{2} F\right) W=0$. Moreover, it is simple to verify that $\left(C_{2 \mathrm{P}}+D_{2 \mathrm{P}} F\right) W=0$ and $P W=0$.

By duality, one can show that for any $\beta \in \lambda\left(A_{a a Q}^{-}\right) \cup \lambda\left(A_{c c Q}^{c}\right), \beta \in \lambda\left(A+K C_{1}\right)$ and the corresponding left eigenvector, say $V$, i.e., $V^{\mathrm{H}}\left(A+K C_{1}\right)=\beta V^{\mathrm{H}}$, satisfies $V^{\mathrm{H}}\left(E+K D_{1}\right)=0$ and $V^{\mathrm{H}} Q=0$. In view of (6.2.6), we have

$$
\begin{aligned}
A_{\mathrm{cmp}} W= & {\left[A+E E^{\prime} P+B F+(I-Q P)^{-1} Q C_{2 \mathrm{P}}^{\prime}\left(C_{2 \mathrm{P}}+D_{2 \mathrm{P}} F\right)\right.} \\
& \left.+(I-Q P)^{-1} K C_{1}+(I-Q P)^{-1} K D_{1} E^{\prime} P\right] W \\
= & (I-Q P)^{-1} K C_{1} W+(A+B F) W
\end{aligned}
$$

and

$$
\begin{aligned}
V^{\mathrm{H}} A_{\mathrm{cmp}}= & V^{\mathrm{H}}(I-Q P)\left[A+E E^{\prime} P+B F+(I-Q P)^{-1} Q C_{2 \mathrm{P}}^{\prime}\left(C_{2 \mathrm{P}}+D_{2 \mathrm{P}} F\right)\right. \\
& \left.+(I-Q P)^{-1} K C_{1}+(I-Q P)^{-1} K D_{1} E^{\prime} P\right] \\
= & V^{\mathrm{H}} B F+V^{\mathrm{H}}\left(A+K C_{1}\right) .
\end{aligned}
$$

Therefore,
$\left[\begin{array}{cc}A & B F \\ -(I-Q P)^{-1} K C_{1} & A_{\mathrm{cmp}}\end{array}\right]\left[\begin{array}{l}W \\ W\end{array}\right]=\left[\begin{array}{c}(A+B F) W \\ A_{\mathrm{cmp}} W-(I-Q P)^{-1} K C_{1}\end{array}\right]=\alpha\left[\begin{array}{l}W \\ W\end{array}\right]$,
and

$$
\left[\begin{array}{ll}
C_{2} & D_{2} F
\end{array}\right]\left[\begin{array}{l}
W \\
W
\end{array}\right]=\left(C_{2}+D_{2} F\right) W=0
$$

This shows that $\alpha$ is an output decoupling zero of $T_{h w}(s)$. Similarly,

$$
\begin{aligned}
& {\left[\begin{array}{ll}
V^{\mathrm{H}} & -V^{\mathrm{H}}
\end{array}\right]\left[\begin{array}{cc}
A & B F \\
-(I-Q P)^{-1} K C_{1} & A_{\mathrm{cmp}}
\end{array}\right]} \\
& =\left[V^{\mathrm{H}}(I-Q P)\left[A+(I-Q P)^{-1} K C_{1}\right] \quad V^{\mathrm{H}}\left(B F-A_{\mathrm{cmp}}\right)\right] \\
& =\beta\left[\begin{array}{ll}
V^{\mathrm{H}} & -V^{\mathrm{H}}
\end{array}\right] \text {, }
\end{aligned}
$$

and

$$
\left[\begin{array}{ll}
V^{\mathrm{H}} & -V^{\mathrm{H}}
\end{array}\right]\left[\begin{array}{c}
E \\
-(I-Q P)^{-1} K D_{1}
\end{array}\right]=V^{\mathrm{H}}\left(E+K D_{1}\right)=0 .
$$

This implies that $\beta$ is an input decoupling zero of $T_{h w}(s)$.
The first part of item 3 in Theorem 6.2 .2 can be verified easily by using (6.4.6) and the fact that

$$
\operatorname{Im}(P)=\left[\mathcal{S}^{+}\left(\Sigma_{\mathbf{P}}\right)\right]^{\perp}
$$

The second part is the dual of the first case. This completes the proof of Theorem 6.2.2.

### 6.4.E. Proof of Theorem 6.3.1

First, note that the subsystem i) $\left(A_{\mathrm{PQ}}, B_{\mathrm{PQ}}, C_{2 \mathrm{P}}, D_{2 \mathrm{P}}\right)$ is right invertible and of minimum phase; and ii) the subsystem ( $A_{\mathrm{PQ}}, E_{\mathrm{PQ}}, C_{1 \mathrm{P}}, D_{1 \mathrm{PQ}}$ ) is left invertible and of minimum phase. It follows from Theorem 7.4.2 that there indeed exist gain matrices $F$ and $K_{\mathrm{R}}$ such that the resulting reduced order output feedback control law (6.3.10) internally stabilizes $\Sigma_{\mathrm{PQ}}$ and makes the $H_{\infty}$ norm of the closed-loop transfer matrix strictly less than any given $\varepsilon$. The second result of Theorem 6.3.1 follows from Lemma 6.4.3.

## Chapter 7

## Continuous-time $H_{\infty}$ Almost Disturbance Decoupling

### 7.1. Introduction

We CONSIDER IN this chapter the problem of $H_{\infty}$ almost disturbance decoupling with measurement feedback and internal stability for continuous-time linear systems. Although in principle it is a special case of the general $H_{\infty}$ control problem, i.e., the case that $\gamma^{*}=0$, the problem of almost disturbance decoupling has a vast history behind it, occupying a central part of classical as well as modern control theory. Several important problems, such as robust control, decentralized control, non-interactive control, model reference or tracking control, $H_{2}$ and $H_{\infty}$ optimal control problems can all be recast into an almost disturbance decoupling problem. Roughly speaking, the basic almost disturbance decoupling problem is to find an output feedback control law such that in the closed-loop system the disturbances are quenched, say in an $\mathcal{L}_{p}$ sense, up to any pre-specified degree of accuracy while maintaining internal stability. Such a problem was originally formulated by Willems ([111] and [112]) and labelled ADDPMS (the almost disturbance decoupling problem with measurement feedback and internal stability). In the case that, instead of a measurement feedback, a state feedback is used, the above problem is termed ADDPS (the almost disturbance decoupling problem with internal stability). The prefix $H_{\infty}$ in the acronyms $H_{\infty}$-ADDPMS and $H_{\infty}$-ADDPS is used to specify that the degree of accuracy in disturbance quenching is measured in $\mathcal{L}_{2}$-sense.

There is extensive literature on the almost disturbance decoupling problem (See, for example, the recent work [109], [74] and [75] and the references therein). In [109], several variations of the disturbance decoupling problems and their solvability conditions are summarized, and the necessary and sufficient conditions are given, under which the $H_{\infty}$-ADDPMS and $H_{\infty}$-ADDPS for continuous-time linear systems are solvable. These conditions are given in terms of geometry subspaces and for strictly proper systems (i.e., without direct feedthrough terms from the control input to the output to be controlled and from the disturbance input to the measurement output). Under these conditions, [74] constructs feedback laws, parameterized explicitly in a single parameter $\varepsilon$, that solve the $H_{\infty}$-ADDPMS and the $H_{\infty}$-ADDPS. These results were later extended to proper systems (i.e., with direct feedthrough terms) in [75]. We emphasize that in all the results mentioned above, the internal stability was always with respect to a closed set in the complex plane. Such a closeness restriction, while facilitating the development of the the above results, excludes systems with disturbance affected purely imaginary invariant zero dynamics from consideration. Only recently was this "final" restriction on the internal stability restriction removed by Scherer [96], thus allowing purely imaginary invariant zero dynamics to be affected by the disturbance. More specifically, Scherer [96] gave a set of necessary and sufficient conditions under which the $H_{\infty}$-ADDPMS and the $H_{\infty}$-ADDPS, with internal stability being with respect to the open left-half plane, is solvable for general proper linear systems. When the stability is with respect to the open left-half plane, the $H_{\infty}$-ADDPMS and the $H_{\infty}$-ADDPS will be referred to as the general $H_{\infty}$-ADDPMS and the general $H_{\infty}$-ADDPS, respectively. The explicit construction algorithm for feedback laws that solve these general $H_{\infty}$-ADDPMS and $H_{\infty}$-ADDPS under Scherer's necessary and sufficient conditions has only appeared in a very recent paper of Chen, Lin and Hang [20]. The objective of this chapter is to present: 1) easily checkable conditions for the general $H_{\infty}$-ADDPS and $H_{\infty}$-ADDPMS; and 2) explicit algorithms to construct solutions that solve these problems. The latter were reported in Chen, Lin and Hang [20].

More specifically, we consider the general $H_{\infty}$-ADDPMS and the general $H_{\infty}$-ADDPS, for the following general continuous-time linear system,

$$
\Sigma:\left\{\begin{array}{l}
\dot{x}=A x+B u+E w  \tag{7.1.1}\\
y=C_{1} x+D_{1} w \\
h=C_{2} x+D_{2} u+D_{22} w
\end{array}\right.
$$

where $x \in \mathbb{R}^{n}$ is the state, $u \in \mathbb{R}^{m}$ is the control input, $y \in \mathbb{R}^{\ell}$ is the measurement, $w \in \mathbb{R}^{q}$ is the disturbance and $h \in \mathbb{R}^{p}$ is the output to be con-
trolled. As usual, for convenient reference in future development, throughout this chapter, we define $\Sigma_{P}$ to be the subsystem characterized by the matrix quadruple ( $A, B, C_{2}, D_{2}$ ) and $\Sigma_{\mathrm{Q}}$ to be the subsystem characterized by the matrix quadruple $\left(A, E, C_{1}, D_{1}\right)$. The following dynamic feedback control laws are investigated:

$$
\Sigma_{c}:\left\{\begin{align*}
\dot{x}_{c} & =A_{c} x_{c}+B_{c} y  \tag{7.1.2}\\
u & =C_{c} x_{c}+D_{c} y
\end{align*}\right.
$$

The controller $\Sigma_{c}$ of (7.1.2) is said to be internally stabilizing when applied to the system $\Sigma$, if the following matrix is asymptotically stable:

$$
A_{\mathrm{cl}}:=\left[\begin{array}{cc}
A+B D_{c} C_{1} & B C_{c}  \tag{7.1.3}\\
B_{c} C_{1} & A_{c}
\end{array}\right]
$$

i.e., all its eigenvalues lie in the open left-half complex plane. Denote by $T_{h w}$ the corresponding closed-loop transfer matrix from the disturbance $w$ to the output to be controlled $h$, i.e.,

$$
\begin{gather*}
T_{h w}=\left[\begin{array}{ll}
C_{2}+D_{2} D_{c} C_{1} & D_{2} C_{c}
\end{array}\right]\left(s I-\left[\begin{array}{cc}
A+B D_{c} C_{1} & B C_{c} \\
B_{c} C_{1} & A_{c}
\end{array}\right]\right)^{-1}\left[\begin{array}{c}
E+B D_{c} D_{1} \\
B_{c} D_{1}
\end{array}\right] \\
+\left(D_{2} D_{c} D_{1}+D_{22}\right) \tag{7.1.4}
\end{gather*}
$$

Then the general $H_{\infty}$-ADDPMS and the general $H_{\infty}$-ADDPS can be formally defined as follows.

Definition 7.1.1. The $H_{\infty}$ almost disturbance decoupling problem with measurement feedback and with internal stability (the $H_{\infty}$-ADDPMS) for the continuous time system (7.1.1) is said to be solvable if, for any given positive scalar $\gamma>0$, there exists at least one controller of the form (7.1.2) such that,

1. in the absence of disturbance, the closed-loop system comprising the system (7.1.1) and the controller (7.1.2) is asymptotically stable, i.e., the matrix $A_{\mathrm{cl}}$ as given by (7.1.3) is asymptotically stable; and
2. the closed-loop system has an $\mathcal{L}_{2}$-gain, from the disturbance $w$ to the controlled output $h$, that is less than or equal to $\gamma$, i.e.,

$$
\begin{equation*}
\|h\|_{2} \leq \gamma\|w\|_{2}, \quad \forall w \in \mathcal{L}_{2} \text { and for }\left(x(0), x_{c}(0)\right)=(0,0) \tag{7.1.5}
\end{equation*}
$$

Equivalently, the $H_{\infty}$-norm of the closed-loop transfer matrix from $w$ to $h, T_{h w}$, is less than or equal to $\gamma$, i.e., $\left\|T_{h w}\right\|_{\infty} \leq \gamma$.

In the case that $C_{1}=I$ and $D_{1}=0$, the general $H_{\infty}$-ADDPMS as defined above becomes the general $H_{\infty}$-ADDPS, where only a static state feedback, instead the dynamic output feedback (7.1.2) is necessary.

Clearly, the $H_{\infty}$-ADDPMS for $\Sigma$ of (7.1.1) is equivalent to the general $H_{\infty}$ control problem for $\Sigma$ with $\gamma^{*}=0$. As stated earlier, one of the objectives of this chapter is to construct families of feedback laws of the form (7.1.2), parameterized in a single parameter, say $\varepsilon$, that, under the necessary and sufficient conditions of Scherer [96], solve the above defined general $H_{\infty}$-ADDPMS and $H_{\infty}$-ADDPS for general systems whose subsystems $\Sigma_{\mathrm{P}}$ and $\Sigma_{\mathrm{Q}}$ may have invariant zeros on the imaginary axis. The feedback laws we are to construct are observer-based. A family of static state feedback laws parameterized in a single parameter is first constructed to solve the general $H_{\infty}$-ADDPS. A class of observers parameterized in the same parameter $\varepsilon$ is then constructed to implement the state feedback laws and thus obtain a family of dynamic measurement feedback laws parameterized in a single parameter $\varepsilon$ that solve the general $H_{\infty}$-ADDPMS. The basic tools we use in the construction of such families of feedback laws are: 1) the special coordinate basis, developed by Sannuti and Saberi [93] and Saberi and Sannuti [89] (see also Chapter 2), in which a linear system is decomposed into several subsystems corresponding to its finite and infinite zero structures as well as its invertibility structures; 2) a block diagonal controllability canonical form (see also Chapter 2) that puts the dynamics of imaginary invariant zeros into a special canonical form under which the lowgain design technique can be applied; and 3) the $H_{\infty}$ low-and-high gain design technique. The development of such an $H_{\infty}$ low-and-high gain design technique was originated in [62] and [63] in the context of $H_{\infty}$-ADDPMS for special classes of nonlinear systems that specialized to a SISO (and hence square invertible) linear system having no invariant zero in the open right-half plane.

### 7.2. Solvability Conditions

In this section, we first recall the necessary and sufficient conditions of Scherer [96] under which the general $H_{\infty}$-ADDPMS and $H_{\infty}$-ADDPS are solvable. Then we will convert the geometric conditions of Scherer into easily checkable ones using the properties of the special coordinate basis. The following result is a slight generalization of Scherer [96].

Theorem 7.2.1. Consider the general measurement feedback system (7.1.1). Then the general $H_{\infty}$ almost disturbance decoupling problem for (7.1.1) with internal stability ( $H_{\infty}$-ADDPMS) is solvable, if and only if the following conditions are satisfied:

1. $(A, B)$ is stabilizable;
2. $\left(A, C_{1}\right)$ is detectable;
3. $D_{22}+D_{2} S D_{1}=0$, where $S=-\left(D_{2}^{\prime} D_{2}\right)^{\dagger} D_{2}^{\prime} D_{22} D_{1}^{\prime}\left(D_{1} D_{1}^{\prime}\right)^{\dagger}$;
4. $\operatorname{Im}\left(E+B S D_{1}\right) \subset \mathcal{S}^{+}\left(\Sigma_{\mathbf{P}}\right) \cap\left\{\cap_{\lambda \in \mathbf{C}^{0}} \mathcal{S}_{\lambda}\left(\Sigma_{\mathbf{P}}\right)\right\}$;
5. $\operatorname{Ker}\left(C_{2}+D_{2} S C_{1}\right) \supset \mathcal{V}^{+}\left(\Sigma_{Q}\right) \cup\left\{\cup_{\lambda \in \mathbf{C}^{0}} \mathcal{V}_{\lambda}\left(\Sigma_{\mathbf{Q}}\right)\right\} ;$ and
6. $\mathcal{V}^{+}\left(\Sigma_{Q}\right) \subset \mathcal{S}^{+}\left(\Sigma_{P}\right)$.

T
Remark 7.2.1. Note that if $\Sigma_{P}$ is right invertible and of minimum phase, and $\Sigma_{Q}$ is left invertible and of minimum phase, then Conditions 4 to 6 of Theorem 7.2.1 are automatically satisfied. Hence, the solvability conditions of the $H_{\infty}$-ADDPMS for such a case reduce to:

1. $(A, B)$ is stabilizable;
2. $\left(A, C_{1}\right)$ is detectable; and
3. $D_{22}+D_{2} S D_{1}=0$, where $S=-\left(D_{2}^{\prime} D_{2}\right)^{\dagger} D_{2}^{\prime} D_{22} D_{1}^{\prime}\left(D_{1} D_{1}^{\prime}\right)^{\dagger}$.

Remark 7.2.2. It is simple to verify that for the case when all states of the system (7.1.1) are fully measurable, i.e., $C_{1}=I$ and $D_{1}=0$, then the solvability conditions for the general $H_{\infty}$-ADDPS reduce to the following:

1. $(A, B)$ is stabilizable;
2. $D_{22}=0$; and
3. $\operatorname{Im}(E) \subset \mathcal{S}^{+}\left(\Sigma_{\mathrm{P}}\right) \cap\left\{\cap_{\lambda \in \mathbb{C}^{0}} \mathcal{S}_{\lambda}\left(\Sigma_{\mathrm{P}}\right)\right\}$.

Moreover, in this case, a static state feedback law, i.e., $u=F x$, where $F$ is a constant matrix and might be parameterized by certain tuning parameters, exists that solves the general $H_{\infty}$-ADDPS.

Theorem 7.2.1 is quite elegant as it is expressed in terms of the well-known geometric conditions. However, it might be hard to verify these geometric conditions numerically. In what follows, we will present a simple method to check the solvability conditions for the $H_{\infty}$-ADDPMS for general continuoustime systems.

Step 7.2.0: Let $S=-\left(D_{2}^{\prime} D_{2}\right)^{\dagger} D_{2}^{\prime} D_{22} D_{1}^{\prime}\left(D_{1} D_{1}^{\prime}\right)^{\dagger}$. If $D_{22}+D_{2} S D_{1} \neq 0$, the algorithm stops here. Otherwise, go to Step 7.2.1.

Step 7.2.1: Compute the special coordinate basis of $\Sigma_{\mathbf{P}}$, i.e., the quadruple ( $A, B, C_{2}, D_{2}$ ). For easy reference, we append a subscript ' p ' to all submatrices and transformations in the SCB associated with $\Sigma_{\mathrm{P}}$, e.g., $\Gamma_{s \mathrm{P}}$ is the state transformation of the SCB of $\Sigma_{\mathrm{P}}$, and $A_{a a \mathrm{P}}^{0}$ is associated with invariant zero dynamics of $\Sigma_{\mathrm{P}}$ on the imaginary axis.

Step 7.2.2: Next, we denote the set of eigenvalues of $A_{a a \mathrm{P}}^{0}$ with a non-negative imaginary part as $\left\{\omega_{\mathrm{P} 1}, \omega_{\mathrm{P} 2}, \cdots, \omega_{\mathrm{P}} k_{\mathrm{P}}\right\}$ and for $i=1,2, \cdots, k_{\mathrm{P}}$, choose complex matrices $V_{i \mathrm{P}}$, whose columns form a basis for the eigenspace

$$
\begin{equation*}
\left\{x \in \mathbb{C}^{n_{a \mathrm{P}}^{0}} \mid x^{\mathrm{H}}\left(\omega_{\mathrm{P} i} I-A_{a a \mathrm{P}}^{0}\right)=0\right\} \tag{7.2.1}
\end{equation*}
$$

where $n_{a \mathrm{p}}^{0}$ is the dimension of $\mathcal{X}_{a \mathrm{p}}^{0}$. Then, let

$$
V_{\mathrm{P}}:=\left[\begin{array}{llll}
V_{1 \mathrm{P}} & V_{2 \mathrm{P}} & \cdots & V_{k \mathrm{P}} \tag{7.2.2}
\end{array}\right] .
$$

We also compute $n_{x \mathrm{P}}:=\operatorname{dim}\left(\mathcal{X}_{a \mathrm{P}}^{+}\right)+\operatorname{dim}\left(\mathcal{X}_{b \mathrm{P}}\right)$, and

$$
\Gamma_{s \mathrm{P}}^{-1}\left(E+B S D_{1}\right):=\left[\begin{array}{c}
E_{a \mathrm{P}}^{-}  \tag{7.2.3}\\
E_{a \mathrm{P}}^{0} \\
E_{a \mathrm{P}}^{+} \\
E_{\mathrm{bP}} \\
E_{\mathrm{cP}} \\
E_{d \mathrm{P}}
\end{array}\right]
$$

Step 7.2.3: Let $\Sigma_{\mathrm{Q}}^{\star}$ be the dual system of $\Sigma_{\mathbf{Q}}$ and be characterized by a quadruple $\left(A^{\prime}, C_{1}^{\prime}, E^{\prime}, D_{1}^{\prime}\right)$. We compute the special coordinate basis of $\Sigma_{\mathrm{Q}}^{\star}$. Again, for easy reference, we append a subscript ' $Q$ ' to all sub-matrices and transformations in the SCB associated with $\Sigma_{Q}^{\star}$, e.g., $\Gamma_{s Q}$ is the state transformation of the SCB of $\Sigma_{Q}^{\star}$, and $A_{a a Q}^{0}$ is associated with invariant zero dynamics of $\Sigma_{\mathrm{Q}}^{\star}$ on the imaginary axis.

Step 7.2.4: Next, denote the set of eigenvalues of $A_{a a Q}^{0}$ with a non-negative imaginary part as $\left\{\omega_{\mathrm{Q} 1}, \omega_{\mathrm{Q} 2}, \cdots, \omega_{\mathrm{Q} k_{\mathrm{Q}}}\right\}$ and for $i=1,2, \cdots, k_{\mathrm{Q}}$, choose complex matrices $V_{i Q}$, whose columns form a basis for the eigenspace

$$
\begin{equation*}
\left\{x \in \mathbb{C}^{n_{a Q}^{0}} \mid x^{\mathrm{H}}\left(\omega_{Q i} I-A_{a a Q}^{0}\right)=0\right\}, \tag{7.2.4}
\end{equation*}
$$

where $n_{a Q}^{0}$ is the dimension of $\mathcal{X}_{a Q}^{0}$. Then, let

$$
V_{\mathrm{Q}}:=\left[\begin{array}{llll}
V_{1 \mathrm{Q}} & V_{2 \mathrm{Q}} & \cdots & V_{k_{\mathrm{Q}}} \tag{7.2.5}
\end{array}\right] .
$$

We next compute $n_{x Q}:=\operatorname{dim}\left(\mathcal{X}_{a Q}^{+}\right)+\operatorname{dim}\left(\mathcal{X}_{b Q}\right)$, and

$$
\Gamma_{s Q}^{-1}\left(C_{2}+D_{2} S C_{1}\right)^{\prime}:=\left[\begin{array}{c}
E_{a Q}^{-}  \tag{7.2.6}\\
E_{a Q}^{0} \\
E_{a Q}^{+} \\
E_{b Q} \\
E_{c Q} \\
E_{d Q}
\end{array}\right] \text {. }
$$

Step 7.2.5: Finally, compute

$$
\Gamma_{s \mathrm{P}}^{-1}\left(\Gamma_{s Q}^{-1}\right)^{\prime}=\left[\begin{array}{ccc}
X_{a}^{-0} & \star & \star  \tag{7.2.7}\\
\star & \Gamma & \star \\
\star & \star & X_{c d}
\end{array}\right]
$$

where $X_{a}^{-0}$ and $X_{c d}$ are of dimensions $\left(n_{a \mathrm{P}}^{-}+n_{a \mathrm{P}}^{0}\right) \times\left(n_{a_{\mathrm{Q}}}^{-}+n_{a \mathrm{Q}}^{0}\right)$ and $\left(n_{c \mathrm{P}}+n_{d \mathrm{P}}\right) \times\left(n_{c Q}+n_{d Q}\right)$, respectively, and finally $\Gamma$ is a sub-matrix of dimension $n_{x \mathrm{P}} \times n_{x \mathrm{Q}}$.

囚
We have the following proposition.
Proposition 7.2.1. Consider the general measurement feedback continuoustime system (7.1.1). Then the $H_{\infty}$ almost disturbance decoupling problem for (7.1.1) with internal stability ( $H_{\infty}$-ADDPMS) is solvable, i.e., $\gamma^{*}=0$, if and only if the following conditions are satisfied:

1. $(A, B)$ is stabilizable;
2. $\left(A, C_{1}\right)$ is detectable;
3. $D_{22}+D_{2} S D_{1}=0$, where $S=-\left(D_{2}^{\prime} D_{2}\right)^{\dagger} D_{2}^{\prime} D_{22} D_{1}^{\prime}\left(D_{1} D_{1}^{\prime}\right)^{\dagger}$;
4. $E_{a \mathrm{P}}^{+}=0, E_{b \mathrm{P}}=0$ and $V_{\mathrm{P}}^{\mathrm{H}} E_{a \mathrm{P}}^{0}=0$;
5. $E_{a Q}^{+}=0, E_{b Q}=0$ and $V_{Q}^{\mathrm{H}} E_{a \mathrm{Q}}^{0}=0$; and
6. $\Gamma=0$.

Proof. It is simple to see that the first three conditions are necessary for the $H_{\infty}$-ADDPMS for (7.1.1) to be solvable. Next, it follows trivially from the properties of the special coordinate basis that the geometric condition, $\operatorname{Im}\left(E+B S D_{1}\right) \subset \mathcal{S}^{+}\left(\Sigma_{\mathrm{P}}\right) \cap\left\{\cap_{\lambda \in \mathbb{C}^{0}} \mathcal{S}_{\lambda}\left(\Sigma_{\mathrm{P}}\right)\right\}$, is equivalent to $E_{a \mathrm{P}}^{+}=0, E_{b \mathrm{P}}=$ 0 and $V_{\mathrm{P}}^{\mathrm{H}} E_{a \mathrm{P}}^{0}=0$. Dually, the geometric condition, $\operatorname{Ker}\left(C_{2}+D_{2} S C_{1}\right) \supset$ $\mathcal{V}^{+}\left(\Sigma_{\mathrm{Q}}\right) \cup\left\{\cup_{\lambda \in \mathbb{C}^{0}} \mathcal{V}_{\lambda}\left(\Sigma_{\mathrm{Q}}\right)\right\}$, is equivalent to $E_{a \mathrm{Q}}^{+}=0, E_{b \mathrm{Q}}=0$ and $V_{\mathrm{Q}}^{\mathrm{H}} E_{a \mathrm{Q}}^{0}=0$. Next, again following the properties of the special coordinate basis, we have

$$
\mathcal{S}^{+}\left(\Sigma_{\mathbf{P}}\right)=\operatorname{Ker}\left\{\left[\begin{array}{lll}
0 & I_{n_{x P}} & 0
\end{array}\right] \Gamma_{s \mathrm{P}}^{-1}\right\}, \quad \mathcal{V}^{+}\left(\Sigma_{Q}\right)=\operatorname{Im}\left\{\left(\Gamma_{s Q}^{-1}\right)^{\prime}\left[\begin{array}{c}
0 \\
I_{n_{x Q}} \\
0
\end{array}\right]\right\} .
$$

Hence, it is straightforward to verify that $\mathcal{V}^{+}\left(\Sigma_{\mathbf{Q}}\right) \subset \mathcal{S}^{+}\left(\Sigma_{\mathbf{P}}\right)$ is equivalent to

$$
\left[\begin{array}{lll}
0 & I_{n_{x} \mathrm{P}} & 0
\end{array}\right] \Gamma_{s \mathrm{P}}^{-1}\left(\Gamma_{s Q}^{-1}\right)^{\prime}\left[\begin{array}{c}
0 \\
I_{n_{x Q}} \\
0
\end{array}\right]=\Gamma=0 .
$$

Thus, the result follows.

### 7.3. Solutions to Full State Feedback Case

In this section, we consider feedback law design for the general $H_{\infty}$ almost disturbance decoupling problem with internal stability and with full state feedback, where internal stability is with respect to the open left-half plane, i.e., the general $H_{\infty}$-ADDPS. More specifically, we present a design procedure that constructs a family of parameterized static state feedback laws,

$$
\begin{equation*}
u=F(\varepsilon) x \tag{7.3.1}
\end{equation*}
$$

that solves the general $H_{\infty}$ - ADDPS for the following system,

$$
\left\{\begin{array}{l}
\dot{x}=A x+B u+E w  \tag{7.3.2}\\
y=x \\
h=C_{2} x+D_{2} u+D_{22} w
\end{array}\right.
$$

That is, under this family of state feedback laws, the resulting closed-loop system is asymptotically stable for sufficiently small $\varepsilon$ and the $H_{\infty}$-norm of the closed-loop transfer matrix from $w$ to $h, T_{h w}(s, \varepsilon)$, tends to zero as $\varepsilon$ tends to zero, where

$$
\begin{equation*}
T_{h w}(s, \varepsilon)=\left[C_{2}+D_{2} F(\varepsilon)\right][s I-A-B F(\varepsilon)]^{-1} E+D_{22} \tag{7.3.3}
\end{equation*}
$$

Clearly, $D_{22}=0$ is a necessary condition for the solvability of the general $H_{\infty}$-ADDPS. We present an algorithm for obtaining this $F(\varepsilon)$, following the asymptotic time-scale and eigenstructure assignment (ATEA) procedure. We first use the special coordinate basis of the given system (See Theorem 2.3.1) to decompose the system into several subsystems according to its finite and infinite zero structures as well as its invertibility structures. The new component here is the low-gain design for the part of the zero dynamics corresponding to all purely imaginary invariant zeros. As will be clear shortly, the low-gain component is critical in handling the case when the zero dynamics corresponding to purely imaginary invariant zeros is affected by disturbance. It is well-known that the disturbance affected purely imaginary zero dynamics is difficult to handle and has always been excluded from consideration until recently.

We have in the following a step-by-step algorithm.
Step 7.S.1: (Decomposition of $\Sigma_{\mathrm{P}}$ ). Transform the subsystem $\Sigma_{\mathbf{P}}$, i.e., the quadruple ( $A, B, C_{2}, D_{2}$ ), into the special coordinate basis (SCB) as given by Theorem 2.3.1 of Chapter 2. Denote the state, output and input transformation matrices as $\Gamma_{s \mathrm{P}}, \Gamma_{o \mathrm{P}}$ and $\Gamma_{i \mathrm{P}}$, respectively.

Step 7.S.2: (Gain matrix for the subsystem associated with $\mathcal{X}_{c}$ ). Let $F_{c}$ be any arbitrary $m_{c} \times n_{c}$ matrix subject to the constraint that

$$
\begin{equation*}
A_{c c}^{c}=A_{c c}-B_{c} F_{c}, \tag{7.3.4}
\end{equation*}
$$

is a stable matrix. Note that the existence of such an $F_{c}$ is guaranteed by the property of SCB, i.e., $\left(A_{c c}, B_{c}\right)$ is controllable.

Step 7.S.3: (Gain matrix for the subsystems associated with $\mathcal{X}_{a}^{+}$and $\mathcal{X}_{b}$ ). Let

$$
F_{a b}^{+}:=\left[\begin{array}{cc}
F_{a 0}^{+} & F_{b 0}  \tag{7.3.5}\\
F_{a d}^{+} & F_{b d}
\end{array}\right]
$$

be any arbitrary $\left(m_{0}+m_{d}\right) \times\left(n_{a}^{+}+n_{b}\right)$ matrix subject to the constraint that

$$
A_{a b}^{+c}:=\left[\begin{array}{cc}
A_{a a}^{+} & L_{a b}^{+} C_{b}  \tag{7.3.6}\\
0 & A_{b b}
\end{array}\right]-\left[\begin{array}{cc}
B_{0 a}^{+} & L_{a d}^{+} \\
B_{0 b} & L_{a b}
\end{array}\right] F_{a b}^{+}
$$

is a stable matrix. Again, note that the existence of such an $F_{a b}$ is guaranteed by the stabilizability of $(A, B)$ and Property 2.3 .1 of the special coordinate basis. For future use, let us partition [ $\left.\begin{array}{ll}F_{a d}^{+} & F_{b d}\end{array}\right]$ as,

$$
\left[\begin{array}{ll}
F_{a d}^{+} & F_{b d}
\end{array}\right]=\left[\begin{array}{cc}
F_{a d 1}^{+} & F_{b d 1}  \tag{7.3.7}\\
F_{a d 2}^{+} & F_{b d 2} \\
\vdots & \vdots \\
F_{a d m_{d}}^{+} & F_{b d m_{d}}
\end{array}\right]
$$

where $F_{a d i}^{+}$and $F_{b d i}$ are of dimensions $1 \times n_{a}^{+}$and $1 \times n_{b}$, respectively.
Step 7.S.4: (Gain matrix for the subsystem associated with $\mathcal{X}_{a}^{0}$ ). The construction of this gain matrix is carried out in the following sub-steps.

Step 7.S.4.1: (Preliminary coordinate transformation). Recalling the definition of ( $A_{\text {con }}, B_{\text {con }}$ ), i.e., (2.3.27), we have

$$
A_{\mathrm{con}}-B_{\mathrm{con}} F_{a b}^{+}=\left[\begin{array}{ccc}
A_{a a}^{-} & 0 & A_{a a b}^{-}  \tag{7.3.8}\\
0 & A_{a a}^{0} & A_{a a b}^{0} \\
0 & 0 & A_{a b}^{+c}
\end{array}\right], \quad B_{\mathrm{con}}=\left[\begin{array}{cc}
B_{0 a}^{-} & L_{a d}^{-} \\
B_{0 a}^{0} & L_{a d}^{0} \\
B_{0 a b}^{+} & L_{a b d}^{+}
\end{array}\right],
$$

where

$$
\begin{gather*}
B_{0 a b}^{+}=\left[\begin{array}{c}
B_{0 a}^{+} \\
B_{0 b}
\end{array}\right], \quad L_{a b d}^{+}=\left[\begin{array}{c}
L_{a d}^{+} \\
L_{b d}
\end{array}\right],  \tag{7.3.9}\\
A_{a a b}^{0}=\left[\begin{array}{ll}
0 & L_{a b}^{0} C_{b}
\end{array}\right]-\left[\begin{array}{ll}
B_{0 a}^{0} & L_{a d}^{0}
\end{array}\right] F_{a b}^{+}, \tag{7.3.10}
\end{gather*}
$$

and

$$
A_{a a b}^{-}=\left[\begin{array}{ll}
0 & L_{a b}^{-} C_{b}
\end{array}\right]-\left[\begin{array}{ll}
B_{0 a}^{-} & L_{a d}^{-} \tag{7.3.11}
\end{array}\right] F_{a b}^{+} .
$$

Clearly ( $A_{\text {con }}-B_{\text {con }} F_{a b}^{+}, B_{\text {con }}$ ) remains stabilizable. Construct the following nonsingular transformation matrix,

$$
\Gamma_{a b}=\left[\begin{array}{ccc}
I_{n_{a}^{-}} & 0 & 0  \tag{7.3.12}\\
0 & 0 & I_{n_{a}^{+}+n_{b}} \\
0 & I_{n_{a}^{0}} & T_{a}^{0}
\end{array}\right]^{-1}
$$

where $T_{a}^{0}$ is the unique solution to the following Lyapunov equation,

$$
\begin{equation*}
A_{a a}^{0} T_{a}^{0}-T_{a}^{0} A_{a b}^{+c}=A_{a a b}^{0} \tag{7.3.13}
\end{equation*}
$$

We note here that such a unique solution to the above Lyapunov equation always exists since all the eigenvalues of $A_{a a}^{0}$ are on the imaginary axis and all the eigenvalues of $A_{a b}^{+c}$ are in the open lefthalf plane. It is now easy to verify that

$$
\begin{gather*}
\Gamma_{a b}^{-1}\left(A_{\mathrm{con}}-B_{\mathrm{con}} F_{a b}^{+}\right) \Gamma_{a b}=\left[\begin{array}{ccc}
A_{a a}^{-} & A_{a a b}^{-} & 0 \\
0 & A_{a b}^{+c} & 0 \\
0 & 0 & A_{a a}^{0}
\end{array}\right],  \tag{7.3.14}\\
\Gamma_{a b}^{-\mathbf{1}} B_{\mathrm{con}}=\left[\begin{array}{cc}
B_{0 a}^{-} & L_{a d}^{-} \\
B_{0 a b}^{+} & L_{a b d}^{+} \\
B_{0 a}^{0}+T_{a}^{0} B_{0 a b}^{+} & L_{a d}^{0}+T_{a}^{0} L_{a b d}^{+}
\end{array}\right] \tag{7.3.15}
\end{gather*}
$$

Hence, the matrix pair $\left(A_{a a}^{0}, B_{a}^{0}\right)$ is controllable, where

$$
B_{a}^{0}=\left[B_{0 a}^{0}+T_{a}^{0} B_{0 a b}^{+} \quad L_{a d}^{0}+T_{a}^{0} L_{a b d}^{+}\right]
$$

Step 7.S.4.2: (Further coordinate transformation). Following the proof of Theorem 2.2.2, find nonsingular transformation matrices $\Gamma_{s a}^{0}$ and $\Gamma_{i a}^{0}$ such that $\left(A_{a a}^{0}, B_{a}^{0}\right)$ can be transformed into the block diagonal controllability canonical form,

$$
\left(\Gamma_{s a}^{0}\right)^{-1} A_{a a}^{0} \Gamma_{s a}^{0}=\left[\begin{array}{cccc}
A_{1} & 0 & \cdots & 0  \tag{7.3.16}\\
0 & A_{2} & \cdots & 0 \\
\vdots & \vdots & \ddots & \vdots \\
0 & 0 & \cdots & A_{l}
\end{array}\right]
$$

and

$$
\left(\Gamma_{s a}^{0}\right)^{-1} B_{a}^{0} \Gamma_{i a}^{0}=\left[\begin{array}{ccccc}
B_{1} & B_{12} & \cdots & B_{1 l} & \star  \tag{7.3.17}\\
0 & B_{2} & \cdots & B_{2 l} & \star \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & B_{l} & \star
\end{array}\right]
$$

where $l$ is an integer and for $i=1,2, \cdots, l$,

$$
A_{i}=\left[\begin{array}{ccccc}
0 & 1 & 0 & \cdots & 0 \\
0 & 0 & 1 & \cdots & 0 \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
0 & 0 & 0 & \cdots & 1 \\
-a_{n_{i}}^{i} & -a_{n_{i}-1}^{i} & -a_{n_{i}-2}^{i} & \cdots & -a_{1}^{i}
\end{array}\right], \quad B_{i}=\left[\begin{array}{c}
0 \\
0 \\
\vdots \\
0 \\
1
\end{array}\right] .
$$

We note that all the eigenvalues of $A_{i}$ are on the imaginary axis. Here the $\star$ 's represent sub-matrices of less interest.
Step 7.S.4.3: (Subsystem design). For each $\left(A_{i}, B_{i}\right)$, let $F_{i}(\varepsilon) \in \mathbf{R}^{1 \times n_{i}}$ be the state feedback gain such that

$$
\begin{equation*}
\lambda\left\{A_{i}+B_{i} F_{i}(\varepsilon)\right\}=-\varepsilon+\lambda\left(A_{i}\right) \in \mathbb{C}^{-} \tag{7.3.18}
\end{equation*}
$$

Note that $F_{i}(\varepsilon)$ is unique.
Step 7.S.4.4: (Composition of gain matrix for subsystem associated with $\left.\mathcal{X}_{a}^{0}\right)$. Let

$$
F_{a}^{0}(\varepsilon):=\Gamma_{i a}^{0}\left[\begin{array}{ccccc}
F_{1}(\varepsilon) & 0 & \cdots & 0 & 0  \tag{7.3.19}\\
0 & F_{2}(\varepsilon) & \cdots & 0 & 0 \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & F_{l-1}(\varepsilon) & 0 \\
0 & 0 & \cdots & 0 & F_{l}(\varepsilon) \\
0 & 0 & \cdots & 0 & 0
\end{array}\right]\left(\Gamma_{s a}^{0}\right)^{-1}
$$

where $\varepsilon \in(0,1]$ is a design parameter whose value is to be specified later.

Clearly, we have

$$
\begin{equation*}
\left\|F_{a}^{0}(\varepsilon)\right\| \leq f_{a}^{0} \varepsilon, \quad \varepsilon \in(0,1] \tag{7.3.20}
\end{equation*}
$$

for some positive constant $f_{a}^{0}$, independent of $\varepsilon$. For future use, we define and partition $F_{a b}(\varepsilon) \in \mathbb{R}^{\left(m_{0}+m_{d}\right) \times\left(n_{a}+n_{b}\right)}$ as

$$
F_{a b}(\varepsilon)=\left[\begin{array}{l}
F_{a b 0}(\varepsilon)  \tag{7.3.21}\\
F_{a b d}(\varepsilon)
\end{array}\right]=\left[\begin{array}{lll}
0_{m_{0} \times n_{a}^{-}} & 0_{m_{0} \times\left(n_{a}^{+}+n_{b}\right)} & F_{a 0}^{0}(\varepsilon) \\
0_{m_{d} \times n_{a}^{-}} & 0_{m_{d} \times\left(n_{a}^{+}+n_{b}\right)} & F_{a d}^{0}(\varepsilon)
\end{array}\right] \Gamma_{a b}^{-1}
$$

and

$$
F_{a b d}(\varepsilon)=\left[\begin{array}{c}
F_{a b d 1}(\varepsilon)  \tag{7.3.22}\\
F_{a b d 2}(\varepsilon) \\
\vdots \\
F_{a b d m_{d}}(\varepsilon)
\end{array}\right],
$$

where $F_{a 0}^{0}(\varepsilon)$ and $F_{a d}^{0}(\varepsilon)$ are defined as

$$
F_{a}^{0}(\varepsilon)=\left[\begin{array}{l}
F_{a 0}^{0}(\varepsilon)  \tag{7.3.23}\\
F_{a d}^{0}(\varepsilon)
\end{array}\right]
$$

We also partition $F_{a d}^{0}(\varepsilon)$ as,

$$
F_{a d}^{0}(\varepsilon)=\left[\begin{array}{c}
F_{a d 1}^{0}(\varepsilon)  \tag{7.3.24}\\
F_{a d 2}^{0}(\varepsilon) \\
\vdots \\
F_{a d m_{d}}^{0}(\varepsilon)
\end{array}\right]
$$

Step 7.S.5: (Gain matrix for the subsystem associated with $\mathcal{X}_{d}$ ). This step makes use of subsystems, $i=1$ to $m_{d}$, represented by (2.3.14) of Chapter 2. Let $\Lambda_{i}=\left\{\lambda_{i 1}, \lambda_{i 2}, \cdots, \lambda_{i q_{i}}\right\}, i=1$ to $m_{d}$, be the sets of $q_{i}$ elements all in $\mathbf{C}^{-}$, which are closed under complex conjugation, where $q_{i}$ and $m_{d}$ are as defined in Theorem 2.3.1 but associated with the special coordinate basis of $\Sigma_{\mathrm{p}}$. Let $\Lambda_{d}:=\Lambda_{1} \cup \Lambda_{2} \cup \cdots \cup \Lambda_{m_{d}}$. For $i=1$ to $m_{d}$, we define

$$
\begin{equation*}
p_{i}(s):=\prod_{j=1}^{q_{i}}\left(s-\lambda_{i j}\right)=s^{q_{i}}+F_{i 1} s^{q_{i}-1}+\cdots+F_{i q_{i}-1} s+F_{i q_{i}} \tag{7.3.25}
\end{equation*}
$$

and

$$
\begin{equation*}
\tilde{F}_{i}(\varepsilon):=\frac{1}{\varepsilon^{q_{i}}} F_{i} S_{i}(\varepsilon) \tag{7.3.26}
\end{equation*}
$$

where

$$
F_{i}=\left[\begin{array}{llll}
F_{i q_{i}} & F_{i q_{i}-1} & \cdots & F_{i 1} \tag{7.3.27}
\end{array}\right], \quad S_{i}(\varepsilon)=\operatorname{diag}\left\{1, \varepsilon, \varepsilon^{2}, \cdots, \varepsilon^{q_{i}-1}\right\}
$$

Step 7.S.6: (Composition of parameterized gain matrix $F(\varepsilon)$ ). In this step, various gains calculated in Steps 7.S. 2 to 7.S.5 are put together to form a composite state feedback gain matrix $F(\varepsilon)$. Let

$$
\begin{gather*}
\tilde{F}_{a b d}(\varepsilon):=\left[\begin{array}{c}
F_{a b d 1}(\varepsilon) F_{1 q_{1}} / \varepsilon^{q_{1}} \\
F_{a b d 2}(\varepsilon) F_{2 q_{2}} / \varepsilon^{q_{2}} \\
\vdots \\
F_{a b d m_{d}}(\varepsilon) F_{m_{d} q_{m_{d}}} / \varepsilon^{q_{m_{d}}}
\end{array}\right],  \tag{7.3.28}\\
\tilde{F}_{a d}^{+}(\varepsilon):=\left[\begin{array}{c}
F_{a d 1}^{+} F_{1 q_{1}} / \varepsilon^{q_{1}} \\
F_{a d 2}^{+} F_{2 q_{2}} / \varepsilon^{q_{2}} \\
\vdots \\
F_{a d m_{d}}^{+} F_{m_{d} q_{m_{d}}} / \varepsilon^{q_{m_{d}}}
\end{array}\right], \tag{7.3.29}
\end{gather*}
$$

and

$$
\tilde{F}_{b d}(\varepsilon):=\left[\begin{array}{c}
F_{b d 1} F_{1 q_{1}} / \varepsilon^{q_{1}}  \tag{7.3.30}\\
F_{b d 2} F_{2 q_{2}} / \varepsilon^{q_{2}} \\
\vdots \\
F_{b d m_{d}} F_{m_{d} q_{m_{d}}} / \varepsilon^{q_{m_{d}}}
\end{array}\right] .
$$

Then define the state feedback gain $F(\varepsilon)$ as

$$
\begin{equation*}
F(\varepsilon):=-\Gamma_{i \mathrm{P}}\left(\tilde{F}_{a b c d}^{\star}(\varepsilon)+\tilde{F}_{a b c d}(\varepsilon)\right) \Gamma_{s \mathrm{P}}^{-1} \tag{7.3.31}
\end{equation*}
$$

where

$$
\begin{gather*}
\tilde{F}_{a b c d}^{\star}(\varepsilon)=\left[\begin{array}{cccccc}
C_{0 a}^{-} & C_{0 a}^{0} & C_{0 a}^{+}+F_{a 0}^{+} & C_{0 b}+F_{b 0} & C_{0 c} & C_{0 d} \\
E_{d a}^{-} & E_{d a}^{0} & E_{d a}^{+}+\tilde{F}_{a d}^{+}(\varepsilon) & E_{d b}+\tilde{F}_{b d}(\varepsilon) & E_{d c} & \tilde{F}_{d}(\varepsilon)+E_{d} \\
E_{c a}^{-} & E_{c a}^{0} & E_{c a}^{+} & 0 & F_{c} & 0
\end{array}\right], \\
\tilde{F}_{a b c d}(\varepsilon)=\left[\begin{array}{ccc}
F_{a b 0}(\varepsilon) & 0 & 0 \\
\tilde{F}_{a b d}(\varepsilon) & 0 & 0 \\
0 & 0 & 0
\end{array}\right], \tag{7.3.32}
\end{gather*}
$$

and where

$$
\begin{gather*}
E_{d}=\left[\begin{array}{ccc}
E_{11} & \cdots & E_{1 m_{d}} \\
\vdots & \ddots & \vdots \\
E_{m_{d} 1} & \cdots & E_{m_{d} m_{d}}
\end{array}\right],  \tag{7.3.34}\\
\tilde{F}_{d}(\varepsilon)=\operatorname{diag}\left\{\tilde{F}_{1}(\varepsilon), \tilde{F}_{2}(\varepsilon), \cdots, \tilde{F}_{m_{d}}(\varepsilon)\right\} . \tag{7.3.35}
\end{gather*}
$$

We have the following theorem.
Theorem 7.3.1. Consider the given system (7.3.2) that satisfies all the conditions in Remark 7.2.2. Then the closed-loop system comprising (7.3.2) and the static state feedback law $u=F(\varepsilon) x$, with $F(\varepsilon)$ given by (7.3.31), has the following properties: For any given $\gamma>0$, there exists a positive scalar $\varepsilon^{*}>0$ such that for all $0<\varepsilon \leq \varepsilon^{*}$,

1. the closed-loop system is asymptotically stable, i.e., $\lambda\{A+B F(\varepsilon)\} \subset \mathbb{C}^{-}$;
2. the $H_{\infty}$-norm of the closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$ is less than $\gamma$, i.e., $\left\|T_{h w}(s, \varepsilon)\right\|_{\infty}<\gamma$.

Hence, by Definition 7.1.1, the control law $u=F(\varepsilon) x$ solves the general $H_{\infty^{-}}$ ADDPS for the given system (7.3.2).

Proof. See Subsection 7.5.A.

We illustrate the above result in the following example.

Example 7.3.1. Let us consider a given system of (7.1.1) characterized by $C_{1}=I, D_{1}=0$ and

$$
\begin{align*}
A & =\left[\begin{array}{rrrrrr}
0 & 1 & 0 & 0 & 0 & 1 \\
0 & 0 & 1 & 0 & 0 & 2 \\
0 & 0 & 0 & 0 & 0 & 3 \\
0 & 0 & 0 & -1 & 3 & 4 \\
0 & 0 & 0 & 0 & 4 & 5 \\
1 & 2 & 3 & 4 & 5 & 6
\end{array}\right], \quad B=\left[\begin{array}{ll}
0 & 0 \\
0 & 0 \\
0 & 0 \\
0 & 0 \\
1 & 0 \\
0 & 1
\end{array}\right], \quad E=\left[\begin{array}{ll}
1 & 0 \\
2 & 0 \\
0 & 0 \\
0 & 0 \\
0 & 0 \\
3 & 1
\end{array}\right],  \tag{7.3.36}\\
C_{2} & =\left[\begin{array}{llllll}
0 & 0 & 0 & 0 & 1 & 2 \\
0 & 0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 1 & 0
\end{array}\right], \quad D_{2}=\left[\begin{array}{ll}
1 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right], \quad D_{22}=\left[\begin{array}{ll}
0 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right] . \tag{7.3.37}
\end{align*}
$$

The subsystem $\Sigma_{\mathbf{P}}$ is already in the form of the special coordinate basis. It is simple to verify that: i) $(A, B)$ is stabilizable; ii) $\Sigma_{\mathrm{P}}$ has three invariant zeros at 0 and one stable invariant zero at -1 ; iii) $\Sigma_{\mathrm{P}}$ has one infinite zero of order zero and one infinite zero of order one; iv) $\Sigma_{\mathbf{P}}$ is left invertible; and v)

$$
\mathcal{S}^{+}\left(\Sigma_{\mathrm{P}}\right)=\operatorname{Im}\left\{\left[\begin{array}{lllll}
1 & 0 & 0 & 0 & 0  \tag{7.3.38}\\
0 & 1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1 & 0 \\
0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 1
\end{array}\right]\right\}
$$

and

$$
\cap_{\lambda \in \mathbb{C}^{0}} \mathcal{S}_{\lambda}\left(\Sigma_{\mathrm{P}}\right)=\operatorname{Im}\left\{\left[\begin{array}{llll}
1 & 0 & 0 & 0  \tag{7.3.39}\\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 1
\end{array}\right]\right\}
$$

Hence,

$$
\mathcal{S}^{+}\left(\Sigma_{\mathrm{P}}\right) \cap\left\{\cap_{\lambda \in \mathbf{C}^{0}} \mathcal{S}_{\lambda}\left(\Sigma_{\mathrm{P}}\right)\right\}=\operatorname{Im}\left\{\left[\begin{array}{llll}
1 & 0 & 0 & 0  \tag{7.3.40}\\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 1
\end{array}\right]\right\} .
$$



Figure 7.3.1: Max. singular values of $T_{h w}$ - State feedback.

Obviously, $\operatorname{Im}(E) \subset \mathcal{S}^{+}\left(\Sigma_{\mathbf{P}}\right) \cap\left\{\cap_{\lambda \in \mathbb{C}^{0}} \mathcal{S}_{\lambda}\left(\Sigma_{\mathbf{P}}\right)\right\}$ and by Remark 7.2.2, the $H_{\infty^{-}}$ ADDPS is achievable for the given system. Following our algorithm, we obtain a state feedback gain matrix

$$
F(\varepsilon)=\left[\begin{array}{rrrrrr}
0 & 0 & 0 & 0 & -6 & -2  \tag{7.3.41}\\
-\varepsilon^{2} / 3-1 & 2 \varepsilon^{2} / 9-\varepsilon-2 & 2 \varepsilon / 3-\varepsilon^{2} / 27-4 & -4 & -5 & -1 / \varepsilon-6
\end{array}\right]
$$

which places the closed-loop poles of $A+B F(\varepsilon)$ asymptotically at $-1,-2,-\varepsilon$, $-\varepsilon,-\varepsilon$ and $-1 / \varepsilon$. The maximum singular value plots of the corresponding closed-loop transfer matrix $T_{h w}(s, \varepsilon)$ in Figure 7.3 .1 clearly show that the $H_{\infty^{-}}$ ADDPS is attained as $\varepsilon$ tends smaller and smaller.

### 7.4. Solutions to Output Feedback Case

We present in this section the designs of both full order and reduced order output feedback controllers that solve the general $H_{\infty}$-ADDPMS for the given system (7.1.1). Here, by full order controller, we mean that the order of the controller is exactly the same as the given system (7.1.1), i.e, is equal to $n$. A reduced order controller, on the other hand, refers to a controller whose
dynamical order is less than $n$. We will assume without loss of any generality that $D_{22}=0$ in the given system (7.1.1) throughout this section.

### 7.4.1. Full Order Output Feedback

The following is a step-by-step algorithm for constructing a parameterized full order output feedback controller that solves the general $H_{\infty}$-ADDPMS:

Step 7.F.C.1: (Construction of the gain matrix $F_{\mathrm{P}}(\varepsilon)$ ). Define an auxiliary system

$$
\left\{\begin{array}{l}
\dot{x}=A x+B u+E w  \tag{7.4.1}\\
y=x \\
h=C_{2} x+D_{2} u+D_{22} w
\end{array}\right.
$$

and then perform Step 7.S. 1 to 7.S. 6 of the previous section to the above system to obtain a parameterized gain matrix $F(\varepsilon)$. We let $F_{\mathrm{P}}(\varepsilon)=F(\varepsilon)$.

Step 7.F.C.2: (Construction of the gain matrix $K_{\mathrm{Q}}(\varepsilon)$ ). Define another auxiliary system

$$
\left\{\begin{array}{l}
\dot{x}=A^{\prime} x+C_{1}^{\prime} u+C_{2}^{\prime} w  \tag{7.4.2}\\
y=x \\
h=E^{\prime} x+D_{1}^{\prime} u+D_{22}^{\prime} w
\end{array}\right.
$$

and then perform Step 7.S. 1 to 7.S. 6 of the previous section to the above system to get the parameterized gain matrix $F(\varepsilon)$. We let $K_{\mathrm{Q}}(\varepsilon)=F(\varepsilon)^{\prime}$.

Step 7.F.C.3: (Construction of the full order controller $\Sigma_{\mathrm{FC}}(\varepsilon)$ ). Finally, the parameterized full order output feedback controller is given by

$$
\Sigma_{\mathrm{FC}}(\varepsilon):\left\{\begin{array}{c}
\dot{x}_{c}=A_{\mathrm{FC}}(\varepsilon) x_{c}+B_{\mathrm{FC}}(\varepsilon) y,  \tag{7.4.3}\\
u=C_{\mathrm{FC}}(\varepsilon) x_{c}+D_{\mathrm{FC}}(\varepsilon) y,
\end{array}\right.
$$

where

$$
\left.\begin{array}{l}
A_{\mathrm{FC}}(\varepsilon):=A+B F_{\mathrm{P}}(\varepsilon)+K_{\mathrm{Q}}(\varepsilon) C_{1},  \tag{7.4.4}\\
B_{\mathrm{FC}}(\varepsilon):=-K_{\mathrm{Q}}(\varepsilon), \\
C_{\mathrm{FC}}(\varepsilon):=F_{\mathrm{P}}(\varepsilon), \\
D_{\mathrm{FC}}(\varepsilon):=0 .
\end{array}\right\}
$$

We have the following theorem.

Theorem 7.4.1. Consider the given system (7.1.1) with $D_{22}=0$ satisfying all the conditions in Theorem 7.2.1. Then the closed-loop system comprising (7.1.1) and the full order output feedback controller (7.4.3) has the following properties: For any given $\gamma>0$, there exists a positive scalar $\varepsilon^{*}>0$ such that for all $0<\varepsilon \leq \varepsilon^{*}$,

1. the resulting closed-loop system is asymptotically stable; and
2. the $H_{\infty}$-norm of the resulting closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$ is less than $\gamma$, i.e., $\left\|T_{h w}(s, \varepsilon)\right\|_{\infty}<\gamma$.

By Definition 7.1.1, the control law (7.4.3) solves the general $H_{\infty}$-ADDPMS for the given system (7.1.1).

Proof. See Subsection 7.5.B.
因

We illustrate the above result in the following example.

Example 7.4.1. We re-consider the system (7.1.1) with $A, B, E, C_{2}, D_{2}$ and $D_{22}$ as in Example 7.3 .1 but with

$$
C_{1}=\left[\begin{array}{rrrrrr}
-1 & -1 & 0 & 1 & 1 & 1  \tag{7.4.5}\\
0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & 0 & 0 & 0 & 1
\end{array}\right], \quad D_{1}=\left[\begin{array}{ll}
1 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right]
$$

Using the software toolboxes of Chen [9] and Lin [60], we can easily obtain the following properties of $\Sigma_{\mathrm{Q}}$ : i) $\left(A, C_{1}\right)$ is detectable; ii) $\Sigma_{\mathrm{Q}}$ has two stable invariant zeros at -1 and -0.5616 , one imaginary axis invariant zero at 0 , and one unstable invariant zero at 3.5616 ; iii) $\Sigma_{Q}$ has one infinite zero of order zero and one infinite zero of order one; iv) $\Sigma_{Q}$ is left invertible; and v)

It is straightforward to see that $\operatorname{Ker}\left(C_{2}\right) \supset \mathcal{V}^{+}\left(\Sigma_{\mathbf{Q}}\right) \cup\left\{\cup_{\lambda \in \mathbb{C}^{0}} \mathcal{V}_{\lambda}\left(\Sigma_{\mathbf{Q}}\right)\right\}$ and $\mathcal{V}^{+}\left(\Sigma_{\mathbf{Q}}\right) \subset \mathcal{S}^{+}\left(\Sigma_{\mathbf{P}}\right)$. By Theorem 7.2.1, the $H_{\infty}-$ ADDPMS is solvable for the


Figure 7.4.1: Max. singular values of $T_{h w}$ - Full order output feedback.
given system. Following our algorithm, we obtain a full order output feedback controller of the form (7.4.3) with $F_{\mathrm{P}}(\varepsilon)$ as given in (7.3.41) and

$$
K_{\mathbf{Q}}(\varepsilon)=\left[\begin{array}{rrr}
2.4375 & 1 & 0.1813  \tag{7.4.7}\\
2.4028 & 2 & -0.0808 \\
0 & 0 & -3.1758 \\
0 & -3 & -4 \\
0 & -8.2462 & -5 \\
-3 & -2 & -1 / \varepsilon-3
\end{array}\right]
$$

which places the closed-loop eigenvalues of $A+K_{\mathrm{Q}}(\varepsilon) C_{1}$ asymptotically at $-0.5616,-1,-4.2462,-4.2787,-\varepsilon$ and $-1 / \varepsilon$. The maximum singular value plots of the corresponding closed-loop transfer matrix $T_{h w}(s, \varepsilon)$ in Figure 7.4.1 show that the $H_{\infty}$-ADDPMS is attained as $\varepsilon$ tends to zero.

### 7.4.2. Reduced Order Output Feedback

In this subsection, we follow the procedure of Chen et al [26] to design a reduced order output feedback controller. We will show that such as a controller structure with appropriately chosen gain matrices also solves the general $H_{\infty^{-}}$ ADDPMS for the system (7.1.1). First, without loss of generality and for
simplicity of presentation, we assume that the matrices $C_{1}$ and $D_{1}$ are already in the form,

$$
C_{1}=\left[\begin{array}{cc}
0 & C_{1,02}  \tag{7.4.8}\\
I_{k} & 0
\end{array}\right] \quad \text { and } \quad D_{1}=\left[\begin{array}{c}
D_{1,0} \\
0
\end{array}\right]
$$

where $k=\ell-\operatorname{rank}\left(D_{1}\right)$ and $D_{1,0}$ is of full rank. Then the given system (7.1.1) can be written as

$$
\left\{\begin{align*}
\binom{\dot{x}_{1}}{\dot{x}_{2}} & =\left[\begin{array}{ll}
A_{11} & A_{12} \\
A_{21} & A_{22}
\end{array}\right]\binom{x_{1}}{x_{2}}+\left[\begin{array}{l}
B_{1} \\
B_{2}
\end{array}\right] u+\left[\begin{array}{l}
E_{1} \\
E_{2}
\end{array}\right] w  \tag{7.4.9}\\
\binom{y_{0}}{y_{1}} & =\left[\begin{array}{cc}
0 & C_{1,02} \\
I_{k} & 0
\end{array}\right]\binom{x_{1}}{x_{2}} \\
h & =\left[\begin{array}{cc}
D_{1,0} \\
0
\end{array}\right] w
\end{align*}\right.
$$

where the original state $x$ is partitioned into two parts, $x_{1}$ and $x_{2}$; and $y$ is partitioned into $y_{0}$ and $y_{1}$ with $y_{1} \equiv x_{1}$. Thus, one needs to estimate only the state $x_{2}$ in the reduced order controller design. Next, define an auxiliary subsystem $\Sigma_{\mathrm{QR}}$ characterized by a matrix quadruple ( $A_{\mathrm{R}}, E_{\mathrm{R}}, C_{\mathrm{R}}, D_{\mathrm{R}}$ ), where

$$
\left(A_{\mathrm{R}}, E_{\mathrm{R}}, C_{\mathrm{R}}, D_{\mathrm{R}}\right)=\left(A_{22}, E_{2},\left[\begin{array}{c}
C_{1,02}  \tag{7.4.10}\\
A_{12}
\end{array}\right],\left[\begin{array}{c}
D_{1,0} \\
E_{1}
\end{array}\right]\right)
$$

The following is a step-by-step algorithm that constructs the reduced order output feedback controller for the general $H_{\infty}$-ADDPMS.

Step 7.R.C.1: (Construction of the gain matrix $F_{\mathbf{P}}(\varepsilon)$ ). Define an auxiliary system

$$
\left\{\begin{array}{l}
\dot{x}=A x+B u+E w  \tag{7.4.11}\\
y=x \\
h=C_{2} x+D_{2} u+D_{22} w
\end{array}\right.
$$

and then perform Step 7.S.1 to 7.S. 6 of Section 7.3 to the above system to get the parameterized gain matrix $F(\varepsilon)$. We let $F_{\mathrm{P}}(\varepsilon)=F(\varepsilon)$.

Step 7.R.C.2: (Construction of the gain matrix $K_{\mathrm{R}}(\varepsilon)$ ). Define another auxiliary system

$$
\left\{\begin{array}{l}
\dot{x}=A_{\mathrm{R}}^{\prime} x+C_{\mathrm{R}}^{\prime} u+C_{2,2}^{\prime} w  \tag{7.4.12}\\
y=x \\
h=E_{\mathrm{R}}^{\prime} x+D_{\mathrm{R}}^{\prime} u+D_{22}^{\prime} w
\end{array}\right.
$$

and then perform Step 7.S. 1 to 7.S. 6 of Section 7.3 to the above system to get the parameterized gain matrix $F(\varepsilon)$. We let $K_{\mathrm{R}}(\varepsilon)=F(\varepsilon)^{\prime}$.

Step 7.R.C.3: (Construction of the reduced order controller $\Sigma_{\mathrm{Rc}}(\varepsilon)$ ). Let us partition $F_{\mathrm{P}}(\varepsilon)$ and $K_{\mathrm{R}}(\varepsilon)$ as,

$$
F_{\mathrm{P}}(\varepsilon)=\left[\begin{array}{lll}
F_{\mathrm{P} 1}(\varepsilon) & F_{\mathrm{P} 2}(\varepsilon)
\end{array}\right] \quad \text { and } \quad K_{\mathrm{R}}(\varepsilon)=\left[\begin{array}{ll}
K_{\mathrm{R} 0}(\varepsilon) & K_{\mathrm{R} 1}(\varepsilon) \tag{7.4.13}
\end{array}\right]
$$

in conformity with the partitions of $x=\binom{x_{1}}{x_{2}}$ and $y=\binom{y_{0}}{y_{1}}$, respectively. Then define

$$
\begin{equation*}
G_{\mathrm{R}}(\varepsilon)=\left[-K_{\mathrm{R} 0}(\varepsilon), A_{21}+K_{\mathrm{R} 1}(\varepsilon) A_{11}-\left(A_{\mathrm{R}}+K_{\mathrm{R}}(\varepsilon) C_{\mathrm{R}}\right) K_{\mathrm{R} 1}(\varepsilon)\right] \tag{7.4.14}
\end{equation*}
$$

Finally, the parameterized reduced order output feedback controller is given by

$$
\Sigma_{\mathrm{RC}}(\varepsilon):\left\{\begin{array}{c}
\dot{x}_{c}=A_{\mathrm{RC}}(\varepsilon) x_{c}+B_{\mathrm{RC}}(\varepsilon) y  \tag{7.4.15}\\
u=C_{\mathrm{RC}}(\varepsilon) x_{c}+D_{\mathrm{RC}}(\varepsilon) y
\end{array}\right.
$$

where

$$
\begin{align*}
& A_{\mathrm{RC}}(\varepsilon):=A_{\mathrm{R}}+B_{2} F_{\mathrm{P} 2}(\varepsilon)+K_{\mathrm{R}}(\varepsilon) C_{\mathrm{R}}+K_{\mathrm{R} 1}(\varepsilon) B_{1} F_{\mathrm{P} 2}(\varepsilon) \\
& B_{\mathrm{RC}}(\varepsilon):=G_{\mathrm{R}}(\varepsilon)+\left[B_{2}+K_{\mathrm{R} 1}(\varepsilon) B_{1}\right]\left[0, \quad F_{\mathrm{P} 1}(\varepsilon)-F_{\mathrm{P} 2}(\varepsilon) K_{\mathrm{R} 1}(\varepsilon)\right], \\
& C_{\mathrm{RC}}(\varepsilon):=F_{\mathrm{P} 2}(\varepsilon), \\
& D_{\mathrm{RC}}(\varepsilon):=\left[0, \quad F_{\mathrm{P} 1}(\varepsilon)-F_{\mathrm{P} 2}(\varepsilon) K_{\mathrm{R} 1}(\varepsilon)\right] . \tag{7.4.16}
\end{align*}
$$

We have the following theorem.
Theorem 7.4.2. Consider the given system (7.1.1) with $D_{22}=0$ satisfying all the conditions in Theorem 7.2.1. Then the closed-loop system comprising (7.1.1) and the reduced order output feedback controller (7.4.15) has the following properties: For any given $\gamma>0$, there exists a positive scalar $\varepsilon^{*}>0$ such that for all $0<\varepsilon \leq \varepsilon^{*}$,

1. the resulting closed-loop system is asymptotically stable; and
2. the $H_{\infty}$-norm of the resulting closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$ is less than $\gamma$, i.e., $\left\|T_{h w}(s, \varepsilon)\right\|_{\infty}<\gamma$.

By Definition 7.1.1, the control law (7.4.15) solves the general $H_{\infty}$-ADDPMS for the given system (7.1.1).

We illustrate the above result in the following example.

Example 7.4.2. We again consider the given system as in Examples 7.3.1 and 7.4.1. As all the five conditions of Theorem 7.2 .1 are satisfied, the $H_{\infty^{-}}$ ADDPMS for the given system can be solved using a reduced order output feedback controller. We will construct such a controller in the following. First, it is simple to show the transformation $T_{s}$ and $T_{o}$,

$$
T_{s}=\left[\begin{array}{llllll}
0 & 0 & 1 & 0 & 0 & 0  \tag{7.4.17}\\
0 & 0 & 0 & 1 & 0 & 0 \\
0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & 0 & 0 & 0 & 1 \\
1 & 0 & 0 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0
\end{array}\right], \quad T_{o}=\left[\begin{array}{lll}
1 & 1 & 1 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{array}\right]
$$

will transform $C_{1}$ and $D_{1}$ to the form of (7.4.8), i.e.,

$$
T_{o}^{-1} C_{1} T_{s}=\left[\begin{array}{c|c}
0 & C_{1,02}  \tag{7.4.18}\\
\hline I_{k} & 0
\end{array}\right]=\left[\begin{array}{rr|rrrr}
0 & 0 & -1 & -1 & 0 & 1 \\
\hline 1 & 0 & 0 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0
\end{array}\right]
$$

and

$$
T_{o}^{-1} D_{1}=\left[\frac{D_{1,0}}{0}\right]=\left[\begin{array}{ll}
1 & 0  \tag{7.4.19}\\
\hline 0 & 0 \\
0 & 0
\end{array}\right]
$$

Moreover, we have

$$
\begin{gather*}
T_{s}^{-1} A T_{s}=\left[\begin{array}{l|l}
A_{11} & A_{12} \\
\hline A_{21} & A_{22}
\end{array}\right]=\left[\begin{array}{ll|llll}
4 & 5 & 0 & 0 & 0 & 0 \\
5 & 6 & 1 & 2 & 3 & 4 \\
\hline 0 & 1 & 0 & 1 & 0 & 0 \\
0 & 2 & 0 & 0 & 1 & 0 \\
0 & 3 & 0 & 0 & 0 & 0 \\
3 & 4 & 0 & 0 & 0 & -1
\end{array}\right],  \tag{7.4.20}\\
T_{s}^{-1} B=\left[\frac{B_{1}}{B_{2}}\right]=\left[\begin{array}{ll}
1 & 0 \\
0 & 1 \\
\hline 0 & 0 \\
0 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right], \quad T_{s}^{-1} E=\left[\frac{E_{1}}{\hdashline E_{2}}\right]=\left[\begin{array}{ll}
0 & 0 \\
3 & 1 \\
\hline 1 & 0 \\
2 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right], \tag{7.4.21}
\end{gather*}
$$

and $A_{\mathrm{R}}=A_{22}, E_{\mathrm{R}}=E_{2}$, and

$$
C_{\mathrm{R}}=\left[\begin{array}{rrrr}
-1 & -1 & 0 & 1  \tag{7.4.22}\\
0 & 0 & 0 & 0 \\
1 & 2 & 3 & 4
\end{array}\right], \quad D_{\mathrm{R}}=\left[\begin{array}{ll}
1 & 0 \\
0 & 0 \\
3 & 1
\end{array}\right]
$$

Following our algorithm, we obtain

$$
\begin{align*}
F_{\mathrm{P}}(\varepsilon) T_{s} & =\left[F_{\mathrm{P} 1}(\varepsilon) \mid F_{\mathrm{P} 2}(\varepsilon)\right] \\
& =\left[\begin{array}{rr|rrrr}
-6 & -2 & 0 & 0 & 0 & 0 \\
-5 & -1 / \varepsilon-6 & -\varepsilon^{2} / 3-1 & 2 \varepsilon^{2} / 9-\varepsilon-2 & 2 \varepsilon / 3-\varepsilon^{2} / 27-4 & -4
\end{array}\right] \tag{7.4.23}
\end{align*}
$$

and

$$
K_{\mathrm{R}}(\varepsilon)=\left[K_{\mathrm{R} 0}(\varepsilon) \mid K_{\mathrm{R} 1}(\varepsilon)\right]=\left[\begin{array}{r|rr}
1.2000+0.1219 \varepsilon & 0 & -0.6663+0.4025 \varepsilon  \tag{7.4.24}\\
0.8187-0.0609 \varepsilon & 0 & -0.8534-0.2012 \varepsilon \\
-0.1219 \varepsilon & 0 & -0.4025 \varepsilon \\
0 & 0 & 0
\end{array}\right]
$$

which place the eigenvalues of $A_{\mathrm{R}}+K_{\mathrm{R}}(\varepsilon) C_{\mathrm{R}}$ at $-0.5616,-1,-3.8303$ and $-\varepsilon$. Also, we obtain a reduced order output feedback controller of the form (7.4.15) with all sub-matrices as defined in (7.4.18) to (7.4.24), and with $B_{\mathrm{RC}}(\varepsilon)$ and $D_{\mathrm{RC}}(\varepsilon)$ being slightly modified to

$$
\begin{equation*}
B_{\mathrm{RC}}(\varepsilon)=G_{\mathrm{R}}(\varepsilon) T_{o}^{-1}+\left[B_{2}+K_{\mathrm{R} 1}(\varepsilon) B_{1}\right]\left[0, F_{\mathrm{P} 1}(\varepsilon)-F_{\mathrm{P} 2}(\varepsilon) K_{\mathrm{R} 1}(\varepsilon)\right] T_{o}^{-1} \tag{7.4.25}
\end{equation*}
$$

and

$$
\begin{equation*}
D_{\mathrm{RC}}(\varepsilon)=\left[0, F_{\mathrm{P} 1}(\varepsilon)-F_{\mathrm{P} 2}(\varepsilon) K_{\mathrm{R} 1}(\varepsilon)\right] T_{o}^{-1} \tag{7.4.26}
\end{equation*}
$$

respectively. The maximum singular value plots of the corresponding closedloop transfer matrix $T_{h w}(s, \varepsilon)$ in Figure 7.4.2 also show that the $H_{\infty}$-ADDPMS is attained as $\varepsilon$ tends to zero.

### 7.5. Proofs of Main Results

We present the proofs of all the main results of this chapter in this section.

### 7.5.A. Proof of Theorem 7.3.1

Under the feedback law $u=F(\varepsilon) x$, the closed-loop system on the special coordinate basis can be written as follows,

$$
\begin{align*}
& \dot{x}_{a}^{-}=A_{a a}^{-} x_{a}^{-}+B_{0 a}^{-} h_{0}+L_{a d}^{-} h_{d}+L_{a b}^{-} h_{b}+E_{a}^{-} w  \tag{7.5.1}\\
& \dot{x}_{a}^{0}=A_{a a}^{0} x_{a}^{0}+B_{0 a}^{0} h_{0}+L_{a d}^{0} h_{d}+L_{a b}^{0} h_{b}+E_{a}^{0} w  \tag{7.5.2}\\
& \dot{x}_{a b}^{+}=A_{a b}^{+c} x_{a b}^{+}-B_{0 a b}^{+} F_{a 0}^{0}(\varepsilon)\left[x_{a}^{0}+T_{a}^{0} x_{a b}^{+}\right] \\
& \quad \quad+L_{a b d}^{+}\left[F_{a d}^{+}, F_{b d}\right] x_{a b}^{+}+L_{a b d}^{+} h_{d}+E_{a b}^{+} w \tag{7.5.3}
\end{align*}
$$



Figure 7.4.2: Max. singular values of $T_{h w}$ - Reduced order output feedback.

$$
\begin{align*}
h_{b} & =\left[0_{m_{b} \times n_{a}^{+}}, C_{b}\right] x_{a b}^{+},  \tag{7.5.4}\\
\dot{x}_{c} & =A_{c c}^{c}+L_{c 0} h_{0}+L_{c b} h_{b}+L_{c d} h_{d}+E_{c} w,  \tag{7.5.5}\\
h_{0} & =-\left[F_{a 0}^{+}, F_{b 0}\right] x_{a b}^{+}-F_{a 0}^{0}(\varepsilon)\left(x_{a}^{0}+T_{a}^{0} x_{a b}^{+}\right),  \tag{7.5.6}\\
\dot{x}_{i} & =A_{q_{i}} x_{i}+L_{i 0} h_{0}+L_{i d} h_{d}-\frac{1}{\varepsilon^{q_{i}}} B_{q_{i}}\left[F_{a d i}^{+} F_{i q_{i}} x_{a}^{+}+F_{b d i} F_{i q_{i}} x_{b}\right. \\
& \left.\quad+F_{a d i}^{0}(\varepsilon) F_{i q_{i}}\left[x_{a}^{0}+T_{a}^{0} x_{a b}^{+}\right]+F_{i} S_{i}(\varepsilon) x_{i}\right]+E_{i} w,  \tag{7.5.7}\\
& \quad h_{i}=C_{q_{i}} x_{i}, \quad i=1,2, \cdots, m_{d}, \tag{7.5.8}
\end{align*}
$$

where $x_{a b}^{+}=\left[\left(x_{a}^{+}\right)^{\prime}, x_{b}^{\prime}\right]^{\prime}$ and $B_{0 a b}^{+}$and $L_{a b d}^{+}$are as defined in Step 7.S.4.1 of the state feedback design algorithm. We have also used Condition 2 of $\mathrm{Re}-$ mark 7.2.2, i.e., $D_{22}=0$, and $E_{a}^{-}, E_{a}^{0}, E_{a b}^{+}, E_{b}, E_{c}$ and $E_{i}, i=1,2, \cdots, m_{d}$, are defined as follows,

$$
\Gamma_{s \mathrm{P}}^{-1} E=\left[\begin{array}{llllllll}
\left(E_{a}^{-}\right)^{\prime} & \left(E_{a}^{0}\right)^{\prime} & \left(E_{a b}^{+}\right)^{\prime} & E_{c}^{\prime} & E_{1}^{\prime} & E_{2}^{\prime} & \cdots & E_{m_{d}}^{\prime} \tag{7.5.9}
\end{array}\right]^{\prime}
$$

Condition 4 of the theorem then implies that

$$
\begin{equation*}
E_{a b}^{+}=0 \tag{7.5.10}
\end{equation*}
$$

and

$$
\begin{equation*}
\operatorname{Im}\left(E_{a}^{0}\right) \subset \mathcal{S}\left(A_{a a}^{0}\right):=\cap_{w \in \lambda\left(A_{a a}^{0}\right)} \operatorname{Im}\left\{w I-A_{a a}^{0}\right\} \tag{7.5.11}
\end{equation*}
$$

To complete the proof, we will make two state transformations on the closedloop system (7.5.1)-(7.5.8). The first state transformation is given as follows,

$$
\begin{align*}
\bar{x}_{a b}= & \Gamma_{a b}^{-1} x_{a b}  \tag{7.5.12}\\
\bar{x}_{c} & =x_{c}  \tag{7.5.13}\\
\bar{x}_{i 1} & =x_{i 1}+F_{a d i}^{+} x_{a}^{+}+F_{b d i} x_{b}+F_{a d i}^{0}(\varepsilon)\left[x_{a}^{0}+T_{a}^{0} x_{a b}^{+}\right], \\
& \quad i=1,2, \cdots, m_{d},  \tag{7.5.14}\\
\bar{x}_{i j} & =x_{i j}, j=2,3, \cdots, q_{i}, i=1,2, \cdots, m_{d}, \tag{7.5.15}
\end{align*}
$$

where $x_{a b}=\left[\left(x_{a}^{-}\right)^{\prime},\left(x_{a}^{0}\right)^{\prime},\left(x_{a b}^{+}\right)^{\prime}\right]^{\prime}$ and $\bar{x}_{a b}=\left[\left(\bar{x}_{a}^{-}\right)^{\prime},\left(\bar{x}_{a b}^{+}\right)^{\prime},\left(\bar{x}_{a}^{0}\right)^{\prime}\right]^{\prime}$. In the new state variables (7.5.13)-(7.5.15), the closed-loop system becomes,

$$
\begin{align*}
\dot{\bar{x}}_{a}^{-}= & A_{a a}^{-} \bar{x}_{a}^{-}+A_{a a b}^{-} \bar{x}_{a b}^{+}-\left[B_{0 a}^{-}, L_{a d}^{-}\right] F_{a}^{0}(\varepsilon) \bar{x}_{a}^{0}+L_{a d}^{-} \bar{h}_{d}+E_{a}^{-} w,  \tag{7.5.16}\\
\dot{\bar{x}}_{a b}^{+}= & A_{a b}^{+c} \bar{x}_{a b}^{+}-\left[B_{0 a b}^{+}, L_{a b d}^{+}\right] F_{a}^{0}(\varepsilon) \bar{x}_{a}^{0}+L_{a b d}^{+} \bar{h}_{d},  \tag{7.5.17}\\
\dot{\bar{x}}_{a}^{0}= & \left(A_{a a}^{0}-B_{a}^{0} F_{a}^{0}(\varepsilon)\right) \bar{x}_{a}^{0}+\left(L_{a d}^{0}+T_{a}^{0} L_{a b d}^{+}\right) \bar{h}_{d}+E_{a}^{0} w,  \tag{7.5.18}\\
\dot{\bar{x}}_{c}= & A_{c c}^{c} \bar{x}_{c}+\left(L_{c b}\left[0, C_{b}\right]-\left[L_{c 0}, L_{c d}\right] F_{a b}^{+}\right) \bar{x}_{a b}^{+} \\
& \quad-\left[L_{c 0}, L_{c d}\right] F_{a}^{0}(\varepsilon) \bar{x}_{a}^{0}+L_{c d} \bar{h}_{d}+E_{c} w,  \tag{7.5.19}\\
h_{0}=- & {\left[F_{a 0}^{+}, F_{b 0}\right] \bar{x}_{a b}^{+}-F_{a 0}^{0}(\varepsilon) \bar{x}_{a}^{0}, }  \tag{7.5.20}\\
\dot{\bar{x}}_{i}= & A_{q_{i}} \bar{x}_{i}-\frac{1}{\varepsilon^{q_{i}}} B_{q_{i}} F_{i} S_{i}(\varepsilon) \bar{x}_{i}+L_{i a b}^{+}(\varepsilon) \bar{x}_{a b}^{+}+L_{i a}^{01}(\varepsilon) F_{a}^{0}(\varepsilon) \bar{x}_{a}^{0} \\
& \quad+L_{i a}^{02}(\varepsilon) F_{a}^{0}(\varepsilon) A_{a a}^{0} \bar{x}_{a}^{0}+\bar{L}_{i d}(\varepsilon) \bar{h}_{d}+\bar{E}_{i}(\varepsilon) w,  \tag{7.5.21}\\
\bar{h}_{i}= & h_{i}+\left[F_{a d i}^{+}, F_{b d i}\right] \bar{x}_{a b}^{+}+F_{a d i}^{0} \bar{x}_{a}^{0}=C_{q_{i}} \bar{x}_{i}, i=1,2, \cdots, m_{d},  \tag{7.5.22}\\
\bar{h}_{d}= & {\left[\bar{h}_{1}, \bar{h}_{2}, \cdots, \bar{h}_{m_{d}}\right], } \tag{7.5.23}
\end{align*}
$$

where $A_{a a b}^{-}, A_{a a b}^{0}, B_{a}^{0}$ and $L_{a b d}^{+}$are as defined in Step 7.S.4.1 of the state feedback law design algorithm, and $L_{i a b}^{+}(\varepsilon), L_{i a}^{01}(\varepsilon), L_{i a}^{02}(\varepsilon), \bar{L}_{i d}(\varepsilon)$ and $\bar{E}_{i}(\varepsilon)$ are defined in an obvious way and, by (7.3.20), satisfy

$$
\begin{equation*}
\left\|L_{i a b}^{+}(\varepsilon)\right\| \leq l_{i a b}^{+}, \quad\left\|L_{i a}^{01}(\varepsilon)\right\| \leq l_{i a}^{01}, \quad\left\|L_{i a}^{02}(\varepsilon)\right\| \leq l_{i a}^{02} \tag{7.5.24}
\end{equation*}
$$

and

$$
\begin{equation*}
\left\|\bar{L}_{i d}(\varepsilon)\right\| \leq \bar{l}_{i d}, \quad\left\|\bar{E}_{i}(\varepsilon)\right\| \leq \bar{e}_{i}, \quad \varepsilon \in(0,1] \tag{7.5.25}
\end{equation*}
$$

for some nonnegative constants $l_{i a b}^{+}, l_{i a}^{01}, l_{i a}^{02} \bar{e}_{i}$ and $\bar{l}_{i d}$, independent of $\varepsilon$.
We now proceed to construct the second transformation. We need to recall the following preliminary results from [63].

Lemma 7.5.1. Let the triple $\left(A_{i}, B_{i}, F_{i}(\varepsilon)\right)$ be as given in Steps 7.S.4. 2 and 7.S.4.3 of the state feedback design algorithm. Then, there exists a nonsingular state transformation matrix $Q_{i}(\varepsilon) \in \mathbb{R}^{n_{i} \times n_{i}}$ such that

1. $Q_{i}(\varepsilon)$ transforms $A_{i}-B_{i} F_{i}(\varepsilon)$ into a real Jordan form, i.e.,

$$
\begin{align*}
Q_{i}^{-1}(\varepsilon)\left[A_{i}\right. & \left.-B_{i} F_{i}(\varepsilon)\right] Q_{i}(\varepsilon)=J_{i}(\varepsilon) \\
& =\operatorname{blkdiag}\left\{J_{i 0}(\varepsilon), J_{i 1}(\varepsilon), J_{i 2}(\varepsilon), \cdots, J_{i p_{i}}(\varepsilon)\right\} \tag{7.5.26}
\end{align*}
$$

where

$$
J_{i 0}(\varepsilon)=\left[\begin{array}{cccc}
-\varepsilon & 1 & &  \tag{7.5.27}\\
& \ddots & \ddots & \\
& & -\varepsilon & 1 \\
& & & -\varepsilon
\end{array}\right]_{r_{i 0} \times r_{i 0}}
$$

and for each $j=1$ to $p_{i}$,

$$
J_{i j}(\varepsilon)=\left[\begin{array}{cccc}
J_{i j}^{\star}(\varepsilon) & I_{2} & & .  \tag{7.5.28}\\
& \ddots & \ddots & \\
& & J_{i j}^{\star}(\varepsilon) & I_{2} \\
& & & J_{i j}^{\star}(\varepsilon)
\end{array}\right]_{2 r_{i j} \times 2 r_{i j}} \quad, \quad J_{i j}^{\star}(\varepsilon)=\left[\begin{array}{cc}
-\varepsilon & \beta_{i j} \\
-\beta_{i j} & -\varepsilon
\end{array}\right],
$$

with $\beta_{i j}>0$ for all $j=1$ to $p_{i}$ and $\beta_{i j} \neq \beta_{i k}$ for $j \neq k$.
2. Both $\left\|Q_{i}(\varepsilon)\right\|$ and $\left\|Q_{i}^{-1}(\varepsilon)\right\|$ are bounded, i.e.,

$$
\begin{equation*}
\left\|Q_{i}(\varepsilon)\right\| \leq \theta_{i}, \quad\left\|Q_{i}^{-1}(\varepsilon)\right\| \leq \theta_{i}, \quad \varepsilon \in(0,1] \tag{7.5.29}
\end{equation*}
$$

for some positive constant $\theta_{i}$, independent of $\varepsilon$.
3. If $E_{i} \in \mathbb{R}^{n_{i} \times q}$ is such that

$$
\begin{equation*}
\operatorname{Im}\left(E_{i}\right) \subset \cap_{w \in \lambda\left(A_{i}\right)} \operatorname{Im}\left(w I-A_{i}\right) \tag{7.5.30}
\end{equation*}
$$

then, there exists a $\delta_{i} \geq 0$, independent of $\varepsilon$, such that

$$
\begin{equation*}
\left\|Q_{i}^{-1}(\varepsilon) E_{i}\right\| \leq \delta_{i}, \quad \varepsilon \in(0,1] \tag{7.5.31}
\end{equation*}
$$

and, if we partition $Q_{i}^{-1}(\varepsilon) E_{i}$ according to that of $J_{i}(\varepsilon)$ as,

$$
Q_{i}^{-1}(\varepsilon) E_{i}=\left[\begin{array}{c}
E_{i 0}(\varepsilon)  \tag{7.5.32}\\
E_{i 1}(\varepsilon) \\
\vdots \\
E_{i p_{i}}(\varepsilon)
\end{array}\right], \quad E_{i 0}(\varepsilon)=\left[\begin{array}{c}
E_{i 01}(\varepsilon) \\
E_{i 02}(\varepsilon) \\
\vdots \\
E_{i 0 r_{i 0}}(\varepsilon)
\end{array}\right]_{r_{i 0} \times 1},
$$

and

$$
E_{i j}(\varepsilon)=\left[\begin{array}{c}
E_{i j 1}(\varepsilon)  \tag{7.5.33}\\
E_{i j 2}(\varepsilon) \\
\vdots \\
E_{i j r_{i j}}(\varepsilon)
\end{array}\right]_{2 r_{i j} \times 1}
$$

then, there exists a $\beta_{i} \geq 0$, independent of $\varepsilon$, such that, for each $j=0$, to $p_{i}$,

$$
\begin{equation*}
\left\|E_{i j r_{i j}}(\varepsilon)\right\| \leq \beta_{i} \varepsilon \tag{7.5.34}
\end{equation*}
$$

4. If we define a scaling matrix $S_{a i}(\varepsilon)$ as

$$
\begin{equation*}
S_{a i}(\varepsilon)=\operatorname{blkdiag}\left\{S_{a i 0}(\varepsilon), S_{a i 1}(\varepsilon), S_{a i 2}(\varepsilon), \cdots, S_{a i p_{i}}(\varepsilon)\right\} \tag{7.5.35}
\end{equation*}
$$

where

$$
\begin{equation*}
S_{a i 0}(\varepsilon)=\operatorname{diag}\left\{\varepsilon^{r_{i 0}-1}, \varepsilon^{r_{i 0}-2}, \cdots, \varepsilon, 1\right\} \tag{7.5.36}
\end{equation*}
$$

and for $j=1$ to $p_{i}$,

$$
\begin{equation*}
S_{a i j}(\varepsilon)=\operatorname{blkdiag}\left\{\varepsilon^{r_{i j}-1} I_{2}, \varepsilon^{r_{i j}-2} I_{2}, \cdots, \varepsilon I_{2}, I_{2}\right\} \tag{7.5.37}
\end{equation*}
$$

then, there exists a $\kappa_{i} \geq 0$ independent of $\varepsilon$ such that,

$$
\begin{equation*}
\left|F_{i}(\varepsilon) Q_{i}(\varepsilon) S_{a i}^{-1}(\varepsilon)\right| \leq \kappa_{i} \varepsilon, \quad\left|F_{i}(\varepsilon) A_{i} Q_{i}(\varepsilon) S_{a i}^{-1}(\varepsilon)\right| \leq \kappa_{i} \varepsilon \tag{7.5.38}
\end{equation*}
$$

Proof. This is a combination of the results of [63], and (2.2.13) of [61].
Lemma 7.5.2. Let

$$
\begin{equation*}
\tilde{J}_{i}(\varepsilon)=\operatorname{blkdiag}\left\{\tilde{J}_{i 0}, \tilde{J}_{i 1}(\varepsilon), \cdots, \tilde{J}_{i p_{i}}(\varepsilon)\right\} \tag{7.5.39}
\end{equation*}
$$

where

$$
\tilde{J}_{i 0}=\left[\begin{array}{cccc}
-1 & 1 & &  \tag{7.5.40}\\
& \ddots & \ddots & \\
& & -1 & 1 \\
& & & -1
\end{array}\right]_{r_{i 0} \times r_{i 0}}
$$

and for each $j=1$ to $p_{i}$,

$$
\tilde{J}_{i j}(\varepsilon)=\left[\begin{array}{cccc}
\tilde{J}_{i j}^{\star}(\varepsilon) & I_{2} & &  \tag{7.5.41}\\
& \ddots & \ddots & \\
& & \tilde{J}_{i j}^{\star}(\varepsilon) & I_{2} \\
& & & \tilde{J}_{i j}^{\star}(\varepsilon)
\end{array}\right]_{2 r_{i j} \times 2 r_{i j}} \quad, \quad \tilde{J}_{i j}^{\star}(\varepsilon)=\left[\begin{array}{cc}
-1 & \beta_{i j} / \varepsilon \\
-\beta_{i j} / \varepsilon & -1
\end{array}\right]
$$

with $\beta_{i j}>0$ for all $j=1$ to $p_{i}$ and $\beta_{j} \neq \beta_{k}$ for $j \neq k$. Then the unique positive definite solution $\tilde{P}_{i}$ to the Lyapunov equation,

$$
\begin{equation*}
\tilde{J}_{i}(\varepsilon)^{\prime} \tilde{P}_{i}+\tilde{P}_{i} \tilde{J}_{i}(\varepsilon)=-I \tag{7.5.42}
\end{equation*}
$$

is independent of $\varepsilon$.
Proof. See [63].

We now define the following second state transformation on the closed-loop system,

$$
\begin{align*}
\tilde{x}_{a}^{-}= & \bar{x}_{a}^{-}, \quad \tilde{x}_{a b}^{+}=\bar{x}_{a b}^{+},  \tag{7.5.43}\\
\tilde{x}_{a}^{0}= & {\left[\left(\tilde{x}_{a 1}^{0}\right)^{\prime},\left(\tilde{x}_{a 2}^{0}\right)^{\prime}, \cdots\left(\tilde{x}_{a l}^{0}\right)^{\prime}\right]^{\prime}=S_{a}(\varepsilon) Q^{-1}(\varepsilon)\left(\Gamma_{s a}^{0}\right)^{-1} \bar{x}_{a}^{0}, }  \tag{7.5.44}\\
& S_{a}(\varepsilon)=\operatorname{blkdiag}\left\{S_{a 1}(\varepsilon), S_{a 2}(\varepsilon), \cdots, S_{a l}(\varepsilon)\right\}, \\
& Q(\varepsilon)=\operatorname{blkdiag}\left\{Q_{1}(\varepsilon), Q_{2}(\varepsilon), \cdots, Q_{l}(\varepsilon)\right\}, \\
\tilde{x}_{c}= & \varepsilon \bar{x}_{c},  \tag{7.5.45}\\
\tilde{x}_{d}= & {\left[\tilde{x}_{1}^{\prime}, \tilde{x}_{2}^{\prime}, \cdots, \tilde{x}_{m_{d}}^{\prime}\right]^{\prime}, \quad \tilde{x}_{i}=S_{i}(\varepsilon) \bar{x}_{i}, i=1,2, \cdots, m_{d}, } \tag{7.5.46}
\end{align*}
$$

under which the closed-loop system becomes,

$$
\begin{align*}
\dot{\tilde{x}}_{a}^{-} & =A_{a a}^{-} \tilde{x}_{a}^{-}+A_{a a b}^{-}(\varepsilon) \tilde{x}_{a b}^{+}+A_{a a}^{-0}(\varepsilon) \tilde{x}_{a}^{0}+L_{a d}^{-} \tilde{h}_{d}+E_{a}^{-} w,  \tag{7.5.47}\\
\dot{\tilde{x}}_{a b}^{+} & =A_{a b}^{+c} \tilde{x}_{a b}^{+}+A_{a b a}^{+0}(\varepsilon) \tilde{x}_{a}^{0}+L_{a b d}^{+} \tilde{h}_{d},  \tag{7.5.48}\\
\dot{\tilde{x}}_{a}^{0} & =\tilde{J}(\varepsilon) \tilde{x}_{a}^{0}+\tilde{B}(\varepsilon) \tilde{x}_{a}^{0}+\tilde{L}_{a d}^{0}(\varepsilon) \tilde{h}_{d}+\tilde{E}_{a}^{0}(\varepsilon) w,  \tag{7.5.49}\\
\dot{\tilde{x}}_{c} & =A_{c c}^{c} \tilde{x}_{c}+\varepsilon\left[A_{c a b}^{+} \tilde{x}_{a b}^{+}+A_{c a}^{0}(\varepsilon) \tilde{x}_{a}^{0}+L_{c d} \tilde{h}_{d}+E_{c} w\right],  \tag{7.5.50}\\
h_{0} & =-\left[F_{a 0}^{+}, F_{b 0}\right] \tilde{x}_{a b}^{+}-\tilde{F}_{a 0}^{0}(\varepsilon) \tilde{x}_{a}^{0}  \tag{7.5.51}\\
\varepsilon \dot{x}_{i} & =\left(A_{q_{i}}-B_{q_{i}} F_{i}\right) \tilde{x}_{i}+\varepsilon \tilde{L}_{i a b}^{+}(\varepsilon) \tilde{x}_{a b}^{+}+\varepsilon \tilde{L}_{i a}^{0}(\varepsilon) \tilde{x}_{a}^{0} \\
& \quad+\varepsilon \tilde{L}_{i d}(\varepsilon) \tilde{h}_{d}+\varepsilon \tilde{E}_{i}(\varepsilon) w,  \tag{7.5.52}\\
\tilde{h}_{i} & =\bar{h}_{i}=h_{i}+\left[F_{a d i}^{+}, F_{b d i}\right] \tilde{x}_{a b}^{+}+\tilde{F}_{a d i}^{0}(\varepsilon) \tilde{x}_{a}^{0}=C_{q_{i}} \tilde{x}_{i},  \tag{7.5.53}\\
\tilde{h}_{d} & =\left[\tilde{h}_{1}, \tilde{h}_{2}, \cdots, \tilde{h}_{m_{d}}\right]^{\prime}, \tag{7.5.54}
\end{align*}
$$

where

$$
\begin{align*}
& A_{a a}^{-0}(\varepsilon)=-\left[B_{0 a}^{+}, L_{a d}^{-}\right] F_{a}^{0}(\varepsilon) \Gamma_{s a}^{0} Q(\varepsilon) S_{a}^{-1}(\varepsilon),  \tag{7.5.55}\\
& A_{a b a}^{+0}(\varepsilon)=-\left[B_{0 a b}^{+}, L_{a b d}^{+}\right] F_{a}^{0}(\varepsilon) \Gamma_{s a}^{0} Q(\varepsilon) S_{a}^{-1}(\varepsilon),  \tag{7.5.56}\\
& \tilde{J}(\varepsilon)=\operatorname{blkdiag}\left\{\varepsilon \tilde{J}_{1}(\varepsilon), \varepsilon \tilde{J}_{2}(\varepsilon), \cdots, \varepsilon \tilde{J}_{l}(\varepsilon)\right\},  \tag{7.5.57}\\
& \tilde{B}(\varepsilon)=\left[\begin{array}{ccccc}
0 & \tilde{B}_{12}(\varepsilon) & \tilde{B}_{13}(\varepsilon) & \cdots & \tilde{B}_{1 l}(\varepsilon) \\
0 & 0 & \tilde{B}_{23}(\varepsilon) & \cdots & \tilde{B}_{2 l}(\varepsilon) \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
0 & 0 & 0 & \cdots & 0
\end{array}\right] \text {, }  \tag{7.5.58}\\
& \tilde{B}_{j k}(\varepsilon)=S_{a j}(\varepsilon) Q_{j}^{-1}(\varepsilon) B_{j k} F_{k}(\varepsilon) Q_{k}(\varepsilon) S_{a k}^{-1}(\varepsilon), \\
& j=1,2, \cdots, l, k=j+1, j+2, \cdots, l, \\
& \tilde{L}_{a d}^{0}(\varepsilon)=S_{a}(\varepsilon) Q^{-1}(\varepsilon)\left(\Gamma_{s a}^{0}\right)^{-1}\left(L_{a d}^{0}+T_{a}^{0} L_{a b d}^{+}\right),  \tag{7.5.59}\\
& \tilde{E}_{a}^{0}(\varepsilon)=S_{a}(\varepsilon) Q^{-1}(\varepsilon)\left(\Gamma_{s a}^{0}\right)^{-1} E_{a}^{0} \\
& =\left[\begin{array}{llll}
\left(\tilde{E}_{a 1}^{0}(\varepsilon)\right)^{\prime} & \left(\tilde{E}_{a 2}^{0}(\varepsilon)\right)^{\prime} & \cdots & \left(\tilde{E}_{a l}^{0}(\varepsilon)\right)^{\prime}
\end{array}\right]^{\prime}, \tag{7.5.60}
\end{align*}
$$

$$
\begin{align*}
A_{c a b}^{+} & =\left[L_{c b}\left[0, C_{b}\right]-\left[L_{c 0}, L_{c d}\right] F_{a b}^{+}\right.  \tag{7.5.61}\\
A_{c a}^{0}(\varepsilon) & =-\left[L_{c 0}, L_{c d}\right] F_{a}^{0}(\varepsilon) \Gamma_{s a}^{0} Q(\varepsilon) S_{a}^{-1}(\varepsilon)  \tag{7.5.62}\\
\tilde{F}_{a 0}^{0}(\varepsilon) & =F_{a 0}^{0}(\varepsilon) S_{a}^{-1}(\varepsilon) Q(\varepsilon) \Gamma_{s a}^{0}  \tag{7.5.63}\\
\tilde{L}_{i a b}^{+}(\varepsilon) & =S_{i}(\varepsilon) L_{i a b}^{+}(\varepsilon)  \tag{7.5.64}\\
\tilde{L}_{i a}^{0}(\varepsilon) & =S_{i}(\varepsilon)\left[L_{i a}^{01}(\varepsilon) F_{a}^{0}(\varepsilon)+L_{i a}^{02}(\varepsilon) F_{a}^{0}(\varepsilon) A_{a a}^{0}\right] \Gamma_{s a}^{0} Q(\varepsilon) S_{a}^{-1}(\varepsilon),  \tag{7.5.65}\\
\tilde{L}_{i d}(\varepsilon) & =S_{i}(\varepsilon) \bar{L}_{i d}(\varepsilon)  \tag{7.5.66}\\
\tilde{E}_{i}(\varepsilon) & =S_{i}(\varepsilon) \bar{E}_{i}(\varepsilon)  \tag{7.5.67}\\
\tilde{F}_{a d i}^{0}(\varepsilon) & =F_{a d i}^{0}(\varepsilon) \Gamma_{s a}^{0} Q(\varepsilon) S_{a}^{-1}(\varepsilon) \tag{7.5.68}
\end{align*}
$$

and where, for $i=1$ to $l, \tilde{J}_{i}(\varepsilon)$ is as defined in Lemma 7.5.2.
By (7.3.20), (7.5.24), (7.5.25), and Lemma 7.5.1, we have that, for all $\varepsilon \in$ $(0,1]$,

$$
\begin{gather*}
\left\|A_{a a b}^{-}(\varepsilon)\right\| \leq a_{a a b}^{-},\left\|\tilde{L}_{a d}^{0}(\varepsilon)\right\| \leq \tilde{l}_{a d}^{0}, \quad\left\|A_{c a b}^{+}\right\| \leq a_{c a b}^{+},  \tag{7.5.69}\\
\left\|A_{a a}^{-0}(\varepsilon)\right\| \leq a_{a a}^{-0} \varepsilon, \quad\left\|A_{a b a}^{+0}(\varepsilon)\right\| \leq a_{a a}^{+0} \varepsilon, \quad\left\|A_{c a}^{0}(\varepsilon)\right\| \leq a_{c a}^{0} \varepsilon, \quad\left\|\tilde{F}_{a 0}^{0}(\varepsilon)\right\| \leq \tilde{f}_{a 0}^{0} \varepsilon \tag{7.5.70}
\end{gather*}
$$

for $i=1$ to $m_{d}$,

$$
\begin{equation*}
\left.\| \tilde{L}_{i a b}^{+}(\varepsilon)\right\} \leq \tilde{l}_{a b}^{+}, \quad\left\|\tilde{L}_{i a}^{0}(\varepsilon)\right\| \leq \tilde{l}_{a}^{0} \varepsilon \tag{7.5.71}
\end{equation*}
$$

and

$$
\begin{equation*}
\left\|\tilde{L}_{i d}(\varepsilon)\right\| \leq \tilde{l}_{d}, \quad\left\|\tilde{F}_{a d i}^{0}(\varepsilon)\right\| \leq \tilde{f}_{a d}^{0} \varepsilon, \quad\left\|\tilde{E}_{i}(\varepsilon)\right\| \leq \tilde{e} \tag{7.5.72}
\end{equation*}
$$

for $i=1$ to $l$,

$$
\begin{equation*}
\left\|\tilde{E}_{a i}^{0}(\varepsilon)\right\| \leq \tilde{e}_{a}^{0} \varepsilon \tag{7.5.73}
\end{equation*}
$$

and finally, for $j=1$ to $l, k=j+1$ to $l$,

$$
\begin{equation*}
\left\|\tilde{B}_{j k}(\varepsilon)\right\| \leq \tilde{b}_{j k} \varepsilon \tag{7.5.74}
\end{equation*}
$$

where $a_{a a b}^{-}, \tilde{l}_{a d}^{0}, a_{c a b}^{+}, a_{a a}^{-0}, a_{a a}^{+0}, \tilde{e}_{a}^{0}, a_{c a}^{0}, \tilde{f}_{a 0}^{0}, \tilde{l}_{a b}^{+}, \tilde{l}_{a}^{0}, \tilde{l}_{d}, \tilde{f}_{a d}^{0}, \tilde{e}$ and $\tilde{b}_{j k}$ are some positive constants, independent of $\varepsilon$.

We next construct a Lyapunov function for the closed loop system (7.5.47)(7.5.54). We do this by composing Lyapunov functions for the subsystems. For the subsystem of $\tilde{x}_{a}^{-}$, we choose a Lyapunov function,

$$
\begin{equation*}
V_{a}^{-}\left(\tilde{x}_{a}^{-}\right)=\left(\tilde{x}_{a}^{-}\right)^{\prime} P_{a}^{-} \tilde{x}_{a}^{-}, \tag{7.5.75}
\end{equation*}
$$

where $P_{a}^{-}>0$ is the unique solution to the Lyapunov equation,

$$
\begin{equation*}
\left(A_{a a}^{-}\right)^{\prime} P_{a}^{-}+P_{a}^{-} A_{a a}^{-}=-I \tag{7.5.76}
\end{equation*}
$$

and for the subsystem of $\tilde{x}_{a b}^{+}$, choose a Lyapunov function,

$$
\begin{equation*}
V_{a b}^{+}\left(\tilde{x}_{a b}^{+}\right)=\left(\tilde{x}_{a b}^{+}\right)^{\prime} P_{a b}^{+} \tilde{x}_{a b}^{+}, \tag{7.5.77}
\end{equation*}
$$

where $P_{a b}^{+}>0$ is the unique solution to the Lyapunov equation,

$$
\begin{equation*}
\left(A_{a b}^{+c}\right)^{\prime} P_{a b}^{+}+P_{a b}^{+} A_{a b}^{+c}=-I \tag{7.5.78}
\end{equation*}
$$

The existence of such $P_{a a}^{-}$and $P_{a b}^{+}$is guaranteed by the fact that both $A_{a a}^{-}$and $A_{a b}^{+c}$ are asymptotically stable. For the subsystem of

$$
\begin{equation*}
\tilde{x}_{a}^{0}=\left[\left(\tilde{x}_{a 1}^{0}\right)^{\prime},\left(\tilde{x}_{a 2}^{0}\right)^{\prime}, \cdots,\left(\tilde{x}_{a l}^{0}\right)^{\prime}\right]^{\prime} \tag{7.5.79}
\end{equation*}
$$

we choose a Lyapunov function,

$$
\begin{equation*}
V_{a}^{0}\left(\tilde{x}_{a}^{0}\right)=\sum_{i=1}^{l} \frac{\left(\alpha_{a}^{0}\right)^{i-1}}{\varepsilon}\left(\tilde{x}_{a i}^{0}\right)^{\prime} P_{a i}^{0} \tilde{x}_{a i}^{0} \tag{7.5.80}
\end{equation*}
$$

where $\alpha_{a}^{0}$ is a positive scalar, whose value is to be determined later, and each $P_{a i}^{0}$ is the unique solution to the Lyapunov equation,

$$
\begin{equation*}
\tilde{J}_{i}(\varepsilon)^{\prime} P_{a i}^{0}+P_{a i}^{0} \tilde{J}_{i}(\varepsilon)=-I \tag{7.5.81}
\end{equation*}
$$

which, by Lemma 7.5.2, is independent of $\varepsilon$. Similarly, for the subsystem $\tilde{x}_{c}$, choose a Lyapunov function,

$$
\begin{equation*}
V_{c}\left(\tilde{x}_{c}\right)=\tilde{x}_{c}^{\prime} P_{c} \tilde{x}_{c} \tag{7.5.82}
\end{equation*}
$$

where $P_{c}>0$ is the unique solution to the Lyapunov equation,

$$
\begin{equation*}
\left(A_{c c}^{c}\right)^{\prime} P_{c}+P_{c} A_{c c}^{c}=-I \tag{7.5.83}
\end{equation*}
$$

The existence of such a $P_{c}$ is again guaranteed by the fact that $A_{c c}^{c}$ is asymptotically stable. Finally, for the subsystem of $\tilde{x}_{d}$, choose a Lyapunov function

$$
\begin{equation*}
V_{d}\left(\tilde{x}_{d}\right)=\sum_{i=1}^{m_{d}} \tilde{x}_{i}^{\prime} P_{i} \tilde{x}_{c} \tag{7.5.84}
\end{equation*}
$$

where each $P_{i}$ is the unique solution to the Lyapunov equation

$$
\begin{equation*}
\left(A_{q_{i}}-B_{q_{i}} F_{i}\right)^{\prime} P_{i}+P_{i}\left(A_{q_{i}}-B_{q_{i}} F_{i}\right)=-I . \tag{7.5.85}
\end{equation*}
$$

Once again, the existence of such $P_{i}$ is due to the fact that $A_{q_{i}}-B_{q_{i}} F_{i}$ is asymptotically stable.

We now construct a Lyapunov function for the closed-loop system (7.5.47)(7.5.54) as follows.

$$
\begin{equation*}
V\left(\tilde{x}_{a}^{-}, \tilde{x}_{a b}^{+}, \tilde{x}_{a}^{0}, \tilde{x}_{c}, \tilde{x}_{d}\right)=V_{a}^{-}\left(\tilde{x}_{a}^{-}\right)+\alpha_{a b}^{+} V_{a b}^{+}\left(\tilde{x}_{a b}^{+}\right)+V_{a}^{0}\left(\tilde{x}_{a}^{0}\right)+V_{c}\left(\tilde{x}_{c}\right)+\alpha_{d} V_{d}\left(\tilde{x}_{d}\right) \tag{7.5.86}
\end{equation*}
$$

where $\alpha_{a b}^{+}=2\left\|P_{a}^{-}\right\|^{2}\left(a_{a a b}^{-}\right)^{2}$ and the value of $\alpha_{d}$ is to be determined.
Let us first consider the derivative of $V_{a}^{0}\left(\tilde{x}_{a}^{0}\right)$ along the trajectories of the subsystem $\tilde{x}_{a}^{0}$ and obtain that,

$$
\begin{align*}
\dot{V}_{a}^{0}\left(\tilde{x}_{a}^{0}\right)= & \sum_{i=1}^{l}\left[-\left(\alpha_{a}^{0}\right)^{i-1}\left(\tilde{x}_{a i}^{0}\right)^{\prime} \tilde{x}_{a i}^{0}+2 \sum_{j=i+1}^{l} \frac{\left(\alpha_{a}^{0}\right)^{i-1}}{\varepsilon}\left(\tilde{x}_{a i}^{0}\right)^{\prime} P_{a i}^{0} \tilde{B}_{i j}(\varepsilon) \tilde{x}_{a j}^{0}\right] \\
& +2 \sum_{i=1}^{l} \frac{\left(\alpha_{a}^{0}\right)^{i-1}}{\varepsilon}\left[\left(\tilde{x}_{a i}^{0}\right)^{\prime} P_{a i}^{0} \tilde{L}_{a d}^{0}(\varepsilon) \tilde{h}_{d}+\left(\tilde{x}_{a i}^{0}\right)^{\prime} P_{a i}^{0} \tilde{E}_{a}^{0}(\varepsilon) w\right] \tag{7.5.87}
\end{align*}
$$

Using (7.5.74) it is straightforward to show that there exists an $\alpha_{a}^{0}>0$ such that,

$$
\begin{equation*}
\dot{V}_{a}^{0}\left(\tilde{x}_{a}^{0}\right) \leq-\frac{3}{4}\left\|\tilde{x}_{a}^{0}\right\|^{2}+\frac{\alpha_{1}}{\varepsilon}\left\|\tilde{x}_{a}^{0}\right\| \cdot\left\|\tilde{h}_{d}\right\|+\alpha_{2}\|w\|^{2} \tag{7.5.88}
\end{equation*}
$$

for some nonnegative constants $\alpha_{1}$ and $\alpha_{2}$, independent of $\varepsilon$.
In view of (7.5.88), the derivative of $V$ along the trajectory of the closed-loop system (7.5.47)-(7.5.54) can be evaluated as follows,

$$
\begin{align*}
\dot{V}= & -\left(\tilde{x}_{a}^{-}\right)^{\prime} \tilde{x}_{a}^{-}+2\left(\tilde{x}_{a}^{-}\right)^{\prime} P_{a}^{-} A_{a a b}^{-}(\varepsilon) \tilde{x}_{a b}^{+}+2\left(\tilde{x}_{a}^{-}\right)^{\prime} P_{a}^{-} A_{a a}^{-0}(\varepsilon) \tilde{x}_{a}^{0} \\
& +2\left(\tilde{x}_{a}^{-}\right)^{\prime} P_{a}^{-} L_{a d}^{-} \tilde{h}_{d}+2\left(\tilde{x}_{a}^{0}\right)^{\prime} P_{a}^{-} E_{a}^{-} w-\alpha_{a b}^{+}\left(\tilde{x}_{a b}^{+}\right)^{\prime} \tilde{x}_{a b}^{+} \\
& +2 \alpha_{a b}^{+}\left(x_{a b}^{+}\right)^{\prime} P_{a b}^{+} A_{a b a}^{+0}(\varepsilon) \tilde{x}_{a}^{0}+2 \alpha_{a b}^{+}\left(x_{a b}^{+}\right)^{\prime} P_{a b}^{+} L_{a b d}^{+} \tilde{h}_{d} \\
- & -\frac{3}{4}\left\|\tilde{x}_{a}^{0}\right\|^{2}+\frac{\alpha_{1}}{\varepsilon}\left\|\tilde{x}_{a}^{0}\right\| \cdot\left\|\tilde{h}_{d}\right\|+\alpha_{2}\|w\|^{2}-\tilde{x}_{c}^{\prime} \tilde{x}_{c} \\
+ & +2 \varepsilon \tilde{x}_{c}^{\prime} P_{c}\left[A_{c a b}^{+} \tilde{x}_{a b}^{+}+A_{c a}^{0}(\varepsilon) \tilde{x}_{a}^{0}+L_{c d} \tilde{h}_{d}+E_{c} w\right] \\
& +\alpha_{d} \sum_{i=1}^{m_{d}}\left[-\frac{1}{\varepsilon} \tilde{x}_{i}^{\prime} \tilde{x}_{i}+2 \tilde{x}_{i}^{\prime} P_{i} \tilde{L}_{i a b}^{+}(\varepsilon) \tilde{x}_{a b}^{+}\right. \\
& \left.+2 \tilde{x}_{i}^{\prime} P_{i} \tilde{L}_{i a}^{0}(\varepsilon) \tilde{x}_{a}^{0}+2 \tilde{x}_{i}^{\prime} P_{i} \tilde{L}_{i d}(\varepsilon) \tilde{h}_{d}+2 \tilde{x}_{i}^{\prime} P_{i} \tilde{E}_{i}(\varepsilon) w\right] . \tag{7.5.89}
\end{align*}
$$

Using the majorizations (7.5.69)-(7.5.73) and noting the definition of $\alpha_{a b}^{+}$ (7.5.86), we can easily verify that, there exist an $\alpha_{d}>0$ and an $\varepsilon_{1}^{*} \in(0,1]$ such that, for all $\varepsilon \in\left(0, \varepsilon_{1}^{*}\right]$,

$$
\begin{equation*}
\dot{V} \leq-\frac{1}{2}\left\|\tilde{x}_{a}^{-}\right\|^{2}-\frac{1}{2}\left\|\tilde{x}_{a b}^{+}\right\|^{2}-\frac{1}{2}\left\|\tilde{x}_{a}^{0}\right\|^{2}-\frac{1}{2 \varepsilon}\left\|\tilde{x}_{d}\right\|^{2}+\alpha_{3}\|w\|^{2} \tag{7.5.90}
\end{equation*}
$$

for some positive constant $\alpha_{3}$, independent of $\varepsilon$.

From (7.5.90), it follows that the closed-loop system in the absence of disturbance $w$ is asymptotically stable. It remains to show that, for any given $\gamma>0$, there exists an $\varepsilon^{*} \in\left(0, \varepsilon_{1}^{*}\right]$ such that, for all $\varepsilon \in\left(0, \varepsilon^{*}\right]$,

$$
\begin{equation*}
\|h\|_{2} \leq \gamma\|w\|_{2} \tag{7.5.91}
\end{equation*}
$$

To this end, we integrate both sides of (7.5.90) from 0 to $\infty$. Noting that $V \geq 0$ and $V(t)=0$ at $t=0$, we have,

$$
\begin{equation*}
\left\|\tilde{h}_{d}\right\|_{2} \leq\left(\sqrt{2 \alpha_{3} \varepsilon}\right)\|w\|_{2} \tag{7.5.92}
\end{equation*}
$$

which, when used in (7.5.88), results in,

$$
\begin{equation*}
\left\|\tilde{x}_{a}^{0}\right\|_{2} \leq\left(\sqrt{\frac{2 \alpha_{1}^{2} \alpha_{3}}{\varepsilon}+\alpha_{2}}\right)\|w\|_{2} \tag{7.5.93}
\end{equation*}
$$

Viewing $\tilde{h}_{d}$ as disturbance to the dynamics of $\tilde{x}_{a b}^{+}$also results in,

$$
\begin{equation*}
\left\|\tilde{x}_{a b}^{+}\right\|_{2} \leq\left(\alpha_{4} \sqrt{\varepsilon}\right)\|w\|_{\mathcal{L}_{2}}, \tag{7.5.94}
\end{equation*}
$$

for some positive constant $\alpha_{4}$, independent of $\varepsilon$.
Finally, recalling that

$$
\begin{equation*}
h=\Gamma_{o \mathrm{P}}\left[\tilde{h}_{d}-F_{a b}^{+} \tilde{x}_{a b}^{+}-\tilde{F}_{a d}^{0}(\varepsilon) \tilde{x}_{a}^{0}\right], \tag{7.5.95}
\end{equation*}
$$

where

$$
\tilde{F}_{a d}^{0}(\varepsilon)=\left[\begin{array}{c}
\tilde{F}_{a d 1}^{0}(\varepsilon)  \tag{7.5.96}\\
\tilde{F}_{a d 2}^{0}(\varepsilon) \\
\vdots \\
F_{a d m_{d}}^{0}(\varepsilon)
\end{array}\right],
$$

with each $\tilde{F}_{a d i}(\varepsilon)$ satisfying (7.5.71) and (7.5.72), we have,

$$
\begin{equation*}
\|h\|_{2} \leq\left\|\Gamma_{o \mathrm{P}}\right\|\left(\sqrt{2 \alpha_{3} \varepsilon}+\alpha_{4}\left\|F_{a b}^{+}\right\| \sqrt{\varepsilon}+\alpha_{5} \sqrt{2 \alpha_{1}^{2} \alpha_{3} \varepsilon+\alpha_{2} \varepsilon^{2}}\right)\|w\|_{2} \tag{7.5.97}
\end{equation*}
$$

for some positive constant $\alpha_{5}$ independent of $\varepsilon$.
To complete the proof, we choose $\varepsilon^{*} \in\left(0, \varepsilon_{1}^{*}\right]$ such that,

$$
\begin{equation*}
\left\|\Gamma_{o \mathrm{o}}\right\|\left(\sqrt{2 \alpha_{3} \varepsilon}+\alpha_{4}\left\|F_{a b}^{+}\right\| \sqrt{\varepsilon}+\alpha_{5} \sqrt{2 \alpha_{1}^{2} \alpha_{3} \varepsilon+\alpha_{2} \varepsilon^{2}}\right) \leq \gamma \tag{7.5.98}
\end{equation*}
$$

For use in the proof of measurement feedback results, it is straightforward to verify from the closed-loop equations (7.5.47)-(7.5.54) that the transfer function from $E_{a}^{0} w$ to $h$ is given by

$$
\begin{equation*}
T_{a o}^{0}(s)=T_{a o}(s, \varepsilon)\left[s I-A_{a a}^{0}+B_{a}^{0} F_{a}^{0}(\varepsilon)\right]^{-1} \tag{7.5.99}
\end{equation*}
$$

where $T_{a o}(s, \varepsilon) \rightarrow 0$ pointwise in $s$ as $\varepsilon \rightarrow 0$.

### 7.5.B. Proof of Theorem 7.4.1

It is trivial to show the stability of the closed-loop system comprising the given plant (7.1.1) and the full order output feedback controller (7.4.3). The closedloop poles are given by $\lambda\left\{A+B F_{\mathrm{P}}(\varepsilon)\right\}$, which are in $\mathbb{C}^{-}$for sufficiently small $\varepsilon$ as shown in Theorem 7.3.1, and $\lambda\left\{A+K_{\mathrm{Q}}(\varepsilon) C_{1}\right\}$, which can be dually shown to be in $\mathbb{C}^{-}$for sufficiently small $\varepsilon$ as well. In what follows, we will show that the full order output feedback controller achieves the $H_{\infty}$-ADDPMS for (7.1.1), which satisfies all 5 conditions of Theorem 7.2.1. Without loss of any generality but for simplicity of presentation, hereafter we assume throughout the rest of the proof that the subsystem $\Sigma_{\mathrm{P}}$, i.e., the quadruple $\left(A, B, C_{2}, D_{2}\right)$, has already been transformed into the special coordinate basis as given in Theorem 2.3.1. To be more specific, we have

$$
\begin{align*}
& A=B_{0} C_{2,0}+\left[\begin{array}{cccccc}
A_{a a}^{-} & 0 & 0 & L_{a b}^{-} C_{b} & 0 & L_{a d}^{-} C_{d} \\
0 & A_{a a}^{0} & 0 & L_{a b}^{0} C_{b} & 0 & L_{a d}^{0} C_{d} \\
0 & 0 & A_{a a}^{+} & L_{a b}^{+} C_{b} & 0 & L_{a d}^{+} C_{d} \\
0 & 0 & 0 & A_{b b} & 0 & L_{b d} C_{d} \\
B_{c} E_{c a}^{-} & B_{c} E_{c a}^{0} & B_{c} E_{c a}^{+} & L_{c b} C_{b} & A_{c c} & L_{c d} C_{d} \\
B_{d} E_{d a}^{-} & B_{d} E_{d a}^{0} & B_{d} E_{d a}^{+} & B_{d} E_{d b} & B_{d} E_{d c} & A_{d d}
\end{array}\right] \\
& :=B_{0} C_{2,0}+\tilde{A},  \tag{7.5.100}\\
& B=\left[\begin{array}{ccc}
B_{0 a}^{-} & 0 & 0 \\
B_{0 a}^{0} & 0 & 0 \\
B_{0 a}^{+} & 0 & 0 \\
B_{0 b} & 0 & 0 \\
B_{0 c} & 0 & B_{c} \\
B_{0 d} & B_{d} & 0
\end{array}\right], \quad B_{0}=\left[\begin{array}{c}
B_{0 a}^{-} \\
B_{0 a}^{0} \\
B_{0 a}^{+} \\
B_{0 b} \\
B_{0 c} \\
B_{0 d}
\end{array}\right],  \tag{7.5.101}\\
& C_{2,0}=\left[\begin{array}{llllll}
C_{0 a}^{-} & C_{0 a}^{0} & C_{0 a}^{+} & C_{0 b} & C_{0 c} & C_{0 d}
\end{array}\right],  \tag{7.5.102}\\
& C_{2}=\left[\begin{array}{cccccc}
C_{0 a}^{-} & C_{0 a}^{0} & C_{0 a}^{+} & C_{0 b} & C_{0 c} & C_{0 d} \\
0 & 0 & 0 & 0 & 0 & C_{d} \\
0 & 0 & 0 & C_{b} & 0 & 0
\end{array}\right], \quad D_{2}=\left[\begin{array}{ccc}
I & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right] \text {, }  \tag{7.5.103}\\
& \mathcal{S}^{+}\left(\Sigma_{\mathrm{P}}\right)=\operatorname{Im}\left\{\left[\begin{array}{llll}
I & 0 & 0 & 0 \\
0 & I & 0 & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & I & 0 \\
0 & 0 & 0 & I
\end{array}\right]\right\} . \tag{7.5.104}
\end{align*}
$$

It is simple to note that Condition 4 of Theorem 7.2.1 implies that

$$
E=\left[\begin{array}{c}
E_{a}^{-}  \tag{7.5.105}\\
E_{a}^{0} \\
0 \\
0 \\
E_{c} \\
E_{d}
\end{array}\right] .
$$

Next, for any $\zeta \in \mathcal{V}_{\lambda}\left(\Sigma_{\mathbf{Q}}\right)$ with $\lambda \in \mathbb{C}^{0}$, we partition $\zeta$ as follows,

$$
\zeta=\left(\begin{array}{c}
\zeta_{a}^{-}  \tag{7.5.106}\\
\zeta_{a}^{0} \\
\zeta_{a}^{+} \\
\zeta_{b} \\
\zeta_{c} \\
\zeta_{d}
\end{array}\right)
$$

Then, Condition 5 of Theorem 7.2.1 implies that $C_{2} \zeta=0$, or equivalently

$$
\begin{equation*}
C_{2,0} \zeta=0, \quad C_{b} \zeta_{b}=0 \quad \text { and } \quad C_{d} \zeta_{d}=0 \tag{7.5.107}
\end{equation*}
$$

By Definition 2.3.3, we have

$$
\left[\begin{array}{cc}
A-\lambda I & E  \tag{7.5.108}\\
C_{1} & D_{1}
\end{array}\right]\binom{\zeta}{\eta}=0
$$

for some appropriate vector $\eta$. Clearly, (7.5.108) and (7.5.105) imply that

$$
(A-\lambda I) \zeta=-E \eta=\left(\begin{array}{c}
\star  \tag{7.5.109}\\
\star \\
0 \\
0 \\
\star \\
\star
\end{array}\right)
$$

where ${ }^{*}$ 's are some vectors of not much interests. Note that (7.5.107) implies

$$
\begin{aligned}
&(A-\lambda I) \zeta=\left(B_{0} C_{2,0}+\tilde{A}-\lambda I\right) \zeta=(\tilde{A}-\lambda I) \zeta \\
& \star \\
&=\left[\begin{array}{c}
\star \\
\left(A_{a a}^{+}-\lambda I\right) \zeta_{a}^{+}+L_{a b}^{+} C_{b} \zeta_{b}+L_{a d}^{+} C_{d} \zeta_{d} \\
\left(A_{b b}-\lambda I\right) \zeta_{b}+L_{b d} C_{d} \zeta_{d} \\
\star \\
\star
\end{array}\right]
\end{aligned}
$$

$$
=\left[\begin{array}{c}
\star  \tag{7.5.110}\\
\star \\
\left(A_{a a}^{+}-\lambda I\right) \zeta_{a}^{+} \\
\left(A_{b b}-\lambda I\right) \zeta_{b} \\
\star \\
\star
\end{array}\right] .
$$

(7.5.109) and (7.5.110) imply

$$
\begin{equation*}
\left(A_{a a}^{+}-\lambda I\right) \zeta_{a}^{+}=0 \quad \text { and } \quad\left(A_{b b}-\lambda I\right) \zeta_{b}=0 \tag{7.5.111}
\end{equation*}
$$

Since $A_{a a}^{+}$is unstable, $\left(A_{a a}^{+}-\lambda I\right) \zeta_{a}^{+}=0$ implies that $\zeta_{a}^{+}=0$. Similarly, since $\left(A_{b b}, C_{b}\right)$ is completely observable, $\left(A_{b b}-\lambda I\right) \zeta_{b}=0$ and $C_{b} \zeta_{b}=0$ imply $\zeta_{b}=0$. Thus, $\zeta$ has the following property,

$$
\zeta=\left(\begin{array}{c}
\zeta_{a}^{-}  \tag{7.5.112}\\
\zeta_{a}^{0} \\
0 \\
0 \\
\zeta_{c} \\
\zeta_{d}
\end{array}\right) \in \mathcal{S}^{+}\left(\Sigma_{\mathrm{P}}\right)
$$

Obviously, (7.5.112) together with Condition 6 of Theorem 7.2.1 imply

$$
\begin{equation*}
\mathcal{S}^{+}\left(\Sigma_{\mathrm{P}}\right) \supset \mathcal{V}^{+}\left(\Sigma_{\mathrm{Q}}\right) \cup\left\{\cup_{\lambda \in \mathbb{C}^{0}} \mathcal{V}_{\lambda}\left(\Sigma_{Q}\right)\right\} \tag{7.5.113}
\end{equation*}
$$

Next, it is straightforward to verify that $A-s I$ can be partitioned as

$$
\begin{equation*}
A-s I=X_{1}+X_{2} C_{2}+X_{3}+X_{4}, \tag{7.5.114}
\end{equation*}
$$

where

$$
\begin{gather*}
X_{1}:=\left[\begin{array}{cccccc}
A_{a a}^{-}-s I & 0 & 0 & L_{a b}^{-} C_{b} & 0 & L_{a d}^{-} C_{d} \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
B_{c} E_{c a}^{-} & B_{c} E_{c a}^{0} & B_{c} E_{c a}^{+} & L_{c b} C_{b} & A_{c c}-s I & L_{c d} C_{d} \\
B_{d} E_{d a}^{-} & B_{d} E_{d a}^{0} & B_{d} E_{d a}^{+} & B_{d} E_{d b} & B_{d} E_{d c} & A_{d d}-s I
\end{array}\right],  \tag{7.5.115}\\
X_{2}=\left[\begin{array}{ccc}
B_{0 a}^{-} & 0 & 0 \\
B_{0 a}^{0} & L_{a d}^{0} & L_{a b}^{0} \\
B_{0 a}^{+} & L_{a d}^{+} & L_{a b}^{+} \\
B_{0 b} & L_{b d} & 0 \\
B_{0 c} & 0 & 0 \\
B_{0 d} & 0 & 0
\end{array}\right], \tag{7.5.116}
\end{gather*}
$$

$$
X_{3}=\left[\begin{array}{cccccc}
0 & 0 & 0 & 0 & 0 & 0  \tag{7.5.117}\\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & A_{a a}^{+}-s I & 0 & 0 & 0 \\
0 & 0 & 0 & A_{b b}-s I & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0
\end{array}\right]
$$

and

$$
X_{4}=\left[\begin{array}{cccccc}
0 & 0 & 0 & 0 & 0 & 0  \tag{7.5.118}\\
0 & A_{a a}^{0}-s I & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0
\end{array}\right]
$$

It is simple to see that

$$
\begin{equation*}
\operatorname{Im}\left(X_{1}\right) \subset \mathcal{S}^{+}\left(\Sigma_{\mathrm{P}}\right) \cap\left\{\cap_{\lambda \in \mathbb{C}^{0}} \mathcal{S}_{\lambda}\left(\Sigma_{\mathrm{P}}\right)\right\} \tag{7.5.119}
\end{equation*}
$$

and

$$
\begin{equation*}
\operatorname{Ker}\left(X_{3}\right) \supset \mathcal{S}^{+}\left(\Sigma_{\mathbf{P}}\right) \supset \mathcal{V}^{+}\left(\Sigma_{\mathbf{Q}}\right) \cup\left\{\cup_{\lambda \in \mathbb{C}^{0}} \mathcal{V}_{\lambda}\left(\Sigma_{Q}\right)\right\} \tag{7.5.120}
\end{equation*}
$$

It follows from the proof of Theorem 7.3.1 that as $\varepsilon \rightarrow 0$

$$
\begin{equation*}
\left\|\left[C_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[s I-A-B F_{\mathrm{P}}(\varepsilon)\right]^{-1}\right\|_{\infty}<\kappa_{\mathrm{P}} \tag{7.5.121}
\end{equation*}
$$

where $\kappa_{\mathrm{P}}$ is a finite positive constant and is independent of $\varepsilon$. Moreover, under Condition 4 of Theorem 7.2.1, we have

$$
\begin{equation*}
\left[C_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[s I-A-B F_{\mathrm{P}}(\varepsilon)\right]^{-1} E \rightarrow 0 \tag{7.5.122}
\end{equation*}
$$

and

$$
\begin{equation*}
\left[C_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[s I-A-B F_{\mathrm{P}}(\varepsilon)\right]^{-1} X_{1} \rightarrow 0 \tag{7.5.123}
\end{equation*}
$$

pointwise in $s$ as $\varepsilon \rightarrow 0$. By (7.5.99), we have

$$
\begin{equation*}
\left[C_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[s I-A-B F_{\mathrm{P}}(\varepsilon)\right]^{-1} X_{4} \rightarrow 0 \tag{7.5.124}
\end{equation*}
$$

pointwise in $s$ as $\varepsilon \rightarrow 0$. Dually, one can show that

$$
\begin{equation*}
\left\|\left[s I-A-K_{\mathrm{Q}}(\varepsilon) C_{1}\right]^{-1}\left[E+K_{\mathrm{Q}}(\varepsilon) D_{1}\right]\right\|_{\infty}<\kappa_{\mathrm{Q}} \tag{7.5.125}
\end{equation*}
$$

where $\kappa_{Q}$ is a finite positive constant and is independent of $\varepsilon$. If Condition 5 of Theorem 7.2.1 is satisfied, the following results hold,

$$
\begin{equation*}
C_{2}\left[s I-A-K_{\mathrm{Q}}(\varepsilon) C_{1}\right]^{-1}\left[E+K_{\mathrm{Q}}(\varepsilon) D_{1}\right] \rightarrow 0 \tag{7.5.126}
\end{equation*}
$$

and

$$
\begin{equation*}
X_{3}\left[s I-A-K_{Q}(\varepsilon) C_{1}\right]^{-1}\left[E+K_{Q}(\varepsilon) D_{1}\right] \rightarrow 0, \tag{7.5.127}
\end{equation*}
$$

pointwise in $s$ as $\varepsilon \rightarrow 0$.

Finally, it is simple to verify that the closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$ under the full order output feedback controller (7.4.3) is given by

$$
\begin{aligned}
T_{h w}(s, \varepsilon)= & {\left[C_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[s I-A-B F_{\mathrm{P}}(\varepsilon)\right]^{-1} E } \\
& +C_{2}\left[s I-A-K_{\mathrm{Q}}(\varepsilon) C_{1}\right]^{-1}\left[E+K_{\mathrm{Q}}(\varepsilon) D_{1}\right]+\left[C_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right] \\
& \cdot\left[s I-A-B F_{\mathrm{P}}(\varepsilon)\right]^{-1}(A-s I)\left[s I-A-K_{\mathrm{Q}}(\varepsilon) C_{1}\right]^{-1}\left[E+K_{\mathrm{Q}}(\varepsilon) D_{1}\right] .
\end{aligned}
$$

Using (7.5.114), we can rewrite $T_{h w}(s, \varepsilon)$ as

$$
\begin{aligned}
T_{h w}(s, \varepsilon)= & {\left[C_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[s I-A-B F_{\mathrm{P}}(\varepsilon)\right]^{-1} E } \\
& +C_{2}\left[s I-A-K_{\mathrm{Q}}(\varepsilon) C_{1}\right]^{-1}\left[E+K_{\mathrm{Q}}(\varepsilon) D_{1}\right] \\
& +\left[C_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[s I-A-B F_{\mathrm{P}}(\varepsilon)\right]^{-1}\left(X_{1}+X_{2} C_{2}+X_{3}+X_{4}\right) \\
& \cdot\left[s I-A-K_{\mathrm{Q}}(\varepsilon) C_{1}\right]^{-1}\left[E+K_{\mathrm{Q}}(\varepsilon) D_{1}\right] .
\end{aligned}
$$

Following (7.5.121) to (7.5.127), and some simple manipulations, it is straightforward to show that as $\varepsilon \rightarrow 0, T_{h w}(s, \varepsilon) \rightarrow 0$, pointwise in $s$, which is equivalent to $\left\|T_{h w}\right\|_{\infty} \rightarrow 0$ as $\varepsilon \rightarrow 0$. Hence, the full order output feedback controller (7.4.3) solves the $H_{\infty}$-ADDPMS for the given plant (7.1.1), provided that all five conditions of Theorem 7.2.1 are satisfied.

### 7.5.C. Proof of Theorem 7.4.2

Again, it is trivial to show the stability of the closed-loop system comprising the given plant (7.1.1) and the reduced order measurement feedback controller (7.4.15) as the closed-loop poles are given by $\lambda\left\{A+B F_{\mathrm{P}}(\varepsilon)\right\}$ and $\lambda\left\{A_{\mathrm{R}}+\right.$ $\left.K_{\mathrm{R}}(\varepsilon) C_{\mathrm{R}}\right\}$, which are asymptotically stable for a sufficiently small $\varepsilon$. Next, it is easy to compute the closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$ under the reduced order output feedback controller,

$$
\begin{aligned}
T_{h w}(s, \varepsilon)= & {\left[C_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[s I-A-B F_{\mathrm{P}}(\varepsilon)\right]^{-1} E } \\
& +\left[C_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[s I-A-B F_{\mathrm{P}}(\varepsilon)\right]^{-1}(A-s I)\binom{0}{I_{n-k}} \\
& \cdot\left[s I-A_{\mathrm{R}}-K_{\mathrm{R}}(\varepsilon) C_{\mathrm{R}}\right]^{-1}\left[E_{\mathrm{R}}+K_{\mathrm{R}}(\varepsilon) D_{\mathrm{R}}\right] \\
& +C_{2}\binom{0}{I_{n-k}}\left[s I-A_{\mathrm{R}}-K_{\mathrm{R}}(\varepsilon) C_{\mathrm{R}}\right]^{-1}\left[E_{\mathrm{R}}+K_{\mathrm{R}}(\varepsilon) D_{\mathrm{R}}\right] .
\end{aligned}
$$

It was shown in Chen [10] (i.e., Proposition 2.2.1) that

$$
\begin{equation*}
\binom{0}{I_{n-k}} \mathcal{V}^{+}\left(\Sigma_{\mathbf{Q R}}\right)=\mathcal{V}^{+}\left(\Sigma_{\mathbf{Q}}\right) \tag{7.5.128}
\end{equation*}
$$

Following the same lines of reasoning as in Chen [10], one can also show that

$$
\begin{equation*}
\binom{0}{I_{n-k}} \cup_{\lambda \in \mathbf{C}^{0}} \mathcal{V}_{\lambda}\left(\Sigma_{Q \mathbb{R}}\right)=\cup_{\lambda \in \mathbb{C}^{0} \mathcal{V}_{\lambda}\left(\Sigma_{Q}\right)} \tag{7.5.129}
\end{equation*}
$$

Hence, we have

$$
\begin{equation*}
\binom{0}{I_{n-k}}\left(\mathcal{V}^{+}\left(\Sigma_{\mathrm{QR}}\right) \cup\left\{\cup_{\lambda \in \mathbf{C}^{0}} \mathcal{V}_{\lambda}\left(\Sigma_{\mathrm{QR}}\right)\right\}\right)=\mathcal{V}^{+}\left(\Sigma_{\mathbf{Q}}\right) \cup\left\{\cup_{\lambda \in \mathbb{C}^{0}} \mathcal{V}_{\lambda}\left(\Sigma_{\mathbf{Q}}\right)\right\} \tag{7.5.130}
\end{equation*}
$$

The rest of the proof follows from the same lines as those of Theorem 7.4.1. 因

## Chapter 8

## Infima in Discrete-time $H_{\infty}$ Optimization

IN THIS CHAPTER, we present computational methods for evaluating the infima of discrete-time $H_{\infty}$ optimal control problems. The main contributions of this chapter are the non-iterative algorithms that compute exactly the values of infima for systems satisfying certain geometric conditions. If these conditions are not satisfied, one might have to use iterative schemes based on certain reduced order systems for approximating these infima. Most of the results of this chapter were reported earlier in Chen [13], and Chen, Guo and Lin [17].

### 8.1. Full Information Feedback Case

The main result of this section deals with the non-iterative computation of the infimum for the following full information feedback discrete-time system characterized by:

$$
\left\{\begin{align*}
& x(k+1)=A x(k)+B u(k)+E w(k)  \tag{8.1.1}\\
& y(k)=\binom{I}{0} x(k) \\
& h(k)=C_{2} x(k)+D_{2} u(k)+D_{22} w(k) \\
& I
\end{align*}\right) w(k),
$$

where $x \in \mathbb{R}^{n}$ is the state, $u \in \mathbb{R}^{m}$ is the control input, $w \in \mathbb{R}^{q}$ is the external disturbance input, $y \in \mathbb{R}^{n+q}$ is the measurement output, and $h \in \mathbb{R}^{\ell}$ is the controlled output of $\Sigma$. For ease of reference in future development, we define $\Sigma_{\mathrm{P}}$ to be the subsystem characterized by the matrix quadruple $\left(A, B, C_{2}, D_{2}\right)$. We first make the following assumptions:

Assumption 8.F.1: $(A, B)$ is stabilizable;
Assumption 8.F.2: $\Sigma_{\mathrm{P}}$ has no invariant zero on the unit circle;
Assumption 8.F.3: $\operatorname{Im}(E) \subset \mathcal{V}^{\odot}\left(\Sigma_{\mathbf{P}}\right)+\mathcal{S}^{\odot}\left(\Sigma_{\mathrm{P}}\right)$; and
Assumption 8.F.4: $D_{22}=0$.
®
In what follows, we state a step-by-step algorithm for the computation of the infimum $\gamma^{*}$.

Step 8.F.1: Without loss of generality but for simplicity of presentation, we assume that the quadruple $\left(A, B, C_{2}, D_{2}\right)$, i.e., $\Sigma_{\mathrm{P}}$, has been partitioned in the form of (2.3.4). Then, transform $\Sigma_{\mathrm{P}}$ into the special coordinate basis as described in Chapter 2 (see also (2.3.20) to (2.3.23) for the compact form of the special coordinate basis). In this algorithm, for ease of reference in future development, we introduce an additional permutation matrix to the state transformation $\Gamma_{s}$ such that the new state variables are ordered as follows:

$$
\tilde{x}=\left(\begin{array}{c}
x_{c}  \tag{8.1.2}\\
x_{a}^{-} \\
x_{a}^{+} \\
x_{d} \\
x_{b}
\end{array}\right)
$$

Next, we compute

$$
\Gamma_{s}^{-1} E=\left[\begin{array}{c}
E_{c}  \tag{8.1.3}\\
E_{a}^{-} \\
E_{a}^{+} \\
E_{d} \\
E_{b}
\end{array}\right]
$$

Note that Assumption 8.F. 3 is equivalent to $E_{b}=0$. Also, for economy of notation, we denote $n_{x}$ the dimension of $\mathbb{R}^{n} / \mathcal{V}^{\ominus}\left(\Sigma_{\mathrm{P}}\right)$, which is equivalent to $n_{x}=n_{a}^{+}+n_{d}+n_{b}$. We note that $n_{x}=0$ if and only if the system $\Sigma_{\mathrm{P}}$ is right invertible and is of minimum phase with no infinite zero of order higher than zero.

Step 8.F.2: Define $A_{x}, B_{x}, B_{x 0}, B_{x 1}, E_{x}, C_{x}$ and $D_{x}$ as follows:

$$
\begin{gather*}
A_{x}:=\left[\begin{array}{ccc}
A_{a a}^{+} & L_{a d}^{+} C_{d} & L_{a b}^{+} C_{b} \\
B_{d} E_{d a}^{+} & A_{d d} & B_{d} E_{d b} \\
0 & L_{b d} C_{d} & A_{b b}
\end{array}\right], E_{x}:=\left[\begin{array}{c}
E_{a}^{+} \\
E_{d} \\
E_{b}
\end{array}\right],  \tag{8.1.4}\\
B_{x}:=\left[\begin{array}{ll}
B_{x 0} & B_{x 1}
\end{array}\right]:=\left[\begin{array}{cc}
B_{0 a}^{+} & 0 \\
B_{0 d} & B_{d} \\
B_{0 b} & 0
\end{array}\right], \tag{8.1.5}
\end{gather*}
$$

and

$$
C_{x}:=\Gamma_{o}\left[\begin{array}{ccc}
0 & 0 & 0  \tag{8.1.6}\\
0 & C_{d} & 0 \\
0 & 0 & C_{b}
\end{array}\right], \quad D_{x}=\Gamma_{o}\left[\begin{array}{ll}
I & 0 \\
0 & 0 \\
0 & 0
\end{array}\right] .
$$

It follows from the property of the special coordinate basis that the pair $\left(A_{x}, B_{x}\right)$ is stabilizable. Next, we find a matrix $F_{x}$ such that $A_{x}+B_{x} F_{x}$ has no eigenvalue at -1 . Then define $\tilde{A}_{x}, \tilde{B}_{x}, \tilde{E}_{x}, \tilde{C}_{x}, \tilde{D}_{x}$ and $\tilde{D}_{22}$ as:

$$
\begin{align*}
\tilde{A}_{x} & :=\left(A_{x}+B_{x} F_{x}+I\right)^{-1}\left(A_{x}+B_{x} F_{x}-I\right) \\
\tilde{B}_{x} & :=2\left(A_{x}+B_{x} F_{x}+I\right)^{-2} B_{x} \\
\tilde{E}_{x} & :=2\left(A_{x}+B_{x} F_{x}+I\right)^{-2} E_{x} \\
\tilde{C}_{x} & :=C_{x}+D_{x} F_{x}  \tag{8.1.7}\\
\tilde{D}_{x} & :=D_{x}-\left(C_{x}+D_{x} F_{x}\right)\left(A_{x}+B_{x} F_{x}+I\right)^{-1} B_{x} \\
\tilde{D}_{22} & :=D_{22}-\left(C_{x}+D_{x} F_{x}\right)\left(A_{x}+B_{x} F_{x}+I\right)^{-1} E_{x}
\end{align*}
$$

Step 8.F.3: Solve the following continuous-time algebraic Riccati equation and algebraic Lyapunov equation, both independent of $\gamma$ :

$$
\begin{align*}
0= & {\left[\tilde{A}_{x}-\tilde{B}_{x}\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{D}_{x}^{\prime} \tilde{C}_{x}\right] \tilde{S}_{x}+\tilde{S}_{x}\left[\tilde{A}_{x}-\tilde{B}_{x}\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{D}_{x}^{\prime} \tilde{C}_{x}\right]^{\prime} } \\
& -\tilde{B}_{x}\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{B}_{x}^{\prime}+\tilde{S}_{x}\left[\tilde{C}_{x}^{\prime} \tilde{C}_{x}-\tilde{C}_{x}^{\prime} \tilde{D}_{x}\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{D}_{x}^{\prime} \tilde{C}_{x}\right] \tilde{S}_{x}  \tag{8.1.8}\\
0= & {\left[\tilde{A}_{x}-\tilde{B}_{x}\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{D}_{x}^{\prime} \tilde{C}_{x}\right] \tilde{T}_{x}+\tilde{T}_{x}\left[\tilde{A}_{x}-\tilde{B}_{x}\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{D}_{x}^{\prime} \tilde{C}_{x}\right]^{\prime} } \\
& -\left[\tilde{E}_{x}-\tilde{B}_{x}\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{D}_{x}^{\prime} \tilde{D}_{22}\right]\left[\tilde{E}_{x}-\tilde{B}_{x}\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{D}_{x}^{\prime} \tilde{D}_{22}\right]^{\prime}, \tag{8.1.9}
\end{align*}
$$

for positive definite solution $\tilde{S}_{x}$ and positive semi-definite solution $\tilde{T}_{x}$. For future use, we define

$$
\begin{equation*}
S_{x}:=\left(A_{x}+B_{x} F_{x}+I\right) \tilde{S}_{x}\left(A_{x}^{\prime}+F_{x}^{\prime} B_{x}^{\prime}+I\right) / 2 \tag{8.1.10}
\end{equation*}
$$

and

$$
\begin{equation*}
T_{x}:=\left(A_{x}+B_{x} F_{x}+I\right) \tilde{T}_{x}\left(A_{x}^{\prime}+F_{x}^{\prime} B_{x}^{\prime}+I\right) / 2 \tag{8.1.11}
\end{equation*}
$$

Step 8.F.4: The infimum, $\gamma^{*}$, is given by

$$
\begin{equation*}
\gamma^{*}=\sqrt{\lambda_{\max }\left(\tilde{T}_{x} \tilde{S}_{x}^{-1}\right)}=\sqrt{\lambda_{\max }\left(T_{x} S_{x}^{-1}\right)} \tag{8.1.12}
\end{equation*}
$$

This completes the algorithm for computing $\gamma^{*}$ for the full information feedback case.

因
We have the following theorem.

Theorem 8.1.1. Consider the full information system given by (8.1.1). Then under Assumptions 8.F.1 to 8.F.4,

1. $\gamma^{*}$ given by (8.1.12) is indeed its infimum, and
2. for $\gamma>\gamma^{*}$, the positive semi-definite matrix $P(\gamma)$ given by

$$
P(\gamma)=\left(\Gamma_{s}^{-1}\right)^{\prime}\left[\begin{array}{cc}
0 & 0  \tag{8.1.13}\\
0 & \left(S_{x}-T_{x} / \gamma^{2}\right)^{-1}
\end{array}\right] \Gamma_{s}^{-1}
$$

is the unique solution that satisfies conditions 2.(a)-2.(c) of Theorem 3.2.1. Moreover, such a solution $P(\gamma)$ does not exist when $\gamma<\gamma^{*}$.

Proof. First, we note that it follows from Theorem 2.3.1 and Property 2.3.4 of Chapter 2 that $\left(A_{x}, B_{x}, C_{x}, D_{x}\right)$ is left invertible with no invariant zeros on the unit circle. Following the results of Stoorvogel et al [102] and Lemma 4.2.3, it is straightforward to show that the following three statements are equivalent:

1. There exists a $\gamma$ suboptimal controller for the full information system (8.1.1).
2. There exists a $\gamma$ suboptimal controller for the following auxiliary system

$$
\left\{\begin{array}{rl}
x_{x}(k+1) & =A_{x} \quad x_{x}(k)+B_{x} u_{x}(k)+E_{x} \quad w_{x}(k)  \tag{8.1.14}\\
y_{x}(k) & =\binom{0}{I} x_{x}(k) \\
& +\binom{I}{0} w_{x}(k) \\
h_{x}(k) & =C_{x}
\end{array} x_{x}(k)+D_{x} u_{x}(k)+D_{22} w_{x}(k), ~ \$\right.
$$

where $A_{x}, B_{x}, E_{x}, C_{x}$ and $D_{x}$ are defined as in (8.1.4) to (8.1.6). Note that $D_{22}=0$ by the assumption.
3. There exists a $\gamma$ suboptimal controller for the following auxiliary system

$$
\left\{\begin{array}{l}
\dot{\tilde{x}}_{x}=\tilde{A}_{x} \quad \tilde{x}_{x}+\tilde{B}_{x} \tilde{u}_{x}+\tilde{E}_{x} \quad \tilde{w}_{x}  \tag{8.1.15}\\
\tilde{y}_{x}=\binom{0}{I} \tilde{x}_{x}+\binom{I}{0} \tilde{w}_{x} \\
\tilde{h}_{x}=\tilde{C}_{x} \quad \tilde{x}_{x}+\tilde{D}_{x} \tilde{u}_{x}+\tilde{D}_{22} \quad \tilde{w}_{x}
\end{array}\right.
$$

where $\tilde{A}_{x}, \tilde{B}_{x}, \tilde{E}_{x}, \tilde{C}_{x}, \tilde{D}_{x}$ and $\tilde{D}_{22}$ are as defined in (8.1.7).
For future use, we denote $\Sigma_{x}$ and $\tilde{\Sigma}_{x}$ the matrix quadruples $\left(A_{x}, B_{x}, C_{x}, D_{x}\right)$ and $\left(\tilde{A}_{x}, \tilde{B}_{x}, \tilde{C}_{x}, \tilde{D}_{x}\right)$, respectively. Note that by Theorems 3.1.1 and 3.2.1, items 2 and 3 above are also equivalent to the following:

1. There exists a solution $P_{x}>0$ to the following discrete-time algebraic Riccati equation,

$$
P_{x}=A_{x}^{\prime} P_{x} A_{x}+C_{x}^{\prime} C_{x}-\left[\begin{array}{c}
B_{x}^{\prime} P_{x} A_{x}+D_{x}^{\prime} C_{x}  \tag{8.1.16}\\
E_{x}^{\prime} P_{x} A_{x}
\end{array}\right]^{\prime} G_{x}^{-1}\left[\begin{array}{c}
B_{x}^{\prime} P_{x} A_{x}+D_{x}^{\prime} C_{x} \\
E_{x}^{\prime} P_{x} A_{x}
\end{array}\right]
$$

where

$$
G_{x}:=\left[\begin{array}{cc}
D_{x}^{\prime} D_{x}+B_{x}^{\prime} P_{x} B_{x} & B_{x}^{\prime} P_{x} E_{x}  \tag{8.1.17}\\
E_{x}^{\prime} P_{x} B_{x} & E_{x}^{\prime} P_{x} E_{x}-\gamma^{2} I
\end{array}\right]
$$

such that the following conditions are satisfied

$$
\begin{align*}
V_{x} & :=B_{x}^{\prime} P_{x} B_{x}+D_{x}^{\prime} D_{x}>0  \tag{8.1.18}\\
R_{x} & :=\gamma^{2} I-E_{x}^{\prime} P_{x} E_{x}+E_{x}^{\prime} P_{x} B_{x} V_{x}^{-1} B_{x}^{\prime} P_{x} E_{x}>0 . \tag{8.1.19}
\end{align*}
$$

2. There exists a solution $\tilde{P}_{x}>0$ to the following continuous-time algebraic Riccati equation,

$$
0=\tilde{P}_{x} \tilde{A}_{x}+\tilde{A}_{x}^{\prime} \tilde{P}_{x}+\tilde{C}_{x}^{\prime} \tilde{C}_{x}-\left[\begin{array}{c}
\tilde{B}_{x}^{\prime} \tilde{P}_{x}+\tilde{D}_{x}^{\prime} \tilde{C}_{x}  \tag{8.1.20}\\
\tilde{E}_{x}^{\prime} \tilde{P}_{x}+\tilde{D}_{22}^{\prime} \tilde{C}_{x}
\end{array}\right]^{\prime} \tilde{G}_{x}^{-1}\left[\begin{array}{c}
\tilde{B}_{x}^{\prime} \tilde{P}_{x}+\tilde{D}_{x}^{\prime} \tilde{C}_{x} \\
\tilde{E}_{x}^{\prime} \tilde{P}_{x}+\tilde{D}_{22}^{\prime} \tilde{C}_{x}
\end{array}\right]
$$

with

$$
\begin{equation*}
\tilde{D}_{22}^{\prime}\left[I-\tilde{D}_{x}\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{D}_{x}^{\prime}\right] \tilde{D}_{22}<\gamma^{2} I \tag{8.1.21}
\end{equation*}
$$

and

$$
\tilde{G}_{x}:=\left[\begin{array}{cc}
\tilde{D}_{x}^{\prime} \tilde{D}_{x} & \tilde{D}_{x}^{\prime} \tilde{D}_{22}  \tag{8.1.22}\\
\tilde{D}_{22}^{\prime} \tilde{D}_{x} & \tilde{D}_{22}^{\prime} \tilde{D}_{22}-\gamma^{2} I
\end{array}\right]
$$

Furthermore, the solutions to the above Riccati equations, if they exist, are related by

$$
\begin{equation*}
P_{x}=2\left(A_{x}^{\prime}+I\right)^{-1} \tilde{P}_{x}\left(A_{x}+I\right)^{-1} \tag{8.1.23}
\end{equation*}
$$

Thus, it is equivalent to show that $\gamma^{*}$ given by (8.1.12) is the infimum for the full information system (8.1.1) by showing that it is an infimum for the auxiliary system in (8.1.15). This can be done by first showing the properties of the auxiliary system of (8.1.15) and then applying the results of Chapter 5. We note that the matrix $F_{x}$ in Step 8.F. 2 of the algorithm is a pre-state feedback gain, which is introduced merely to deal with the situation when $A_{x}$ has eigenvalues at -1 and the inverse of $I+A_{x}$ does not exist. For the sake of simplicity but without loss of generality, we will hereafter assume that $A_{x}$ has no eigenvalue at -1 and $F_{x}=0$. We will first show the following two facts associated with
the auxiliary system (8.1.15): There exists a pre-disturbance feedback to the system in (8.1.15) in the form of,

$$
\begin{equation*}
\tilde{u}_{x}=\tilde{F}_{w} \tilde{w}_{x}+\tilde{v}_{x} \tag{8.1.24}
\end{equation*}
$$

such that

1. $\tilde{D}_{22}+\tilde{D}_{x} \tilde{F}_{w}=0$, and
2. $\operatorname{Im}\left(\tilde{E}_{x}+\tilde{B}_{x} \tilde{F}_{w}\right) \subseteq \mathcal{V}^{\odot}\left(\tilde{\Sigma}_{x}\right)+\mathcal{S}^{\odot}\left(\tilde{\Sigma}_{x}\right)$.

In fact, we will show that such an $\tilde{F}_{w}$ is given by

$$
\begin{equation*}
\tilde{F}_{w}=-\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{D}_{x}^{\prime} \tilde{D}_{22} \tag{8.1.25}
\end{equation*}
$$

In order to make our proof simpler, we first apply a pre-state feedback law

$$
u_{x}=F_{x} x_{x}+v_{x}=-\left[\begin{array}{ccc}
0 & 0 & 0  \tag{8.1.26}\\
E_{d a}^{+} & 0 & E_{d b}
\end{array}\right] x_{x}+v_{x}
$$

to the system in (8.1.14) such that the resulting dynamic matrix $A_{x}+B_{x} F_{x}$ has the following format,

$$
\left[\begin{array}{ccc}
A_{a a}^{+} & L_{a d}^{+} C_{d} & L_{a b}^{+} C_{b}  \tag{8.1.27}\\
0 & A_{d d} & 0 \\
0 & L_{b d} C_{d} & A_{b b}
\end{array}\right]
$$

while the rest of the system matrices in (8.1.14) remain unchanged. Hence, it is without loss of generality that we assume that $A_{x}$ is already in the form of (8.1.27). Also, we assume that both $A_{d d}$ and $A_{b b}$ have no eigenvalue at -1 . Then it is simple to verify that
$\left(A_{x}+I\right)^{-1}=\left[\begin{array}{ccc}\left(A_{a a}^{+}+I\right)^{-1} & X_{1} & X_{2} \\ 0 & \left(A_{d d}+I\right)^{-1} & 0 \\ 0 & -\left(A_{b b}+I\right)^{-1} L_{b d} C_{d}\left(A_{d d}+I\right)^{-1} & \left(A_{b b}+I\right)^{-1}\end{array}\right]$,
where

$$
\begin{align*}
& X_{1}=-\left(A_{a a}^{+}+I\right)^{-1}\left[L_{a d}^{+}-L_{a b}^{+} C_{b}\left(A_{b b}+I\right)^{-1} L_{b d}\right] C_{d}\left(A_{d d}+I\right)^{-1}  \tag{8.1.29}\\
& X_{2}=-\left(A_{a a}^{+}+I\right)^{-1} L_{a b}^{+} C_{b}\left(A_{b b}+I\right)^{-1} \tag{8.1.30}
\end{align*}
$$

and

$$
\begin{aligned}
\tilde{D}_{x} & =D_{x}-C_{x}\left(A_{x}+I\right)^{-1} B_{x} \\
& =\Gamma_{o}\left[\begin{array}{cc}
I & 0 \\
-C_{d}\left(A_{d d}+I\right)^{-1} B_{0 d} & -C_{d}\left(A_{d d}+I\right)^{-1} B_{d} \\
X_{3} & C_{b}\left(A_{b b}+I\right)^{-1} L_{b d} C_{d}\left(A_{d d}+I\right)^{-1} B_{d}
\end{array}\right]
\end{aligned}
$$

where

$$
\begin{equation*}
X_{3}=C_{b}\left(A_{b b}+I\right)^{-1} L_{b d} C_{d}\left(A_{d d}+I\right)^{-1} B_{0 d}-C_{b}\left(A_{b b}+I\right)^{-1} B_{0 b} \tag{8.1.31}
\end{equation*}
$$

Define

$$
\tilde{\Gamma}_{o}=\Gamma_{o}\left[\begin{array}{ccc}
I & 0 & 0  \tag{8.1.32}\\
-C_{d}\left(A_{d d}+I\right)^{-1} B_{0 d} & -C_{d}\left(A_{d d}+I\right)^{-1} B_{d} & 0 \\
X_{3} & C_{b}\left(A_{b b}+I\right)^{-1} L_{b d} C_{d}\left(A_{d d}+I\right)^{-1} B_{d} & I
\end{array}\right] .
$$

We note that $\tilde{\Gamma}_{o}$ is nonsingular. This follows from the property of the special coordinate basis (see Theorem 2.3.1) that the triple $\left(A_{d d}, B_{d}, C_{d}\right)$ is square and invertible with no invariant zero, and hence $C_{d}\left(A_{d d}+I\right)^{-1} B_{d}$ is nonsingular. Then we have

$$
\tilde{D}_{x}=\tilde{\Gamma}_{o}\left[\begin{array}{ll}
I & 0  \tag{8.1.33}\\
0 & I \\
0 & 0
\end{array}\right],
$$

and
$\tilde{D}_{22}=-C_{x}\left(A_{x}+I\right)^{-1} E_{x}=\left[\begin{array}{c}0 \\ -C_{d}\left(A_{d d}+I\right)^{-1} E_{d} \\ C_{b}\left(A_{b b}+I\right)^{-1} L_{b d} C_{d}\left(A_{d d}+I\right)^{-1} E_{d}\end{array}\right]=\tilde{\Gamma}_{o}\left[\begin{array}{c}0 \\ X_{4} \\ 0\end{array}\right]$,
where

$$
\begin{equation*}
X_{4}=\left[C_{d}\left(A_{d d}+I\right)^{-1} B_{d}\right]^{-1} C_{d}\left(A_{d d}+I\right)^{-1} E_{d} \tag{8.1.35}
\end{equation*}
$$

It is now obvious to see that the following pre-disturbance feedback law to (8.1.15)

$$
\tilde{u}_{x}=\tilde{F}_{w} \tilde{w}_{x}+\tilde{v}_{x}=-\left[\begin{array}{c}
0  \tag{8.1.36}\\
X_{4}
\end{array}\right] \tilde{w}_{x}+\tilde{v}_{x}
$$

guarantees that $\tilde{D}_{22}+\tilde{D}_{x} \tilde{F}_{w}=0$. We also have

$$
\tilde{E}_{x}+\tilde{B} \tilde{F}_{w}=2\left(A_{x}+I\right)^{-2}\left(E_{x}+B_{x} \tilde{F}_{w}\right)=2\left(A_{x}+I\right)^{-2}\left[\begin{array}{c}
E_{a}^{+}  \tag{8.1.37}\\
E_{d}^{*} \\
0
\end{array}\right]
$$

where

$$
\begin{equation*}
E_{d}^{*}=E_{d}-B_{d}\left[C_{d}\left(A_{d d}+I\right)^{-1} B_{d}\right]^{-1} C_{d}\left(A_{d d}+I\right)^{-1} E_{d} \tag{8.1.38}
\end{equation*}
$$

This shows the first fact. Since $\tilde{D}_{x}$ is of maximal column rank, it follows that the above $\tilde{F}_{w}$ is also equivalent to $-\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{D}_{x}^{\prime} \tilde{D}_{22}$. Next, let us proceed to prove the second fact, i.e.,

$$
\operatorname{Im}\left(\tilde{E}_{x}+\tilde{B}_{x} \tilde{F}_{w}\right) \subseteq \mathcal{V}^{\ominus}\left(\tilde{\Sigma}_{x}\right)+\mathcal{S}^{\odot}\left(\tilde{\Sigma}_{x}\right)
$$

We will have to apply several nonsingular state transformations to the system

$$
\left\{\begin{array}{l}
\dot{\tilde{x}}_{x}=\tilde{A}_{x} \tilde{x}_{x}+\tilde{B}_{x} \tilde{v}_{x}+\left(\tilde{E}_{x}+\tilde{B}_{x} \tilde{F}_{w}\right) \tilde{w}_{x}  \tag{8.1.39}\\
\tilde{h}_{x}=\tilde{C}_{x} \tilde{x}_{x}+\tilde{D}_{x} \tilde{v}_{x}
\end{array}\right.
$$

and transform it into the form of the special coordinate basis as given in Theorem 2.3.1. First let us define a state transformation

$$
\begin{equation*}
\tilde{T}_{x}=\left(A_{x}+I\right)^{-2} \tag{8.1.40}
\end{equation*}
$$

In view of (8.1.28), it is straightforward, although tedious, to verify that

$$
\tilde{T}_{x}=\left[\begin{array}{ccc}
\left(A_{a a}^{+}+I\right)^{-2} & \star & \star  \tag{8.1.41}\\
0 & \left(A_{d d}+I\right)^{-2} & 0 \\
0 & X_{5} & \left(A_{b b}+I\right)^{-2}
\end{array}\right]
$$

where $\star$ 's are matrices of not much interest and

$$
\begin{equation*}
X_{5}=-\left(A_{b b}+I\right)^{-1}\left[L_{b d} C_{d}\left(A_{d d}+I\right)^{-1}+\left(A_{b b}+I\right)^{-1} L_{b d} C_{d}\right]\left(A_{d d}+I\right)^{-1} \tag{8.1.42}
\end{equation*}
$$

and

$$
\begin{align*}
\bar{A}_{x} & :=\tilde{T}_{x}^{-1} \tilde{A}_{x} \tilde{T}_{x}=\left(A_{x}-I\right)\left(A_{x}+I\right)^{-1}  \tag{8.1.43}\\
& =\left[\begin{array}{ccc}
\left(A_{a a}^{+}-I\right)\left(A_{a a}^{+}+I\right)^{-1} & \star & 2\left(A_{a a}^{+}+I\right)^{-1} L_{a b}^{+} C_{b}\left(A_{b b}+I\right)^{-1} \\
0 & \left(A_{d d}-I\right)\left(A_{d d}+I\right)^{-1} & 0 \\
0 & 2\left(A_{b b}+I\right)^{-1} L_{b d} C_{d}\left(A_{d d}+I\right)^{-1} & \left(A_{b b}-I\right)\left(A_{b b}+I\right)^{-1}
\end{array}\right]
\end{align*}
$$

$$
\bar{B}_{x}:=\tilde{T}_{x}^{-1} \tilde{B}_{x}=2 B_{x}=2\left[\begin{array}{cc}
B_{0 a}^{+} & 0  \tag{8.1.44}\\
B_{0 d} & B_{d} \\
B_{0 b} & 0
\end{array}\right]
$$

$\bar{E}_{x}:=\tilde{T}_{x}^{-1}\left(\tilde{E}_{x}+\tilde{B}_{x} \tilde{F}_{w}\right)=2\left[\begin{array}{c}E_{a}^{+} \\ E_{d}^{*} \\ E_{b}\end{array}\right], \quad$ where $E_{b}=0$,
$\bar{C}_{x}:=\tilde{C}_{x} \tilde{T}_{x}$

$$
=\tilde{\Gamma}_{o}\left[\begin{array}{ccc}
0 & 0 & 0  \tag{8.1.46}\\
0 & -\left[C_{d}\left(A_{d d}+I\right)^{-1} B_{d}\right]^{-1} C_{d}\left(A_{d d}+I\right)^{-2} & 0 \\
0 & -C_{b}\left(A_{b b}+I\right)^{-2} L_{b d} C_{d}\left(A_{d d}+I\right)^{-1} & C_{b}\left(A_{b b}+I\right)^{-2}
\end{array}\right]
$$

$\bar{D}_{x}:=\tilde{D}_{x}=\tilde{\Gamma}_{o}\left[\begin{array}{ll}I & 0 \\ 0 & I \\ 0 & 0\end{array}\right]$.

In order to bring the system of (8.1.39) into the standard form of the special coordinate basis, we will have to perform another state transformation that will cause the $(3,2)$ block of $\bar{C}_{x}$ in the right hand side of (8.1.46) to vanish. The following transformation $\bar{T}_{x}$ will do the job,

$$
\bar{T}_{x}=\left[\begin{array}{ccc}
I & 0 & 0  \tag{8.1.48}\\
0 & I & 0 \\
0 & L_{b d} C_{d}\left(A_{d d}+I\right)^{-1} & \left(A_{b b}+I\right)^{2}
\end{array}\right]
$$

It is quite easy to verify this time that

$$
\begin{align*}
& \hat{A}_{x}:=\bar{T}_{x}^{-1} \bar{A}_{x} \bar{T}_{x} \\
& =\left[\begin{array}{ccc}
\left(A_{a a}^{+}-I\right)\left(A_{a a}^{+}+I\right)^{-1} & \star & 2\left(A_{a a}^{+}+I\right)^{-1} L_{a b}^{+} C_{b}\left(A_{b b}+I\right) \\
0 & & \left(A_{d d}-I\right)\left(A_{d d}+I\right)^{-1}
\end{array}\right], \\
& \hat{B}_{x}:=\hat{B}_{x 0}:=\bar{T}_{x}^{-1} \bar{B}_{x}=2\left[\begin{array}{cc}
B_{0 a}^{+} & 0 \\
B_{0 d} & B_{d} \\
\star & -\left(A_{b b}+I\right)^{-2} L_{b d} C_{d}\left(A_{d d}+I\right)^{-1} B_{d}
\end{array}\right],  \tag{8.1.50}\\
& \hat{E}_{x}:=\bar{T}_{x}^{-1} \bar{E}_{x} \\
& =2\left[\begin{array}{c}
E_{a}^{+} \\
E_{d}^{*} \\
\left(A_{b b}+I\right)^{-2}\left[E_{b}-L_{b d} C_{d}\left(A_{d d}^{+}+I\right)^{-1} E_{d}^{*}\right]
\end{array}\right]=2\left[\begin{array}{c}
E_{a}^{+} \\
E_{d}^{*} \\
0
\end{array}\right], \\
& \hat{C}_{x}:=\left[\begin{array}{c}
\hat{C}_{x 0} \\
\hat{C}_{x 1}
\end{array}\right]:=\bar{C}_{x} \bar{T}_{x} \\
& =\tilde{\Gamma}_{o}\left[\begin{array}{ccc}
0 & 0 & 0 \\
0 & -\left[C_{d}\left(A_{d d}+I\right)^{-1} B_{d}\right]^{-1} C_{d}\left(A_{d d}+I\right)^{-2} & 0 \\
\hline 0 & 0 & C_{b}
\end{array}\right],  \tag{8.1.52}\\
& \hat{D}_{x}:=\bar{D}_{x}=\tilde{D}_{x} . \tag{8.1.53}
\end{align*}
$$

Then we have
$\hat{A}_{x}-\hat{B}_{x 0} \hat{C}_{x 0}=\left[\begin{array}{ccc}\left(A_{a a}^{+}-I\right)\left(A_{a a}^{+}+I\right)^{-1} & \star & 2\left(A_{a a}^{+}+I\right)^{-1} L_{a b}^{+} C_{b}\left(A_{b b}+I\right) \\ 0 & A_{a a}^{*} & 0 \\ 0 & 0 & \left(A_{b b}+I\right)^{-1}\left(A_{b b}-I\right)\end{array}\right]$,
where

$$
\begin{equation*}
A_{a a}^{*}=\left(A_{d d}-I\right)\left(A_{d d}+I\right)^{-1}+2 B_{d}\left[C_{d}\left(A_{d d}+I\right)^{-1} B_{d}\right]^{-1} C_{d}\left(A_{d d}+I\right)^{-2} \tag{8.1.55}
\end{equation*}
$$

Define another nonsingular state transformation,

$$
\hat{T}_{x}=\left[\begin{array}{ccc}
I & 0 & \hat{T}_{*}  \tag{8.1.56}\\
0 & I & 0 \\
0 & 0 & I
\end{array}\right]
$$

with $\hat{T}_{*}$ being a solution to the following general Lyapunov equation

$$
\left(I-A_{a a}^{+}\right)\left(I+A_{a a}^{+}\right)^{-1} \hat{T}_{*}+\hat{T}_{*}\left(A_{b b}+I\right)^{-1}\left(A_{b b}-I\right)=2\left(A_{a a}^{+}-I\right)^{-1} L_{a b}^{+} C_{b}\left(A_{b b}+I\right)
$$

It follows from Kailath [48] that such a solution always exists and is unique if $A_{a a}^{+}$and $A_{b b}$ have no common eigenvalue. Then it is straightforward to verify that it would transform the $(1,3)$ block of $\hat{A}_{x}-\hat{B}_{x 0} \hat{C}_{x 0}$ in (8.1.54) to 0 while not changing the structures of other blocks. Hence, $\hat{T}_{x}$ would also transform the system $\left(\hat{A}_{x}, \hat{B}_{x}, \hat{C}_{x}, \hat{D}_{x}\right)$ and $\hat{E}_{x}$ into the standard form of the special coordinate basis as given in Theorem 2.3 .1 since the pair $\left\{\left(A_{b b}+I\right)^{-1}\left(A_{b b}-I\right), C_{b}\right\}$ is completely observable due to the complete observability of $\left(A_{b b}, C_{b}\right)$. It is now clear from the properties of the special coordinate basis that

$$
\operatorname{Im}\left(\hat{E}_{x}\right) \subseteq \mathcal{V}^{\odot}\left(\hat{\Sigma}_{x}\right)+\mathcal{S}^{\odot}\left(\hat{\Sigma}_{x}\right)
$$

where $\hat{\Sigma}_{x}$ is characterized by $\left(\hat{A}_{x}, \hat{B}_{x}, \hat{C}_{x}, \hat{D}_{x}\right)$, which is equivalent to

$$
\operatorname{Im}\left(\tilde{E}_{x}+\tilde{B}_{x} \tilde{F}_{w}\right) \subseteq \mathcal{V}^{\odot}\left(\tilde{\Sigma}_{x}\right)+\mathcal{S}^{\ominus}\left(\tilde{\Sigma}_{x}\right)
$$

This proves the second fact.
Next, let us apply a pre-disturbance feedback law,

$$
\begin{equation*}
\tilde{u}_{x}=\tilde{F}_{w} \tilde{w}_{x}+\tilde{v}_{x}=-\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{D}_{x}^{\prime} \tilde{D}_{22} \tilde{w}_{x}+\tilde{v}_{x} \tag{8.1.57}
\end{equation*}
$$

to the auxiliary system (8.1.15). Again, this pre-feedback law will not affect solutions to the $H_{\infty}$ problem for (8.1.15) or to the solution $\tilde{P}_{x}$ of (8.1.20)(8.1.22). After applying this pre-feedback law, we obtain the following new system

$$
\left\{\begin{array}{lll}
\dot{\tilde{x}}_{x}= & \tilde{A}_{x} & \tilde{x}_{x}+\tilde{B}_{x} \tilde{v}_{x}+\left[\tilde{E}_{x}-\tilde{B}_{x}\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{D}_{x}^{\prime} \tilde{D}_{22}\right]  \tag{8.1.58}\\
\tilde{y}_{x} \\
\tilde{y}_{x}=\binom{0}{I} \tilde{x}_{x} \\
\tilde{h}_{x}= & \binom{I}{0} & \tilde{C}_{x} \\
\tilde{x}_{x}+\tilde{D}_{x} \\
\tilde{v}_{x}+ & 0 & \tilde{w}_{x}
\end{array}\right.
$$

Then it follows from Corollary 3.1.1 that the existence condition of a $\gamma$ suboptimal controller for (8.1.58) is equivalent to the existence of a matrix $\tilde{P}_{x}>0$ such that

$$
\begin{aligned}
& 0=\tilde{P}_{x} \tilde{A}_{x}+\tilde{A}_{x}^{\prime} \tilde{P}_{x}+\tilde{C}_{x}^{\prime} \tilde{C}_{x}-\left(\tilde{P}_{x} \tilde{B}_{x}+\tilde{C}_{x}^{\prime} \tilde{D}_{x}\right)\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1}\left(\tilde{P}_{x} \tilde{B}_{x}+\tilde{C}_{x}^{\prime} \tilde{D}_{x}\right)^{\prime} \\
& \quad+\tilde{P}_{x}\left[\tilde{E}_{x}-\tilde{B}_{x}\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{D}_{x}^{\prime} \tilde{D}_{22}\right]\left[\tilde{E}_{x}-\tilde{B}_{x}\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{D}_{x}^{\prime} \tilde{D}_{22}\right]^{\prime} \tilde{P}_{x} / \gamma^{2}
\end{aligned}
$$

is satisfied. Note that the solution $\tilde{P}_{x}$ to the above Riccati equation is identical to the solution that satisfies (8.1.20)-(8.1.21).

Now, it follows from Theorem 4.1.2 that ( $\tilde{A}_{x}, \tilde{B}_{x}, \tilde{C}_{x}, \tilde{D}_{x}$ ) is left invertible, and is free of infinite zeros and stable invariant zeros as well as invariant zeros on the unit circle. Also, in view of the second fact of the auxiliary system of (8.1.58), it satisfies Assumptions 5.F. 1 to 5.F. 4 of Chapter 5. Following the results of Chapter 5, we can easily show that

$$
\begin{equation*}
\gamma^{*}=\sqrt{\lambda_{\max }\left(\tilde{T}_{x} \tilde{S}_{x}^{-1}\right)} \tag{8.1.59}
\end{equation*}
$$

and for any $\gamma>\gamma^{*}$, the positive definite solution $\tilde{P}_{x}$ of (8.1.20)-(8.1.22) is given by

$$
\begin{equation*}
\tilde{P}_{x}=\left(\tilde{S}_{x}-\tilde{T}_{x} / \gamma^{2}\right)^{-1} \tag{8.1.60}
\end{equation*}
$$

It then follows from (8.1.23) that for any $\gamma>\gamma^{*}$, the positive definite solution $P_{x}$ of (8.1.16)-(8.1.19) is given by

$$
\begin{equation*}
P_{x}=2\left(A_{x}^{\prime}+I\right)^{-1}\left(\tilde{S}_{x}-\tilde{T}_{x} / \gamma^{2}\right)^{-1}\left(A_{x}+I\right)^{-1} \tag{8.1.61}
\end{equation*}
$$

and hence $\gamma^{*}$ can also be obtained from the following expression,

$$
\begin{equation*}
\gamma^{*}=\sqrt{\lambda_{\max }\left(T_{x} S_{x}^{-1}\right)} \tag{8.1.62}
\end{equation*}
$$

where $S_{x}$ and $T_{x}$ are as defined in (8.1.10) and (8.1.11), respectively. Moreover, it is straightforward to verify that

$$
P(\gamma)=\left(\Gamma_{s}^{-1}\right)^{\prime}\left[\begin{array}{cc}
0 & 0 \\
0 & \left(S_{x}-T_{x} / \gamma^{2}\right)^{-1}
\end{array}\right] \Gamma_{s}^{-1}
$$

is the unique solution that satisfies conditions 2.(a)-2.(c) of Theorem 3.2.1.
Finally, note that $\left(\tilde{A}_{x}, \tilde{B}_{x}, \tilde{C}_{x}, \tilde{D}_{x}\right)$ is left invertible, and is free of infinite zeros and stable invariant zeros as well as invariant zeros on the unit circle. It follows from Richardson and Kwong [83] that the solution $\tilde{S}_{x}$ to the Riccati equation (8.1.8) is positive definite because ( $\tilde{A}_{x}, \tilde{B}_{x}$ ) is controllable, and the solution $\tilde{T}_{x}$ to the Lyapunov equation (8.1.9) is positive semi-definite. In fact, both of them are unique. This completes the proof of our algorithm.

The following remarks are in order.

Remark 8.1.1. For the case when $D_{22} \neq 0$, Assumption 8.F. 3 should be replaced by the following conditions:

1. $\tilde{D}_{22}:=D_{22}-C_{x}\left(A_{x}+I\right)^{-1} E_{x}$ is in the range space of $\tilde{D}_{x}$, and
2. $\operatorname{Im}\left[\tilde{E}_{x}-\tilde{B}_{x}\left(\tilde{D}_{x}^{\prime} \tilde{D}_{x}\right)^{-1} \tilde{D}_{x}^{\prime} \tilde{D}_{22}\right] \subseteq \mathcal{V}^{\odot}\left(\tilde{\Sigma}_{x}\right)+\mathcal{S}^{\ominus}\left(\tilde{\Sigma}_{x}\right)$.

Then our algorithm would carry through without any problems. We would also like to note that if $\left(A, B, C_{2}, D_{2}\right)$ is right invertible, then $\left(\tilde{A}_{x}, \tilde{B}_{x}, \tilde{C}_{x}, \tilde{D}_{x}\right)$ is invertible and $\tilde{D}_{x}$ is square and nonsingular, and $\mathcal{V}^{\ominus}\left(\tilde{\Sigma}_{x}\right)+\mathcal{S}^{\odot}\left(\tilde{\Sigma}_{x}\right)=\mathbb{R}^{n_{x}}$. Hence, the above two conditions will be automatically satisfied. Such a result was first reported in Chen [13].

Remark 8.1.2. If Assumptions 8.F. 3 and 8.F. 4 are not satisfied, then one might have to approximate iteratively the infimum $\gamma^{*}$ by finding the smallest non-negative scalar, say $\tilde{\gamma}^{*} \geq 0$, such that the Riccati equation (8.1.20) and (8.1.21) are satisfied.

We illustrate the above results in the following example.
Example 8.1.1. Consider a full information system (8.1.1) characterized by

$$
A=\left[\begin{array}{lllll}
1 & 1 & 1 & 1 & 1  \tag{8.1.63}\\
0 & 0 & 0 & 1 & 1 \\
0 & 0 & 1 & 1 & 1 \\
1 & 1 & 1 & 1 & 1 \\
0 & 0 & 0 & 1 & 1
\end{array}\right], \quad B=\left[\begin{array}{lll}
0 & 0 & 1 \\
0 & 0 & 0 \\
1 & 0 & 0 \\
0 & 1 & 0 \\
0 & 0 & 0
\end{array}\right], \quad E=\left[\begin{array}{l}
1 \\
1 \\
1 \\
1 \\
0
\end{array}\right]
$$

and

$$
C_{2}=\left[\begin{array}{rrrrr}
0 & 0 & -1 & 0 & 0  \tag{8.1.64}\\
0 & 0 & 0 & 1 & 0 \\
0 & 0 & 0 & 0 & 1
\end{array}\right], \quad D_{2}=\left[\begin{array}{lll}
1 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right], \quad D_{22}=\left[\begin{array}{l}
0 \\
0 \\
0
\end{array}\right]
$$

It is can be verified that $(A, B)$ is controllable and $\left(A, B, C_{2}, D_{2}\right)$ is neither right nor left invertible, and is of nonminimum phase with two invariant zeros at 0 and 2, respectively. Moreover, it is already in the form of the special coordinate basis as given in Theorem 2.3.1 and Assumption 8.F. 3 is satisfied as $E_{b}=0$. Hence, Assumptions 8.F. 1 to 8.F. 4 are all satisfied. Following the algorithm, we obtain

$$
\begin{gathered}
\Gamma_{s}=I_{5}, \quad n_{x}=3 \\
A_{x}=\left[\begin{array}{lll}
2 & 1 & 1 \\
1 & 1 & 1 \\
0 & 1 & 1
\end{array}\right], \quad B_{x}=\left[\begin{array}{ll}
1 & 0 \\
0 & 1 \\
0 & 0
\end{array}\right], \quad E_{x}=\left[\begin{array}{l}
1 \\
1 \\
0
\end{array}\right]
\end{gathered}
$$

$$
\begin{gathered}
C_{x}=\left[\begin{array}{lll}
0 & 0 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{array}\right], \quad D_{x}=\left[\begin{array}{ll}
1 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right] \\
\tilde{A}_{x}=\left[\begin{array}{rrr}
0.25 & 0.25 & 0.25 \\
0.50 & -0.50 & 0.50 \\
-0.25 & 0.75 & -0.25
\end{array}\right] \\
\tilde{B}_{x}=\left[\begin{array}{rr}
0.3125 & -0.1875 \\
-0.6250 & 1.3750 \\
0.4375 & -1.0625
\end{array}\right], \quad \tilde{E}_{x}=\left[\begin{array}{r}
0.125 \\
0.750 \\
-0.625
\end{array}\right]
\end{gathered}
$$

and

$$
\tilde{C}_{x}=C_{x}, \quad \tilde{D}_{x}=\left[\begin{array}{rr}
1.000 & 0.000 \\
0.250 & -0.750 \\
-0.125 & 0.375
\end{array}\right], \quad \tilde{D}_{22}=\left[\begin{array}{r}
0.00 \\
-0.50 \\
0.25
\end{array}\right] .
$$

It is simple to verify that $\left(\tilde{A}_{x}, \tilde{B}_{x}, \tilde{C}_{x}, \tilde{D}_{x}\right)$ is left invertible with two invariant zeros at 1 and $1 / 3$, respectively. Solving Riccati equations (8.1.8) and (8.1.9), we obtain

$$
\tilde{S}_{x}=\left[\begin{array}{rrr}
0.227615 & -0.207890 & 0.019725 \\
-0.207890 & 1.202254 & -1.005636 \\
0.019725 & -1.005636 & 1.014089
\end{array}\right]
$$

and

$$
\tilde{T}_{x}=\left[\begin{array}{rrr}
0.09375 & -0.062500 & 0.031250 \\
-0.06250 & 0.041667 & -0.020833 \\
0.03125 & -0.020833 & 0.010417
\end{array}\right]
$$

Finally, we get

$$
S_{x}=\left[\begin{array}{rrr}
0.562306 & -0.145898 & -0.145898 \\
-0.145898 & 0.618034 & -0.381966 \\
-0.145898 & -0.381966 & 0.618034
\end{array}\right], \quad T_{x}=\left[\begin{array}{ccc}
1 / 3 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right],
$$

and the infimum

$$
\gamma^{*}=0.934173
$$

### 8.2. Output Feedback Case

We present in this section a well-conditioned non-iterative algorithm for the exact computation of $\gamma^{*}$ of the following measurement feedback discrete-time system $\Sigma$,

$$
\left\{\begin{align*}
x(k+1) & =A x(k)+B u(k)  \tag{8.2.1}\\
y(k) & =C_{1} x(k) \\
& +D_{1} w(k), \\
h(k) & =C_{2} x(k)+D_{2} u(k)+D_{22} w(k),
\end{align*}\right.
$$

where $x \in \mathbb{R}^{n}$ is the state, $u \in \mathbb{R}^{m}$ is the control input, $w \in \mathbf{R}^{q}$ is the external disturbance input, $y \in \mathbb{R}^{p}$ is the measurement output, and $h \in \mathbf{R}^{\ell}$ is the controlled output of $\Sigma$. Again, for easy reference, we define $\Sigma_{\mathrm{P}}$ to be the subsystem characterized by the matrix quadruple $\left(A, B, C_{2}, D_{2}\right)$ and $\Sigma_{Q}$ to be the subsystem characterized by the matrix quadruple $\left(A, E, C_{1}, D_{1}\right)$. We first make the following assumptions:

Assumption 8.M.1: $(A, B)$ is stabilizable;
Assumption 8.M.2: $\Sigma_{\mathrm{P}}$ has no invariant zero on the unit circle;
Assumption 8.M.3: $\operatorname{Im}(E) \subset \mathcal{V}^{\odot}\left(\Sigma_{\mathrm{P}}\right)+\mathcal{S}^{\odot}\left(\Sigma_{\mathrm{P}}\right)$;
Assumption 8.M.4: $\left(A, C_{1}\right)$ is detectable;
Assumption 8.M.5: $\Sigma_{Q}$ has no invariant zero on the unit circle;
Assumption 8.M.6: $\operatorname{Ker}\left(C_{2}\right) \supset \mathcal{V}^{\odot}\left(\Sigma_{\mathrm{Q}}\right) \cap \mathcal{S}^{\odot}\left(\Sigma_{\mathrm{Q}}\right)$; and
Assumption 8.M.7: $D_{22}=0$.
As in the previous section, we outline a step-by-step algorithm for the computation of $\gamma^{*}$ below:

Step 8.M.1: Define an auxiliary full information problem for

$$
\left\{\begin{align*}
x(k+1) & =A x(k)+B u(k)+E \quad w(k)  \tag{8.2.2}\\
y(k) & =\binom{0}{I} x(k) \\
& +\binom{I}{0} w(k) \\
h(k) & =C_{2} x(k)+D_{2} u(k)+D_{22} w(k)
\end{align*}\right.
$$

and perform Steps 8.F. 1 to 8.F. 3 of the algorithm given in the previous section. For future use and in order to avoid notational confusion, we rename the state transformation of the special coordinate basis for $\Sigma_{\mathrm{P}}$ as $\Gamma_{s \mathrm{P}}$ and the dimension of $A_{x}$ as $n_{x \mathrm{P}}$. Also, rename $S_{x}$ of (8.1.10) and $T_{x}$ of (8.1.11) as $S_{x \mathrm{P}}$ and $T_{x \mathrm{P}}$, respectively.

Step 8.M.2: Define another auxiliary full information problem for

$$
\left\{\begin{align*}
& x(k+1)=A^{\prime} x(k)+C_{1}^{\prime} u(k)+C_{2}^{\prime} w(k)  \tag{8.2.3}\\
& y(k)=\binom{0}{I} x(k) \\
& h(k)=E^{\prime} x(k)+D_{1}^{\prime} u(k)+D_{22}^{\prime} w(k) \\
& 0
\end{align*}\right),
$$

and again perform Steps 8.F. 1 to 8.F. 3 of the algorithm given in Section 8.1 one more time, but for this auxiliary system. Let $\Sigma_{Q}^{\star}$ be the dual
system of $\Sigma_{Q}$ and be characterized by ( $A^{\prime}, C_{1}^{\prime}, E^{\prime}, D_{1}^{\prime}$ ). We rename the state transformation of the special coordinate basis for $\Sigma_{Q}^{\star}$ as $\Gamma_{s Q}$ and the dimension of $A_{x}$ as $n_{x \mathrm{Q}}$, and $S_{x}$ of (8.1.10) and $T_{x}$ of (8.1.11) as $S_{x \mathrm{Q}}$ and $T_{x Q}$, respectively.

Step 8.M.3: Partition

$$
\Gamma_{s \mathrm{P}}^{-1}\left(\Gamma_{s Q}^{-1}\right)^{\prime}=\left[\begin{array}{ll}
\star & \star  \tag{8.2.4}\\
\star & \Gamma
\end{array}\right]
$$

where $\Gamma$ is a $n_{x \mathrm{P}} \times n_{x \mathrm{Q}}$ matrix, and define a constant matrix

$$
M=\left[\begin{array}{cc}
T_{x \mathrm{P}} S_{x \mathrm{P}}^{-1}+\Gamma S_{x Q}^{-1} \Gamma^{\prime} S_{x \mathrm{P}}^{-1} & -\Gamma S_{x Q}^{-1}  \tag{8.2.5}\\
-T_{x Q} S_{x Q}^{-1} \Gamma^{\prime} S_{x \mathrm{P}}^{-1} & T_{x \mathrm{Q}} S_{x Q}^{-1}
\end{array}\right] .
$$

Step 8.M.4: The infimum $\gamma^{*}$ is then given by

$$
\begin{equation*}
\gamma^{*}=\sqrt{\lambda_{\max }(M)} \tag{8.2.6}
\end{equation*}
$$

where $M$ has only real and non-negative eigenvalues.
Proof of the Algorithm. Once the result for the full information case is established, the proof of this algorithm is similar to the one given in Section 5.2 of Chapter 5.

The following remarks are in order.
Remark 8.2.1. Consider the given discrete-time system (8.2.1) that satisfies Assumptions 8.M.1 to 8.M.7. Then for any $\gamma>\gamma^{*}$, where $\gamma^{*}$ is given by (8.2.6), the following $P(\gamma)$ and $Q(\gamma)$,

$$
P(\gamma):=\left(\Gamma_{s \mathrm{P}}^{-1}\right)^{\prime}\left[\begin{array}{cc}
0 & 0  \tag{8.2.7}\\
0 & \left(S_{x \mathrm{P}}-T_{x \mathrm{P}} / \gamma^{2}\right)^{-1}
\end{array}\right] \Gamma_{s \mathrm{P}}^{-1},
$$

and

$$
Q(\gamma):=\left(\Gamma_{s Q}^{-1}\right)^{\prime}\left[\begin{array}{cc}
0 & 0  \tag{8.2.8}\\
0 & \left(S_{x Q}-T_{x Q} / \gamma^{2}\right)^{-1}
\end{array}\right] \Gamma_{s Q}^{-1}
$$

satisfy conditions 2.(a)-2.(g) of Theorem 3.2.1.
Remark 8.2.2. For discrete-time $H_{\infty}$ control, $\gamma^{*}$ for the full information feedback system is in general different from that of the full state feedback system regardless of $D_{22}=0$ or not. For the state feedback case, i.e., $C_{1}=I$ and $D_{1}=0$, we note that the subsystem $\Sigma_{Q}$ is always free of invariant zeros (and hence free of unit circle invariant zeros) and left invertible. Thus, as long as $\Sigma_{\mathrm{P}}$ is free of unit circle invariant zeros and satisfies Assumption 8.M.1 to 8.M.3, one
can apply the above algorithm to get the infimum, $\gamma^{*}$. For this special case $\Gamma_{s Q}$, $n_{x Q}, S_{x Q}$ and $T_{x Q}$ in Step 8.M. 2 of the above algorithm can be directly obtained using the following simple procedure: Compute a nonsingular transformation $\Gamma_{s Q}$ such that

$$
\Gamma_{s Q}^{\prime} E=\left[\begin{array}{c}
0  \tag{8.2.9}\\
\hat{E}
\end{array}\right]
$$

where $\hat{E}$ is a $n_{x Q} \times n_{x Q}$ nonsingular matrix. Then $S_{x Q}$ and $T_{x Q}$ are respectively given by

$$
\begin{equation*}
S_{x Q}=\left(\hat{E}^{-1}\right)^{\prime} \hat{E}^{-1} \quad \text { and } \quad T_{x Q}=0 \tag{8.2.10}
\end{equation*}
$$

and hence

$$
\begin{equation*}
\gamma^{*}=\left[\lambda_{\max }\left(T_{x \mathrm{P}} S_{x \mathrm{P}}^{-1}+\Gamma S_{x \mathrm{Q}}^{-1} \Gamma^{\prime} S_{x \mathrm{P}}^{-1}\right)\right]^{\frac{1}{2}} . \tag{8.2.11}
\end{equation*}
$$

Note that in general, $\gamma^{*} \geq\left\{\lambda_{\max }\left(T_{x \mathrm{P}} S_{x \mathrm{P}}^{-1}\right)\right\}^{\frac{1}{2}}$.

Remark 8.2.3. For the case when $D_{22} \neq 0$, Assumptions 8.M. 3 and 8.M. 6 should be replaced by the conditions given in Remark 8.1.1, which is associated with the full information system of (8.2.2), and a set of conditions similar to those in that remark, but for the full information system of (8.2.3). Then our procedure would again carry through and yield the correct result. Note that if $\Sigma_{\mathrm{P}}$ is right invertible and $\Sigma_{Q}$ is left invertible, then all these conditions will be automatically satisfied. The result will then reduce to that of Chen [13]. 园

Remark 8.2.4. If Assumptions 8.M. 3 and 8.M.6, i.e., the geometric conditions, and Assumption 8.M. 7 are not satisfied, then an iterative scheme might be used to determine the infimum. This can be done by finding the smallest scalar, say $\tilde{\gamma}^{*}$, such that all the following conditions are satisfied:

1. The Riccati equation

$$
\begin{aligned}
0=\tilde{P}_{x} \tilde{A}_{x \mathrm{P}} & +\tilde{A}_{x \mathrm{P}}^{\prime} \tilde{P}_{x}+\tilde{C}_{x \mathrm{P}}^{\prime} \tilde{C}_{x \mathrm{P}}-\left[\begin{array}{c}
\tilde{B}_{x \mathrm{P}}^{\prime} \tilde{P}_{x}+\tilde{D}_{x \mathrm{P}}^{\prime} \tilde{C}_{x \mathrm{P}} \\
\tilde{E}_{x \mathrm{P}}^{\prime} \tilde{P}_{x}+\tilde{D}_{22 \mathrm{P}}^{\prime} \tilde{C}_{x \mathrm{P}}
\end{array}\right]^{\prime} \\
& \times\left[\begin{array}{cc}
\tilde{D}_{x \mathrm{P}}^{\prime} \tilde{D}_{x \mathrm{P}} & \tilde{D}_{x \mathrm{P}}^{\prime} \tilde{D}_{22 \mathrm{P}} \\
\tilde{D}_{22 \mathrm{P}}^{\prime} \tilde{D}_{x \mathrm{P}} & \tilde{D}_{22 \mathrm{P}}^{\prime} \tilde{D}_{22 \mathrm{P}}-\left(\tilde{\gamma}^{*}\right)^{2} I
\end{array}\right]^{-1}\left[\begin{array}{c}
\tilde{B}_{x \mathrm{P}}^{\prime} \tilde{P}_{x}+\tilde{D}_{x \mathrm{P}}^{\prime} \tilde{C}_{x \mathrm{P}} \\
\tilde{E}_{x \mathrm{P}}^{\prime} \tilde{P}_{x}+\tilde{D}_{22 \mathrm{P}}^{\prime} \tilde{C}_{x \mathrm{P}}
\end{array}\right]
\end{aligned}
$$

has a positive definite solution $\tilde{P}_{x}>0$, which satisfies

$$
\tilde{D}_{22 \mathrm{P}}^{\prime}\left[I-\tilde{D}_{x \mathrm{P}}\left(\tilde{D}_{x \mathrm{P}}^{\prime} \tilde{D}_{x \mathrm{P}}\right)^{-1} \tilde{D}_{x \mathrm{P}}^{\prime}\right] \tilde{D}_{22 \mathrm{P}}<\left(\tilde{\gamma}^{*}\right)^{2} I
$$

Here we note that all the sub-matrices in the above Riccati equation are defined as in (8.1.7) but for the auxiliary system (8.2.2) of Step 8.M.1.
2. The Riccati equation

$$
\begin{aligned}
0=\tilde{Q}_{x} \tilde{A}_{x Q} & +\tilde{A}_{x Q}^{\prime} \tilde{Q}_{x}+\tilde{C}_{x Q}^{\prime} \tilde{C}_{x Q}-\left[\begin{array}{c}
\tilde{B}_{x Q}^{\prime} \tilde{Q}_{x}+\tilde{D}_{x Q}^{\prime} \tilde{C}_{x Q} \\
\tilde{E}_{x Q}^{\prime} \tilde{Q}_{x}+\tilde{D}_{22 Q}^{\prime} \tilde{C}_{x Q}
\end{array}\right]^{\prime} \\
& \times\left[\begin{array}{cc}
\tilde{D}_{x Q}^{\prime} \tilde{D}_{x Q} & \tilde{D}_{x Q}^{\prime} \tilde{D}_{22 Q} \\
\tilde{D}_{22 Q}^{\prime} \tilde{D}_{x Q} & \tilde{D}_{22 Q}^{\prime} \tilde{D}_{22 Q}-\left(\tilde{\gamma}^{*}\right)^{2} I
\end{array}\right]^{-1}\left[\begin{array}{c}
\tilde{B}_{x Q}^{\prime} \tilde{Q}_{x}+\tilde{D}_{x Q}^{\prime} \tilde{C}_{x Q} \\
\tilde{E}_{x Q}^{\prime} \tilde{Q}_{x}+\tilde{D}_{22 Q}^{\prime} \tilde{C}_{x Q}
\end{array}\right],
\end{aligned}
$$

has a positive definite solution $\tilde{Q}_{x}>0$, which satisfies

$$
\tilde{D}_{22 Q}^{\prime}\left[I-\tilde{D}_{x Q}\left(\tilde{D}_{x Q}^{\prime} \tilde{D}_{x Q}\right)^{-1} \tilde{D}_{x Q}^{\prime}\right] \tilde{D}_{22 Q}<\left(\tilde{\gamma}^{*}\right)^{2} I
$$

Similarly, we note that all the sub-matrices in the above Riccati equation are defined as in (8.1.7) but for the auxiliary system (8.2.3) of Step 8.M.2.
3. Finally, the coupling condition holds, i.e.,

$$
\begin{equation*}
\lambda_{\max }\left\{\tilde{P}_{x} \Gamma \tilde{Q}_{x} \Gamma^{\prime}\right\}<\left(\tilde{\gamma}^{*}\right)^{2} \tag{8.2.12}
\end{equation*}
$$

where $\Gamma$ is as defined in (8.2.4).
The following example illustrates our computational algorithms.
Example 8.2.1. We consider a discrete-time measurement feedback system (8.2.1) with $A, B, E, C_{2}, D_{2}$ and $D_{22}$ being given as those in Example 8.1.1 of the previous section. We consider the full state feedback case first, i.e., $C_{1}=I$ and $D_{1}=0$. Following the algorithm and the simplified procedure in Remark 8.2.2, we obtain those matrices as in the full information case and

$$
\begin{gathered}
\Gamma_{s Q}=\left[\begin{array}{rrrrr}
1 & 1 & 1 & 0 & 1 \\
-1 & 0 & 0 & 0 & 0 \\
0 & -1 & 0 & 0 & 0 \\
0 & 0 & -1 & 0 & 0 \\
0 & 0 & 0 & 1 & 0
\end{array}\right], \quad n_{x Q}=1 \\
S_{x Q}=1, \quad T_{x Q}=0, \quad \Gamma=\left[\begin{array}{l}
1 \\
1 \\
0
\end{array}\right]
\end{gathered}
$$

and

$$
\gamma^{*}=3.181043
$$

Now, we consider the computation of $\gamma^{*}$ for the given system with an output measurement characterized by

$$
C_{1}=\left[\begin{array}{lllll}
0 & 0 & 0 & 0 & 1 \tag{8.2.13}
\end{array}\right], \quad D_{1}=0
$$

It can be shown that $\left(A, C_{1}\right)$ is detectable and $\left(A, E, C_{1}, D_{1}\right)$ is invertible with three invariant zeros at $0,0.618$ and -1.618 , respectively, and one infinite zero of order 2. Hence, Assumption 8.M. 6 is automatically satisfied. Following the algorithm, we obtain
$M=\left[\begin{array}{rrrrrr}52.08746 & 76.55250 & 66.46233 & -0.95905 & 2.61803 & -4.23607 \\ 92.57546 & 138.46401 & 120.13777 & -1.65303 & 5.23607 & -7.85410 \\ 28.03444 & 42.12461 & 36.88854 & -0.69398 & 2.61803 & -2.61803 \\ 19.20270 & 29.28949 & 24.96658 & 0 & 0 & -1.44097 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ -46.97871 & -70.77709 & -61.686918 & 0.95905 & -3.61803 & 4.23607\end{array}\right]$,
and

$$
\gamma^{*}=15.16907
$$

国

### 8.3. Plants with Unit Circle Zeros

We discuss in this section a non-iterative computational algorithm for the measurement feedback system (8.2.1) whose subsystems $\Sigma_{\mathrm{P}}$ and/or $\Sigma_{\mathrm{Q}}$ have invariant zeros on the unit circle. We assume that $(A, B)$ is stabilizable and $\left(A, C_{1}\right)$ is detectable. Let $F$ and $K$ be matrices of appropriate dimensions such that $A+B F$ and $A+K C_{1}$ have no eigenvalue at -1 and define

$$
\left.\begin{array}{rl}
\tilde{A}_{\mathrm{P}} & :=(A+B F+I)^{-1}(A+B F-I),  \tag{8.3.1}\\
\tilde{B}_{\mathrm{P}} & :=2(A+B F+I)^{-1} B, \\
\tilde{E}_{\mathrm{P}} & :=2(A+B F+I)^{-1} E, \\
\tilde{C}_{2 \mathrm{P}} & :=\left(C_{2}+D_{2} F\right)(A+B F+I)^{-1}, \\
\tilde{D}_{2 \mathrm{P}} & :=D_{2}-\left(C_{2}+D_{2} F\right)(A+B F+I)^{-1} B, \\
\tilde{D}_{22 \mathrm{P}} & :=D_{22}-\left(C_{2}+D_{2} F\right)(A+B F+I)^{-1} E,
\end{array}\right\}
$$

and

$$
\left.\begin{array}{rl}
\tilde{A}_{\mathrm{Q}} & :=\left(A+K C_{1}+I\right)^{-1}\left(A+K C_{1}-I\right), \\
\tilde{C}_{1_{\mathrm{Q}}} & :=2 C_{1}\left(A+K C_{1}+I\right)^{\prime-1} \\
\tilde{C}_{2 \mathrm{Q}} & :=2 C_{2}\left(A+K C_{1}+I\right)^{\prime-1} \\
\tilde{E}_{\mathrm{Q}} & :=\left(A+K C_{1}+I\right)^{-1}\left(E+K D_{1}\right),  \tag{8.3.2}\\
\tilde{D}_{1 \mathrm{Q}} & :=D_{1}-C_{1}\left(A+K C_{1}+I\right)^{-1}\left(E+K D_{1}\right), \\
\tilde{D}_{22 \mathrm{Q}} & :=D_{22}-C_{2}\left(A+K C_{1}+I\right)^{-1}\left(E+K D_{1}\right) .
\end{array}\right\}
$$

Let $\tilde{\Sigma}_{\mathrm{P}}$ denote the system characterized by $\left(\tilde{A}_{\mathrm{P}}, \tilde{B}_{\mathrm{P}}, \tilde{C}_{2 \mathrm{P}}, \tilde{D}_{2 \mathrm{P}}\right)$ and $\tilde{\Sigma}_{\mathrm{Q}}^{*}$ denote the system characterized by $\left(\tilde{A}_{\mathrm{Q}}^{\prime}, \tilde{C}_{1 \mathrm{Q}}^{\prime}, \tilde{E}_{\mathrm{Q}}^{\prime}, \tilde{D}_{1 \mathrm{Q}}^{\prime}\right)$. We also make the following assumptions:

Assumption 8.Z.1: $\operatorname{Im}\left(\tilde{D}_{22 \mathrm{P}}\right) \subset \operatorname{Im}\left(\tilde{D}_{2 \mathrm{P}}\right) ;$
Assumption 8.Z.2: $\operatorname{Im}\left[\tilde{E}_{\mathrm{P}}-\tilde{B}_{\mathrm{P}}\left(\tilde{D}_{2 \mathrm{P}}^{\prime} \tilde{D}_{2 \mathrm{P}}\right)^{\dagger} \tilde{D}_{2 \mathrm{P}}^{\prime} \tilde{D}_{22 \mathrm{P}}\right] \subset \mathcal{V}^{-}\left(\tilde{\Sigma}_{\mathrm{P}}\right)+\mathcal{S}^{-}\left(\tilde{\Sigma}_{\mathrm{F}}\right) ;$
Assumption 8.Z.3: $\operatorname{Im}\left(\tilde{D}_{22 Q}^{\prime}\right) \subset \operatorname{Im}\left(\tilde{D}_{1 Q}^{\prime}\right)$;
Assumption 8.Z.4: $\operatorname{Im}\left[\tilde{C}_{2 Q}^{\prime}-\tilde{C}_{1 Q}^{\prime}\left(\tilde{D}_{1 Q} \tilde{D}_{1 Q}^{\prime}\right)^{\dagger} \tilde{D}_{1 Q} \tilde{D}_{22 Q}^{\prime}\right] \subset \mathcal{V}^{-}\left(\tilde{\Sigma}_{Q}^{\star}\right)+\mathcal{S}^{-}\left(\tilde{\Sigma}_{Q}^{\star}\right)$.

It can be shown that Assumptions 8.Z.1-8.Z.4 are independent of the choice of $F$ and $K$ in (8.3.1) and (8.3.2). The computation of $\gamma^{*}$ for a plant whose subsystems have invariant zeros on the unit circle can be done by slightly modifying the algorithm given in Section 5.3 of Chapter 5. In particular, $\Sigma_{\mathrm{r}}$ in Steps 5.Z.1 and 5.Z. 5 should be replaced by $\tilde{\Sigma}_{\text {P }}$ and Equation (5.3.2) should be replaced by the following

$$
\Gamma_{s \mathrm{P}}^{-1}\left[\tilde{E}_{\mathrm{P}}-\tilde{B}_{\mathrm{P}}\left(\tilde{D}_{2 \mathrm{P}}^{\prime} \tilde{D}_{2 \mathrm{P}}\right)^{\dagger} \tilde{D}_{2 \mathrm{P}}^{\prime} \tilde{D}_{22 \mathrm{P}}\right]=\left[\begin{array}{c}
E_{a \mathrm{P}}^{+}  \tag{8.3.3}\\
E_{b \mathrm{P}} \\
E_{a \mathrm{P}}^{0} \\
E_{a \mathrm{P}}^{-} \\
E_{c \mathrm{P}} \\
E_{d \mathrm{P}}
\end{array}\right]
$$

Also, $\Sigma_{\mathrm{Q}}^{\star}$ in Steps $5 . Z .2$ and 5.Z. 5 should be replaced by $\tilde{\Sigma}_{\mathrm{Q}}^{\star}$ and Equation (5.3.19) should be replaced by

$$
\Gamma_{s Q}^{-1}\left[\tilde{C}_{2 \mathrm{Q}}^{\prime}-\tilde{C}_{1 \mathrm{Q}}^{\prime}\left(\tilde{D}_{1 Q} \tilde{D}_{1 \mathrm{Q}}^{\prime}\right)^{\dagger} \tilde{D}_{1 \mathrm{Q}} \tilde{D}_{22 \mathrm{Q}}^{\prime}\right]=\left[\begin{array}{c}
E_{a \mathrm{Q}}^{+}  \tag{8.3.4}\\
E_{b \mathrm{Q}} \\
E_{a \mathrm{Q}}^{0} \\
E_{a Q}^{-} \\
E_{c \mathrm{Q}} \\
E_{d \mathrm{Q}}
\end{array}\right] .
$$

The rest of the algorithm does not need to be changed at all.

## Chapter 9

## Solutions to Discrete-time $H_{\infty}$ Problem

This Chapter is concerned with the discrete-time $H_{\infty}$ control problem with full state feedback, full information feedback and general measurement feedback. The objective is to present a solution to the discrete-time $H_{\infty}$ control problem. One way to approach this problem is to transform the discrete-time $H_{\infty}$ optimal control problem into an equivalents continuous-time $H_{\infty}$ control problem via bilinear transformation (see Chapter 4). Then the continuoustime controllers that are solutions to the auxiliary problem can be obtained and transformed back to their discrete-time equivalent using inverse bilinear transformation (see again Chapter 4). Another way is to solve this problem directly in discrete-time setting and in terms of the original system's performance. This approach leaves the possibility of directly observing the effect of certain physical parameters. Finally, a novel aspect of this chapter is that we show that if certain states or disturbances are observed directly, then this yields the possibility of deriving a reduced order controller. This result corresponds with the continuous-time reduced order controller structure of Chapter 6. The main results of this chapter are similar to those in [102], but the presentation is quite different.

### 9.1. Full Information and State Feedbacks

We first consider in this section the following full information feedback system,

$$
\left\{\begin{align*}
& x(k+1)=A \quad x(k)+B u(k)+E \quad w(k)  \tag{9.1.1}\\
& y(k)=\binom{I}{0} x(k) \\
& h(k)=C_{2} x(k)+D_{2} u(k)+D_{22} w(k) \\
& I
\end{align*}\right) w(k), ~ \$
$$

where $x \in \mathbb{R}^{n}$ is the state, $u \in \mathbb{R}^{m}$ is the control input, $w \in \mathbb{R}^{q}$ is the external disturbance input, $y \in \mathbb{R}^{n+q}$ is the measurement output, and $h \in \mathbb{R}^{l}$ is the controlled output of $\Sigma$. As usual, we define $\Sigma_{\mathrm{P}}$ to be the subsystem characterized by the matrix quadruple $\left(A, B, C_{2}, D_{2}\right)$. We assume that $\Sigma_{\mathrm{P}}$ has no invariant zero on the unit circle and its infimum is given by $\gamma^{*}$. We are interested in designing a full information feedback control law

$$
\begin{equation*}
u(k)=F_{1} x(k)+F_{2} w(k) \tag{9.1.2}
\end{equation*}
$$

such that when it is applied to the given system (9.1.1), the resulting closed-loop system is asymptotically stable and the resulting closed-loop transfer matrix from $w$ to $h$ has an $H_{\infty}$-norm less than a given $\gamma>\gamma^{*}$.

In what follows, we state a step-by-step algorithm for the computation of $F_{1}$ and $F_{2}$.

Step 9.F.1: Without loss of generality but for simplicity of presentation, we assume that the quadruple $\left(A, B, C_{2}, D_{2}\right)$, i.e., $\Sigma_{\mathbf{P}}$, has been partitioned in the form of (2.3.4). Then, transform $\Sigma_{\mathrm{P}}$ into the special coordinate basis as described in Chapter 2, i.e., find non-singular transformations $\Gamma_{s}, \Gamma_{i}$ and $\Gamma_{o}$ such that

$$
\begin{aligned}
\Gamma_{s}^{-1}\left(A-B_{0} C_{2,0}\right) \Gamma_{s} & =\left[\begin{array}{ccccc}
A_{c c} & B_{c} E_{c a}^{-} & B_{c} E_{c a}^{+} & L_{c d} C_{d} & L_{c b} C_{b} \\
0 & A_{a a}^{-} & 0 & L_{a d}^{-} C_{d} & L_{a b}^{-} C_{b} \\
0 & 0 & A_{a a}^{+} & L_{a d}^{+} C_{d} & L_{a b}^{+} C_{b} \\
B_{d} E_{d c} & B_{d} E_{d a}^{-} & B_{d} E_{d a}^{+} & A_{d d} & B_{d} E_{d b} \\
0 & 0 & 0 & L_{b d} C_{d} & A_{b b}
\end{array}\right], \\
\Gamma_{o}^{-1}\left[\begin{array}{l}
C_{2,0} \\
C_{2,1}
\end{array}\right] \Gamma_{s} & =\left[\begin{array}{ccccc}
C_{0 c} & C_{0 a}^{-} & C_{0 a}^{+} & C_{0 b} & C_{0 d} \\
0 & 0 & 0 & C_{d} & 0 \\
0 & 0 & 0 & 0 & C_{b}
\end{array}\right], \\
\Gamma_{s}^{-1}\left[\begin{array}{ll}
B_{0} & B_{1}
\end{array}\right] \Gamma_{i} & =\left[\begin{array}{ccc}
B_{0 c} & 0 & B_{c} \\
B_{0 a}^{-} & 0 & 0 \\
B_{0 a}^{+} & 0 & 0 \\
B_{0 d} & B_{d} & 0 \\
B_{0 b} & 0 & 0
\end{array}\right], \Gamma_{o}^{-1} D_{2} \Gamma_{i}=\left[\begin{array}{ccc}
I_{m_{0}} & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right] .
\end{aligned}
$$

Note that an additional permutation matrix to the state transformation has been introduced here to the original SCB such that the new state variables are ordered as follows:

$$
\tilde{x}=\left(\begin{array}{c}
x_{c}  \tag{9.1.3}\\
x_{a}^{-} \\
x_{a}^{+} \\
x_{d} \\
x_{b}
\end{array}\right) .
$$

Next, we compute

$$
\Gamma_{s}^{-1} E / \gamma=\left[\begin{array}{c}
E_{c}  \tag{9.1.4}\\
E_{a}^{-} \\
E_{a}^{+} \\
E_{d} \\
E_{b}
\end{array}\right]
$$

Step 9.F.2: Let $F_{c}$ be any appropriate dimensional constant matrix such that all the eigenvalues of $A_{c c}-B_{c} F_{c}$ are on the open unit disc. This can be done as $\left(A_{c c}, B_{c}\right)$ is completely controllable.

Step 9.F.3: Define $A_{x}, B_{x}, E_{x}, C_{x}$ and $D_{x}$ as follows:

$$
\begin{gather*}
A_{x}:=\left[\begin{array}{ccc}
A_{a a}^{+} & L_{a d}^{+} C_{d} & L_{a b}^{+} C_{b} \\
B_{d} E_{d a}^{+} & A_{d d} & B_{d} E_{d b} \\
0 & L_{b d} C_{d} & A_{b b}
\end{array}\right], \quad E_{x}:=\left[\begin{array}{c}
E_{a}^{+} \\
E_{d} \\
E_{b}
\end{array}\right],  \tag{9.1.5}\\
B_{x}:=\left[\begin{array}{cc}
B_{0 a}^{+} & 0 \\
B_{0 d} & B_{d} \\
B_{0 b} & 0
\end{array}\right], \quad D_{22 x}:=D_{22} / \gamma, \tag{9.1.6}
\end{gather*}
$$

and

$$
C_{x}:=\Gamma_{o}\left[\begin{array}{ccc}
0 & 0 & 0  \tag{9.1.7}\\
0 & C_{d} & 0 \\
0 & 0 & C_{b}
\end{array}\right], \quad D_{x}=\Gamma_{o}\left[\begin{array}{cc}
I_{m_{0}} & 0 \\
0 & 0 \\
0 & 0
\end{array}\right] .
$$

Step 9.F.4: Solve the following discrete-time algebraic Riccati equation:

$$
P_{x}=A_{x}^{\prime} P_{x} A_{x}+C_{x}^{\prime} C_{x}-\left[\begin{array}{c}
B_{x}^{\prime} P_{x} A_{x}+D_{x}^{\prime} C_{x}  \tag{9.1.8}\\
E_{x}^{\prime} P_{x} A_{x}+D_{22 x}^{\prime} C_{x}
\end{array}\right]^{\prime} G_{x}^{-1}\left[\begin{array}{c}
B_{x}^{\prime} P_{x} A_{x}+D_{x}^{\prime} C_{x} \\
E_{x}^{\prime} P_{x} A_{x}+D_{22 x}^{\prime} C_{x}
\end{array}\right]
$$

where

$$
G_{x}:=\left[\begin{array}{cc}
D_{x}^{\prime} D_{x}+B_{x}^{\prime} P_{x} B_{x} & B_{x}^{\prime} P_{x} E_{x}  \tag{9.1.9}\\
E_{x}^{\prime} P_{x} B_{x} & E_{x}^{\prime} P_{x} E_{x}+D_{22 x}^{\prime} D_{22 x}-I
\end{array}\right]
$$

for $P_{x}>0$. Note that because $\left(A_{x}, B_{x}, C_{x}, D_{x}\right)$ is left invertible and only has unstable invariant zeros, such a $P_{x}$ always exists provided that
$\gamma>\gamma^{*}$. In fact, one can use the very accurate method given previously in Chapter 4 to obtain this $P_{x}$. For future use in the output feedback case, we compute

$$
X=\left(\Gamma_{s}^{-1}\right)^{\prime}\left[\begin{array}{cc}
0 & 0  \tag{9.1.10}\\
0 & P_{x}
\end{array}\right] \Gamma_{s}^{-1}
$$

Step 9.F.5: Next, compute

$$
\begin{equation*}
F_{1 x}=\left(B_{x}^{\prime} P_{x} B_{x}+D_{x}^{\prime} D_{x}\right)^{-1}\left(B_{x}^{\prime} P_{x} A_{x}+D_{x}^{\prime} C_{x}\right) \tag{9.1.11}
\end{equation*}
$$

and

$$
\begin{equation*}
F_{2 x}=\left(B_{x}^{\prime} P_{x} B_{x}+D_{x}^{\prime} D_{x}\right)^{-1}\left(B_{x}^{\prime} P_{x} E_{x}+D_{x}^{\prime} D_{22 x}\right) \tag{9.1.12}
\end{equation*}
$$

Then, partition $F_{1 x}$ as follows:

$$
F_{1 x}=\left[\begin{array}{lll}
F_{0 a x}^{+} & F_{0 d x} & F_{0 b x}  \tag{9.1.13}\\
F_{d a x}^{+} & F_{d d x} & F_{d b x}
\end{array}\right] .
$$

Step 9.F.6: Finally, the gain matrices $F_{1}$ and $F_{2}$ are respectively given by

$$
F_{1}=-\Gamma_{i}\left[\begin{array}{ccccc}
C_{0 c} & C_{0 a}^{-} & C_{0 a}^{+}+F_{0 a x}^{+} & C_{0 d}+F_{0 d x} & C_{0 b}+F_{0 b x}  \tag{9.1.14}\\
E_{d c} & E_{d a}^{-} & F_{d a x}^{+} & F_{d d x} & F_{d b x} \\
F_{c} & \star & \star & \star & \star
\end{array}\right] \Gamma_{s}^{-1}
$$

and

$$
F_{2}=-\Gamma_{i}\left[\begin{array}{c}
F_{2 x}  \tag{9.1.15}\\
\star
\end{array}\right] \gamma
$$

where $\star$ 's are some arbitrary matrices with appropriate dimensions. 母

We have the following theorem.

Theorem 9.1.1. Consider the full information feedback discrete-time system (9.1.1). Then under the full information feedback law,

$$
\begin{equation*}
u(k)=F_{1} x(k)+F_{2} w(k) \tag{9.1.16}
\end{equation*}
$$

with $F_{1}$ and $F_{2}$ given by (9.1.14) and (9.1.15), respectively, the closed-loop system is asymptotically stable and the $H_{\infty}$-norm of the closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$ is less than $\gamma$. 团

Proof. It is straightforward to verify that the poles of the closed-loop system comprising the given full information system (9.1.1) with the control law (9.1.16) are given by $A_{c c}-B_{c} F_{c}, A_{a a}^{-}$and $A_{x}-B_{x} F_{1 x}$. We note that both $A_{c c}-B_{c} F_{c}$ and $A_{a a}^{-}$are asymptotically stable. Hence, the closed-loop system is stable if and only if $A_{x}-B_{x} F_{1 x}$ is stable. Moreover, it is also simple to show that its closed-loop transfer matrix from $w$ to $h$, say $T_{h w}$, is equal to $\gamma T_{h_{x} w_{x}}$, where $T_{h_{x} w_{x}}$ is the transfer matrix from $w_{x}$ to $h_{x}$ of the closed-loop system comprising the following auxiliary system,

$$
\left\{\begin{array}{l}
\dot{x}_{x}=A_{x} x_{x}+B_{x} u_{x}+E_{x} w_{x}  \tag{9.1.17}\\
y_{x}=\binom{I}{0} x_{x} \\
h_{x}=C_{x} x_{x}+D_{x} u_{x}+D_{22 x} w_{x}
\end{array}\right.
$$

with a full information control law,

$$
\begin{equation*}
u_{x}=-F_{1 x} x_{x}-F_{2 x} w_{x} \tag{9.1.18}
\end{equation*}
$$

Because ( $A_{x}, B_{x}, C_{x}, D_{x}$ ) is left invertible and has only unstable invariant zeros, it follows from the result of [100] that the solution to the Riccati equation (9.1.8) is indeed a positive definite one provided that $\gamma>\gamma^{*}$. Moreover, we also have $A_{x}-B_{x} F_{1 x}$ is asymptotically stable and $\left\|T_{h_{x} w_{x}}\right\|_{\infty}<1$. Hence, the result of Theorem 9.1.1 follows.

We illustrate the above result with a numerical example.
Example 9.1.1. Let us consider a discrete-time full information system (9.1.1) with matrices $A, B, E, C_{2}, D_{2}$ and $D_{22}$ are as given in Example 8.1.1 of Chapter 8. The infimum for this problem was computed in Example 8.1.1 and is given by $\gamma^{*}=0.934173$. Let us choose a $\gamma=0.934174$, which is slightly larger than $\gamma^{*}$. Following the above algorithm, we obtain

$$
F_{1}=\left[\begin{array}{rrrrr}
0 & 0 & -0.745354 & -1.078688 & -1.078688 \\
-1 & -1 & -1.412022 & -1.872678 & -1.872678 \\
-1 & 0 & 0 & 0 & 0
\end{array}\right]
$$

and

$$
F_{2}=\left[\begin{array}{r}
-0.872677 \\
-1.206011 \\
0
\end{array}\right]
$$

The closed-loop poles, i.e., $\lambda\left(A+B F_{1}\right)=\{0,0,0,0,0.38197\}$. The singular value plot of the closed-loop transfer matrix from $w$ to $h$ in Figure 9.1.1 clearly shows that its $H_{\infty}$-norm is less that the given $\gamma=0.934174$.


Figure 9.1.1: Singular values of $T_{h w}$ under full information feedback.

As was shown in Chapter 8, for discrete-time systems, the infimum associated with the given full information feedback system is in general different from that associated with its corresponding full state feedback system, i.e.,

$$
\left\{\begin{align*}
x(k+1) & =A x(k)+B u(k)+E w(k)  \tag{9.1.19}\\
y(k) & =x(k) \\
h(k) & =C_{2} x(k)+D_{2} u(k)+D_{22} w(k)
\end{align*}\right.
$$

Let $\gamma^{*}$ be the infimum associated with the full state feedback problem. Then, for any given $\gamma>\gamma^{*}$, the following algorithm will produce a static state feedback law that achieves the closed-loop stability as well as the required $H_{\infty}$-norm bound of the closed-loop transfer matrix from $w$ to $h$.

Step 9.S. 1 to 9.S.4: These steps are identical to Step 9.F. 1 to 9.F.4, respectively.
Step 9.S.5: Compute

$$
\begin{align*}
H_{x}:= & B_{x}^{\prime} P_{x} B_{x}+D_{x}^{\prime} D_{x}+\left(B_{x}^{\prime} P_{x} E_{x}+D_{x}^{\prime} D_{22 x}\right) \\
& \times\left(I-D_{22 x}^{\prime} D_{22 x}-E_{x}^{\prime} P_{x} E_{x}\right)^{-1}\left(E_{x}^{\prime} P_{x} B_{x}+D_{22 x}^{\prime} D_{x}\right), \tag{9.1.20}
\end{align*}
$$

and

$$
\begin{align*}
F_{x}:= & H_{x}^{-1}\left[B_{x}^{\prime} P_{x} A_{x}+D_{x}^{\prime} C_{x}+\left(B_{x}^{\prime} P_{x} E_{x}+D_{x}^{\prime} D_{22 x}\right)\right. \\
& \left.\times\left(I-D_{22 x}^{\prime} D_{22 x}-E_{x}^{\prime} P_{x} E_{x}\right)^{-1}\left(E_{x}^{\prime} P_{x} A_{x}+D_{22 x}^{\prime} C_{x}\right)\right] \tag{9.1.21}
\end{align*}
$$

Then, partition $F_{x}$ as follows:

$$
F_{x}=\left[\begin{array}{lll}
F_{0 a x}^{+} & F_{0 d x} & F_{0 b x}  \tag{9.1.22}\\
F_{d a x}^{+} & F_{d d x} & F_{d b x}
\end{array}\right] .
$$

Step 9.S.6: The gain matrix $F$ is given by

$$
F=-\Gamma_{i}\left[\begin{array}{ccccc}
C_{0 c} & C_{0 a}^{-} & C_{0 a}^{+}+F_{0 a x}^{+} & C_{0 d}+F_{0 d x} & C_{0 b}+F_{0 b x}  \tag{9.1.23}\\
E_{d c} & E_{d a}^{-} & F_{d a x}^{+} & F_{d d x} & F_{d b x} \\
F_{c} & \star & \star & \star & \star
\end{array}\right] \Gamma_{s}^{-1}
$$

where $\star$ 's are some arbitrary matrices with appropriate dimensions. 囚
Following the lines of reasoning similar to the proof of Theorem 9.1.1, one can show that the static control law,

$$
\begin{equation*}
u(k)=F x(k) \tag{9.1.24}
\end{equation*}
$$

with $F$ given by (9.1.23), will i) achieve the closed-loop stability, and ii) make the $H_{\infty}$-norm of the resulting closed-loop transfer matrix from $w$ to $h$ less than the given $\gamma$. We illustrate this in the following example.

Example 9.1.2. Let us consider a discrete-time full state feedback system (9.1.19) with matrices $A, B, E, C_{2}, D_{2}$ and $D_{22}$ are as given in Example 8.1.1 of Chapter 8. The infimum for this problem was computed in Example 8.2.1 and is given by $\gamma^{*}=3.181043$. Let us choose a $\gamma=3.181044$, which is slightly larger than $\gamma^{*}$. Following the above algorithm, we obtain

$$
F=\left[\begin{array}{rrrrr}
0 & 0 & -0.432563 & -0.885373 & -0.885373 \\
-1 & -1 & -1.479753 & -1.914538 & -1.914538 \\
-1 & 0 & 0 & 0 & 0
\end{array}\right]
$$

The closed-loop poles, i.e., $\lambda(A+B F)=\{0,0,0,0.27093,0.38197\}$ and the singular value plot of the closed-loop transfer matrix from $w$ to $h$ in Figure 9.1.2 clearly shows that its $H_{\infty}$-norm is less that the given $\gamma=3.181044$.

### 9.2. Full Order Output Feedback

We construct solutions to the discrete-time $H_{\infty}$ control problem for the following measurement feedback discrete-time system $\Sigma$,

$$
\left\{\begin{align*}
x(k+1) & =A x(k)+B u(k)  \tag{9.2.1}\\
y(k) & =E w(k), \\
& =C_{1} x(k) \\
h(k) & =D_{1} w(k)+D_{2} u(k)+D_{22} w(k),
\end{align*}\right.
$$



Figure 9.1.2: Singular values of $T_{h w}$ under full state feedback.
where $x \in \mathbb{R}^{n}$ is the state, $u \in \mathbb{R}^{m}$ is the control input, $w \in \mathbb{R}^{q}$ is the external disturbance input, $y \in \mathbb{R}^{p}$ is the measurement output, and $h \in \mathbb{R}^{\ell}$ is the controlled output of $\Sigma$. Again, for the purpose of easy reference, we define $\Sigma_{\mathrm{P}}$ to be the subsystem characterized by the matrix quadruple $\left(A, B, C_{2}, D_{2}\right)$ and $\Sigma_{\mathrm{Q}}$ to be the subsystem characterized by the matrix quadruple $\left(A, E, C_{1}, D_{1}\right)$. We assume in this section that both subsystems $\Sigma_{\mathrm{P}}$ and $\Sigma_{\mathrm{Q}}$ have no invariant zero on the unit circle.

Let $\gamma^{*}$ be the infimum for the given $\Sigma$ of (9.2.1). Given a positive scalar $\gamma>\gamma^{*}$, the following algorithm will produce a measurement feedback control law that achieves i) internal stability for the closed-loop system, and ii) the resulting $\left\|T_{h w}\right\|_{\infty}<\gamma$.

Step 9.M.1: Define an auxiliary full information problem for

$$
\left\{\begin{align*}
& x(k+1)=A x(k)+B u(k)+E \quad w(k)  \tag{9.2.2}\\
& y(k)=\binom{0}{I} x(k) \\
& h(k)=C_{2} x(k)+D_{2} u(k)+D_{22} w(k) \\
& 0
\end{align*}\right) w(k), ~ \$
$$

and perform Steps 9.F. 1 to $9 . F .4$ of the algorithm given in the previous section to get a positive semi-definite matrix $X$. Let $P:=X$ and compute

$$
\begin{equation*}
V:=B^{\prime} P B+D_{2}^{\prime} D_{2} \tag{9.2.3}
\end{equation*}
$$

and

$$
\begin{equation*}
R:=\gamma^{2} I-D_{22}^{\prime} D_{22}-E^{\prime} P E+\left(E^{\prime} P B+D_{22}^{\prime} D_{2}\right) V^{\dagger}\left(B^{\prime} P E+D_{2}^{\prime} D_{22}\right) \tag{9.2.4}
\end{equation*}
$$

where $\dagger$ denotes the Moore-Penrose (pseudo) inverse. It can be shown that $R>0$. Next, compute

$$
\begin{align*}
& A_{s}:=A-B V^{\dagger}\left(B^{\prime} P A+D_{2}^{\prime} C_{2}\right)  \tag{9.2.5}\\
& C_{s}:=C_{2}-D_{2} V^{\dagger}\left(B^{\prime} P A+D_{2}^{\prime} C_{2}\right) \tag{9.2.6}
\end{align*}
$$

and calculate

$$
\begin{aligned}
A_{\mathrm{P}} & :=A+E R^{-1}\left(E^{\prime} P A_{s}+D_{22}^{\prime} C_{s}\right), \\
E_{\mathrm{P}} & :=E R^{-\frac{1}{2}}, \\
C_{1 \mathrm{p}} & :=C_{1}+D_{1} R^{-1}\left(E^{\prime} P A_{s}+D_{22}^{\prime} C_{s}\right), \\
C_{2 \mathrm{P}} & :=\left(V^{\frac{1}{2}}\right)^{\dagger}\left[B^{\prime} P A+D_{2}^{\prime} C_{2}+\left(B^{\prime} P E+D_{2}^{\prime} D_{22}\right) R^{-1}\left(E^{\prime} P A_{s}+D_{22}^{\prime} C_{s}\right)\right] \\
D_{1 \mathrm{p}} & :=D_{1} R^{-\frac{1}{2}}, \\
D_{2 \mathrm{P}} & :=V^{\frac{1}{2}}, \\
D_{22 \mathrm{P}} & :=\left(V^{\frac{1}{2}}\right)^{\dagger}\left(B^{\prime} P E+D_{2}^{\prime} D_{22}\right) R^{-\frac{1}{2}} .
\end{aligned}
$$

Step 9.M.2: Define another auxiliary full information problem for
and again perform Steps 9.F. 1 to $9 . F .4$ of the algorithm in the previous section to get another positive semi-definite matrix $X$ and let $Q:=X$. Also, let

$$
\begin{equation*}
Y:=\left(\gamma^{2} I-Q P\right)^{-1} Q \tag{9.2.8}
\end{equation*}
$$

Step 9.M.3: Next, compute

$$
\begin{align*}
W_{\mathrm{P}}:= & D_{1 \mathrm{P}} D_{1 \mathrm{P}}^{\prime}+C_{1 \mathrm{P}} Y C_{1 \mathrm{P}}^{\prime}, \\
S_{\mathrm{P}}:= & \left(C_{2 \mathrm{P}} Y C_{1 \mathrm{P}}+D_{22 \mathrm{P}} D_{1 \mathrm{P}}^{\prime}\right) W_{\mathrm{P}}^{\dagger}\left(C_{1 \mathrm{P}} Y C_{2 \mathrm{P}}+D_{1 \mathrm{P}} D_{22 \mathrm{P}}^{\prime}\right) \\
& +\gamma^{2} I-D_{22 \mathrm{P}} D_{22 \mathrm{P}}^{\prime}-C_{2 \mathrm{P}} Y C_{2 \mathrm{P}}^{\prime},  \tag{9.2.9}\\
A_{z}:= & A_{\mathrm{P}}-\left(A_{\mathrm{P}} Y C_{1 \mathrm{P}}^{\prime}+E_{\mathrm{P}} D_{1 \mathrm{P}}^{\prime}\right) W_{\mathrm{P}}^{\dagger} C_{1 \mathrm{P}}, \\
E_{z}:= & E_{\mathrm{P}}-\left(A_{\mathrm{P}} Y C_{1 \mathrm{P}}^{\prime}+E_{\mathrm{P}} D_{1 \mathrm{P}}^{\prime}\right) W_{\mathrm{P}}^{\dagger} D_{1 \mathrm{P}},
\end{align*}
$$

and

$$
\begin{align*}
A_{\mathrm{PY}}:= & A_{\mathrm{P}}+\left(A_{z} Y C_{2 \mathrm{P}}^{\prime}+E_{z} D_{22 \mathrm{P}}^{\prime}\right) S_{\mathrm{P}}^{-1} C_{2 \mathrm{P}}, \\
B_{\mathrm{PY}}:= & B+\left(A_{z} Y C_{2 \mathrm{P}}+E_{z} D_{22 \mathrm{P}}^{\prime}\right) S_{\mathrm{P}}^{-1} D_{2 \mathrm{P}}, \\
E_{\mathrm{PY}}:= & {\left[\left(A_{z} Y C_{2 \mathrm{P}}+E_{z} D_{22 \mathrm{P}}^{\prime}\right) S_{\mathrm{P}}^{-1}\left(C_{2 \mathrm{P}} Y C_{1 \mathrm{P}}+D_{22 \mathrm{P}} D_{1 \mathrm{P}}^{\prime}\right)\right.} \\
& \left.+A_{\mathrm{P}} Y C_{1 \mathrm{P}}^{\prime}+E_{\mathrm{P}} D_{1 \mathrm{P}}^{\prime}\right]\left(W_{\mathrm{P}}^{\frac{1}{2}}\right)^{\dagger},  \tag{9.2.10}\\
& \\
C_{2 \mathrm{PY}}:= & S_{\mathrm{P}}^{-\frac{1}{2}} C_{2 \mathrm{P}}, \\
D_{1 \mathrm{PY}}:= & W_{\mathrm{P}}^{\frac{1}{2}} \\
D_{2 \mathrm{PY}}:= & S_{\mathrm{P}}^{-\frac{1}{2}} D_{2 \mathrm{P}}, \\
D_{22 \mathrm{PY}}:= & S_{\mathrm{P}}^{-\frac{1}{2}}\left(C_{2 \mathrm{P}} Y C_{1 \mathrm{P}}^{\prime}+D_{22 \mathrm{P}} D_{1 \mathrm{P}}^{\prime}\right)\left(W_{\mathrm{P}}^{\frac{1}{2}}\right)^{\dagger} .
\end{align*}
$$

It can be shown that i) the quadruple ( $A_{\mathrm{PY}}, B_{\mathrm{PY}}, C_{2 \mathrm{PY}}, D_{2 \mathrm{PY}}$ ) is right invertible and of minimum phase with no infinite zero, and ii) the quadruple $\left(A_{\mathrm{PY}}, E_{\mathrm{PY}}, C_{1 \mathrm{P}}, D_{1 \mathrm{PY}}\right)$ is left invertible and of minimum phase with no infinite zero. Moreover, there exists an appropriate constant matrix $X_{P Y}$ such that $D_{2 \mathrm{PY}}+D_{2 \mathrm{PY}} X_{\mathrm{PY}} D_{1 \mathrm{PY}}=0$.

Step 9.M.4: Let

$$
\begin{align*}
& F_{1 \mathrm{PY}}:=-D_{2 \mathrm{P}}^{\dagger} C_{2 \mathrm{P}}+\left(I-D_{2 \mathrm{P}}^{\dagger} D_{2 \mathrm{P}}\right) F_{0}  \tag{9.2.11}\\
& F_{2 \mathrm{PY}}:=-D_{2 \mathrm{PY}}^{\dagger} D_{22 \mathrm{PY}} \tag{9.2.12}
\end{align*}
$$

where $F_{0}$ is such that $A_{\mathrm{P}}+B F_{1 \mathrm{PY}}=A_{\mathrm{PY}}+B_{\mathrm{PY}} F_{1 \mathrm{PY}}$ has all its eigenvalues inside the unit circle. Also, let

$$
\begin{align*}
& K_{1 \mathrm{PY}}:=-E_{\mathrm{PY}} D_{1 \mathrm{PY}}^{\dagger}+K_{0}\left(I-D_{1 \mathrm{PY}} D_{1 \mathrm{PY}}^{\dagger}\right)  \tag{9.2.13}\\
& K_{2 \mathrm{PY}}:=-D_{1 \mathrm{PY}}^{\dagger} \tag{9.2.14}
\end{align*}
$$

where $K_{0}$ is such that $A_{\mathrm{PY}}+K_{1 \mathrm{PY}} C_{1 \mathrm{P}}$ is stable. We would like to note that a more systematic procedure to compute the above gain matrices will be given in the next chapter.

Step 9.M.5: Finally, we obtain a measurement output feedback control law,

$$
\Sigma_{\mathrm{cmp}}:\left\{\begin{align*}
v(k+1) & =A_{\mathrm{cmp}} v(k)+B_{\mathrm{cmp}} y(k)  \tag{9.2.15}\\
u(k) & =C_{\mathrm{cmp}} v(k)+D_{\mathrm{cmp}} y(k)
\end{align*}\right.
$$

with

$$
\left.\begin{array}{l}
D_{\mathrm{cmp}}:=-F_{2 \mathrm{PY}} K_{2 \mathrm{PY}},  \tag{9.2.16}\\
C_{\mathrm{cmp}}:=F_{1 \mathrm{PY}}-D_{\mathrm{cmp}} C_{1 \mathrm{P}}, \\
B_{\mathrm{cmp}}:=B_{\mathrm{PY}} D_{\mathrm{cmp}}-K_{1 \mathrm{PY}}, \\
A_{\mathrm{cmp}}:=A_{\mathrm{PY}}+B_{\mathrm{PY}} C_{\mathrm{cmp}}+K_{1 \mathrm{PY}} C_{1 \mathrm{p}} .
\end{array}\right\}
$$

Clearly, $v \in \mathbf{R}^{n}$, i.e., the obtained controller $\Sigma_{\text {cmp }}$ has the same dynamical order as that of the given system $\Sigma$.

We have the following theorem.
Theorem 9.2.1. Consider the given discrete-time system $\Sigma$ of (9.2.1) and the controller $\Sigma_{\mathrm{cmp}}$ of (9.2.15) with $A_{\mathrm{cmp}}, B_{\mathrm{cmp}}, C_{\mathrm{cmp}}$ and $D_{\mathrm{cmp}}$ being given by (9.2.16). Also, let $\gamma>\gamma^{*}$ be given. Then, we have

1. the resulting closed-loop system comprising $\Sigma$ and $\Sigma_{\mathrm{cmp}}$ is asymptotically stable; and
2. the $H_{\infty}$-norm of the closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$ is less than $\gamma$.

Proof. The proof of the above theorem can be carried out in two stages: The first stage involves showing that the following two statements are equivalent:

1. The closed-loop system comprising the given system $\Sigma$ of (9.2.1) and the controller $\Sigma_{\mathrm{cmp}}$ of (9.2.15) is internally stable and its transfer matrix from $w$ to $h, T_{h w}\left(\Sigma \times \Sigma_{\mathrm{cmp}}\right)$, has an $H_{\infty}$-norm less than $\gamma$.
2. The closed-loop system comprising an auxiliary system $\Sigma_{\mathrm{PY}}$, where $\Sigma_{\mathrm{PY}}$ is given by

$$
\left\{\begin{align*}
x_{\mathrm{PY}}(k+1) & =A_{\mathrm{PY}} x_{\mathrm{PY}}(k)+B_{\mathrm{PY}} u(k)+E_{\mathrm{PY}} w_{\mathrm{PY}}(k),  \tag{9.2.17}\\
y(k) & =C_{1 \mathrm{P}} x_{\mathrm{PY}}(k) \\
& +D_{1 \mathrm{PY}} w_{\mathrm{PY}}(k), \\
h_{\mathrm{PY}}(k) & =C_{2 \mathrm{PY}} x_{\mathrm{PY}}(k)+D_{2 \mathrm{PY}} u(k)+D_{22 \mathrm{PY}} w_{\mathrm{PY}}(k),
\end{align*}\right.
$$

and the controller $\Sigma_{\mathrm{cmp}}$ of (9.2.15) is internally stable and its transfer matrix from $w_{\mathrm{PY}}$ to $h_{\mathrm{PY}}, T_{h_{\mathrm{PY}} w_{\mathrm{PY}}}\left(\Sigma_{\mathrm{PY}} \times \Sigma_{\mathrm{cmp}}\right)$, has an $H_{\infty}$-norm less than $\gamma$.

The second stage involves showing that the transfer matrix from $w_{P Y}$ to $h_{P Y}$ of the closed-loop system comprising $\Sigma_{\mathrm{PY}}$ and $\Sigma_{\mathrm{cmp}}$ is internally stable and is in fact identically zero for all frequencies, i.e., $T_{h_{P Y} w_{P Y}}\left(\Sigma_{\mathrm{PY}} \times \Sigma_{\mathrm{cmp}}\right) \equiv 0$. It is obvious that $\left\|T_{h_{P Y} w_{P Y}}\left(\Sigma_{P Y} \times \Sigma_{\mathrm{cmp}}\right)\right\|_{\infty}=0<\gamma$. Hence, $\Sigma \times \Sigma_{\mathrm{cmp}}$ is internally stable and $\left\|T_{h w}\left(\Sigma \times \Sigma_{\text {cmp }}\right)\right\|_{\infty}<\gamma$. We refer interested readers to [102] for more detailed proofs of the above two facts (stages).

Remark 9.2.1. It is clear from the above proof that the design of a $\gamma$-suboptimal control law for the original system (9.2.1) is equivalent to the finding of a control law that solves the $H_{\infty}$ disturbance decoupling problem with internal stability for the auxiliary system (9.2.17). One can use a more systematic procedure given in Chapter 10 to find such a control law.

The following is an illustrative example.

Example 9.2.1. Let us consider a discrete-time system (9.2.1) with matrices $A, B, E, C_{2}, D_{2}$ and $D_{22}$ are as given in Example 8.1.1 of Chapter 8 and

$$
C_{1}=\left[\begin{array}{lllll}
0 & 0 & 0 & 0 & 1 \tag{9.2.18}
\end{array}\right], \quad D_{1}=0 .
$$

The infimum for this problem was computed in Example 8.2.1 and is given by $\gamma^{*}=15.16907$. Let us choose a positive scalar $\gamma=15.17$. Following our algorithm, we obtain a full order output feedback control law (9.2.15) with

$$
\begin{gathered}
A_{\text {cmp }}=\left[\begin{array}{rrrr}
0 & 1 & 1.005710 & 1.003529 \\
0 & 0 & 0.005710 & 1.003529 \\
0 & 0 & 0.691710 & 0.191432 \\
\hline & -1.8763781 \\
0 & 0 & -0.310193 & -0.809744 \\
0 & 0 & 0 & 3.217071 \\
1 & -3.281899
\end{array}\right], \\
B_{\text {cmp }}
\end{gathered}=\left[\begin{array}{r}
10.519757 \\
4.307309 \\
2.067505 \\
-4.026815 \\
4.281899
\end{array}\right], \quad D_{\mathrm{cmp}}=\left[\begin{array}{r}
-4.043756 \\
-14.546573 \\
0
\end{array}\right],
$$

and

$$
C_{\mathrm{cmp}}=\left[\begin{array}{rrrrr}
0 & 0 & -0.314000 & -0.812097 & 3.231659 \\
-1 & -1 & -1.315903 & -1.813273 & 12.733300 \\
-1 & 0 & 0 & 0 & 0
\end{array}\right]
$$

The plot of the singular values of the closed-loop transfer matrix from $w$ to $h$ in Figure 9.2 .1 shows that

$$
\begin{equation*}
\left\|T_{h w}\left(\Sigma \times \Sigma_{\mathrm{cmp}}\right)\right\|_{\infty}<\gamma=15.17 \tag{9.2.19}
\end{equation*}
$$

The poles of the closed-loop system are given by

$$
-0.596025,0.618045,0.433068,0.382376,-0.237186,-0.000212,0,0,0,0
$$

which are all inside the unit circle.

### 9.3. Reduced Order Output Feedback

In this section we show that for the singular $H_{\infty}$ control problem, we can always find a suboptimal solution which has a dynamical order less than that of the plant and is of a reduced order observer-based structure. This result is analogous to that obtained in Chapter 6 for the continuous-time problems.


Figure 9.2.1: Singular values of $T_{h w}$ under full order output feedback.

Without loss of generality, we develop such a reduced order observer-based controller for the system $\Sigma_{\text {PY }}$ defined in the previous section, i.e.,

$$
\left\{\begin{align*}
x_{\mathrm{PY}}(k+1) & =A_{\mathrm{PY}} x_{\mathrm{PY}}(k)+B_{\mathrm{PY}} u(k)  \tag{9.3.1}\\
y(k) & =E_{1 \mathrm{PY}} w_{\mathrm{PY}}(k), \\
& +D_{1 \mathrm{PY}} w_{\mathrm{PY}}(k), \\
h_{\mathrm{PY}}(k) & =C_{2 \mathrm{PY}} x_{\mathrm{PY}}(k)+D_{2 \mathrm{PY}} u(k)+D_{22 \mathrm{PY}} w_{\mathrm{PY}}(k) .
\end{align*}\right.
$$

There exists a constant output pre-feedback law $X_{\mathrm{PY}} y$ such that after applying this pre-feedback law, namely setting

$$
\begin{equation*}
u \longrightarrow X_{\mathrm{PY}} y+u \tag{9.3.2}
\end{equation*}
$$

the direct feed-through term from $w_{\mathrm{PY}}$ from $h_{\mathrm{PY}}$ disappears. Hence without loss of generality, hereafter we assume that $D_{22_{\mathrm{PY}}}=0$.

There exists an 'optimal' state feedback gain $F_{P Y}$ in the sense that

$$
\left(C_{2 \mathrm{PY}}+D_{2 \mathrm{PY}} F_{\mathrm{PY}}\right)\left(s I-A_{\mathrm{PY}}-B_{\mathrm{PY}} F_{\mathrm{PY}}\right)^{-1} E_{\mathrm{PY}} \equiv 0
$$

with $A_{\mathrm{PY}}+B_{\mathrm{PY}} F_{\mathrm{PY}}$ stable. We need to construct an observer of low order. Without loss of generality but for simplicity of presentation, we assume that the matrices $C_{1 \mathrm{P}}$ and $D_{1 \mathrm{PY}}$ are already in the form

$$
C_{1 \mathrm{P}}=\left[\begin{array}{cc}
0 & C_{1,02}  \tag{9.3.3}\\
I_{p-m_{0}} & 0
\end{array}\right] \quad \text { and } \quad D_{1 \mathrm{PY}}=\left[\begin{array}{c}
D_{1,0} \\
0
\end{array}\right]
$$

where $m_{0}$ is the rank of $D_{1 \mathrm{PY}}$ and $D_{1,0}$ is of full rank. Then the given system $\Sigma_{\mathrm{PY}}$ can be written as,
where

$$
\begin{equation*}
\binom{x_{1}}{x_{2}}=x_{\mathrm{PY}} \quad \text { and } \quad\binom{y_{0}}{y_{1}}=\boldsymbol{y} \tag{9.3.5}
\end{equation*}
$$

We note that $y_{1} \equiv x_{1}$. Thus, one needs to estimate only the state $x_{2}$ in the reduced-order estimator. Then following closely the procedure given in [22], we first rewrite the state equation for $x_{1}$ in terms of the measured output $y_{1}$ and state $x_{2}$ as follows,

$$
\begin{equation*}
y_{1}(k+1)=A_{11} y_{1}(k)+A_{12} x_{2}(k)+E_{1} w_{\mathrm{PY}}(k)+B_{1} u(k), \tag{9.3.6}
\end{equation*}
$$

where $y_{1}$ and $u$ are known. Observation of $x_{2}$ is made via $y_{0}$ and

$$
\begin{equation*}
\tilde{y}_{1}(k)=A_{12} x_{2}(k)+E_{1} w_{\mathrm{PY}}(k)=y_{1}(k+1)-A_{11} y_{1}(k)-B_{1} u(k) \tag{9.3.7}
\end{equation*}
$$

A reduced-order system for the estimation of state $x_{2}$ is given by

$$
\left\{\begin{align*}
x_{2}(k+1) & =A_{\mathrm{R}} x_{2}(k)+E_{\mathrm{R}} w_{\mathrm{PY}}(k)+\left[\begin{array}{ll}
A_{21} & B_{2}
\end{array}\right]\binom{y_{1}(k)}{u(k)},  \tag{9.3.8}\\
y_{\mathrm{R}}(k) & =C_{\mathrm{R}} x_{2}(k)+D_{\mathrm{R}} w_{\mathrm{PY}}(k)
\end{align*}\right.
$$

where

$$
A_{\mathrm{R}}:=A_{22}, \quad E_{\mathrm{R}}:=E_{2}, \quad C_{\mathrm{R}}:=\left[\begin{array}{c}
C_{1,02}  \tag{9.3.9}\\
A_{12}
\end{array}\right], \quad D_{\mathrm{R}}:=\left[\begin{array}{c}
D_{1,0} \\
E_{1}
\end{array}\right] .
$$

Based on (9.3.8), one can construct a reduced-order observer for $x_{2}$ as,

$$
\hat{x}_{2}(k+1)=A_{\mathrm{R}} \hat{x}_{2}(k)+\left[\begin{array}{ll}
A_{21} & B_{2} \tag{9.3.10}
\end{array}\right]\binom{y_{1}(k)}{u(k)}+K_{\mathrm{R}}\left[C_{\mathrm{R}} \hat{x}_{2}(k)-y_{\mathrm{R}}(k)\right]
$$

where $K_{\mathrm{R}}$ is the observer gain matrix which must be chosen such that $A_{\mathrm{R}}+K_{\mathrm{R}} C_{\mathrm{R}}$ is asymptotically stable and

$$
\begin{equation*}
\left(z I-A_{\mathrm{R}}-K_{\mathrm{R}} C_{\mathrm{R}}\right)^{-1}\left(E_{\mathrm{R}}-K_{\mathrm{R}} D_{\mathrm{R}}\right) \equiv 0 \tag{9.3.11}
\end{equation*}
$$

Following the result of Chen [10], i.e., Proposition 2.2.1, one can show that the quadruple ( $A_{\mathrm{R}}, E_{\mathrm{R}}, C_{\mathrm{R}}, D_{\mathrm{R}}$ ) is left invertible and of minimum phase with no
infinite zero, provided that the quadruple ( $A_{\mathrm{PY}}, E_{\mathrm{PY}}, C_{1 \mathrm{PY}}, D_{1 \mathrm{PY}}$ ) is left invertible and of minimum phase with no infinite zero. The computation of $K_{\mathrm{r}}$ can systematically be done using the procedure given in the next chapter.

At this moment we have a reduced-order observer and an optimal state feedback. However, $y_{\mathrm{R}}$ contains a future measurement, i.e., the term $y_{1}(k+1)$ in (9.3.7). We apply a transformation to remove this term. We partition the reduced order observer gain $K_{\mathrm{R}}=\left[K_{\mathrm{R} 0}, K_{\mathrm{R} 1}\right]$ compatible with the dimensions of the outputs $\left(y_{0}^{\prime}, \tilde{y}_{1}^{\prime}\right)^{\prime}$, and at the same time define a new variable,

$$
v:=\hat{x}_{2}+K_{\mathrm{R} 1} \tilde{y}_{1} .
$$

We then obtain the following reduced order estimator based controller,

$$
\left\{\begin{array}{rl}
v(k+1) & =\left(A_{\mathrm{R}}+K_{\mathrm{R}} C_{\mathrm{R}}\right)  \tag{9.3.12}\\
\hat{x}_{\mathrm{PY}}(k) & =\left[\begin{array}{c}
0 \\
I_{n-p+m_{0}}
\end{array}\right] \quad v(k)+\left(B_{2}+K_{\mathrm{R} 1} B_{1}\right) u(k)+G_{\mathrm{R}} y(k) \\
u(k) & =F_{\mathrm{PY}} \\
0 & I \\
0 & -K_{\mathrm{R} 1}
\end{array}\right] y(k), \quad \hat{x}_{\mathrm{PY}}(k)+\begin{array}{cc}
X_{\mathrm{PY}} & y(k)
\end{array}
$$

where

$$
G_{\mathrm{R}}=\left[-K_{\mathrm{R} 0}, A_{21}+K_{\mathrm{R} 1} A_{11}-\left(A_{\mathrm{R}}+K_{\mathrm{R}} C_{\mathrm{R}}\right) K_{\mathrm{R} 1}\right]
$$

and $F_{P Y}$ is state feedback gain and $X_{P Y}$ is the output pre-feedback gain.
Remark 9.3.1. It is interesting to point out that the state space representation of the reduced order estimator based controller in (9.3.12) might not be minimal and hence the McMillan degree of this controller might be less than the dynamical order of its state space representation (9.3.12). This is mainly due to the stable dynamics which become unobservable in the controlled output $h_{\text {PY }}$ after the preliminary output feedback law (9.3.2).

A very interesting example is the state feedback case for $C_{1}=I$ and $D_{1}=0$. In this case, the preliminary output feedback $X_{P Y}$ in (9.3.2) can be chosen such that after this preliminary feedback $C_{2 \mathrm{PY}}=0$ and $A_{\mathrm{PY}}$ is stable. Hence we can choose $F_{P Y}=0$ but this implies that the reduced order estimator based controller (9.3.12) has a McMillan degree equal to zero and it reduces to the static state feedback solution, $u=X_{\mathrm{PY}} y$.

Finally, we note that the reduced order output feedback control law (9.3.12) can be written in the following standard form,

$$
\Sigma_{\mathrm{cmp}}:\left\{\begin{align*}
v(k+1) & =A_{\mathrm{cmp}} v(k)+B_{\mathrm{cmp}} y(k),  \tag{9.3.13}\\
u(k) & =C_{\mathrm{cmp}} v(k)+D_{\mathrm{cmp}} y(k),
\end{align*}\right.
$$

with

$$
\begin{align*}
& A_{\mathrm{cmp}}:=\left(A_{\mathrm{R}}+K_{\mathrm{R}} C_{\mathrm{R}}\right)+\left(B_{2}+K_{\mathrm{R} 1} B_{1}\right) F_{\mathrm{PY}}\left[\begin{array}{l}
0 \\
I
\end{array}\right] \\
& B_{\mathrm{cmp}}:=\left(B_{2}+K_{\mathrm{R} 1} B_{1}\right)\left(F_{\mathrm{PY}}\left[\begin{array}{cc}
0 & I \\
0 & -K_{\mathrm{R} 1}
\end{array}\right]+X_{\mathrm{PY}}\right)+G_{\mathrm{R}} \\
& C_{\mathrm{cmp}}:=F_{\mathrm{PY}}\left[\begin{array}{l}
0 \\
I
\end{array}\right],  \tag{9.3.14}\\
& D_{\mathrm{cmp}}:=F_{\mathrm{PY}}\left[\begin{array}{cc}
0 & I \\
0 & -K_{\mathrm{R} 1}
\end{array}\right]+X_{\mathrm{PY}} .
\end{align*}
$$

We have the following theorem.
Theorem 9.3.1. Consider the given discrete-time system $\Sigma$ of (9.2.1). Also, let $\gamma>\gamma^{*}$ be given. Then, there exist gain matrices $X_{\mathrm{PY}}, F_{\mathrm{PY}}$ and $K_{\mathrm{R}}$ such that the resulting controller $\Sigma_{\mathrm{cmp}}$ of (9.3.13) with $A_{\mathrm{cmp}}, B_{\mathrm{cmp}}, C_{\mathrm{cmp}}$ and $D_{\mathrm{cmp}}$ being given as in (9.3.14) has the following properties:

1. the resulting closed-loop system comprising $\Sigma$ and $\Sigma_{\mathrm{cmp}}$ is asymptotically stable; and
2. the $H_{\infty}$-norm of the resulting closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$ is less than $\gamma$.

Proof. It is quite obvious because $\Sigma_{\text {PY }}$ has the following properties:

1. There exists a constant matrix $X_{\mathrm{PY}}$ such that $D_{2 \mathrm{PY}}+D_{2 \mathrm{PY}} X_{\mathrm{PY}} D_{1 \mathrm{PY}}=0$;
2. $\left(A_{\mathrm{PY}}, B_{\mathrm{PY}}, C_{2 \mathrm{PY}}, D_{2 \mathrm{PY}}\right)$ is right invertible and of minimum phase with no infinite zero;
3. $\left(A_{\mathrm{PY}}, E_{\mathrm{PY}}, C_{1 \mathrm{P}}, D_{1 \mathrm{PY}}\right)$ is left invertible and of minimum phase with no infinite zero.

A systematic procedure for computing the gain matrices $X_{\mathrm{PY}}, F_{\mathrm{PY}}$ and $K_{\mathrm{R}}$ can be found in Chapter 10.

The following example illustrates the result of this section.
Example 9.3.1. Consider a discrete-time system of the form (9.2.1) with

$$
\begin{gather*}
A=\left[\begin{array}{ll}
1 & 1 \\
2 & 3
\end{array}\right], \quad B=\left[\begin{array}{r}
-2 \\
1
\end{array}\right], \quad E=\left[\begin{array}{l}
1 \\
1
\end{array}\right],  \tag{9.3.15}\\
C_{1}=\left[\begin{array}{ll}
0 & 1 \\
1 & 0
\end{array}\right], \quad D_{1}=\left[\begin{array}{l}
1 \\
0
\end{array}\right], \tag{9.3.16}
\end{gather*}
$$

and

$$
C_{2}=\left[\begin{array}{ll}
0.8 & 0.9 \tag{9.3.17}
\end{array}\right], \quad D_{2}=0, \quad D_{22}=1
$$

It is simple to verify that the subsystem $\left(A, B, C_{2}, D_{2}\right)$ is invertible with an unstable invariant zero at 1.5714 and the subsystem $\left(A, E, C_{1}, D_{1}\right)$ is left invertible with an unstable invariant zero at 2. By utilizing the algorithm for computing $\gamma^{*}$ in the previous chapter, we obtain an exact value of the infimum

$$
\gamma^{*}=3.9631638
$$

In what follows, we will design a $\gamma$-suboptimal measurement output control law with $\gamma=3.963164$. Following the above procedures, we obtain an auxiliary system (9.3.1) with

$$
\left.\begin{array}{c}
A_{\mathrm{PY}}=\left[\begin{array}{ll}
1.14353033 & 1.18520854 \\
2.34861499 & 3.46328599
\end{array}\right], \quad B_{\mathrm{PY}}=\left[\begin{array}{r}
-2 \\
0.96422578
\end{array}\right], \\
E_{\mathrm{PY}}=10^{3} \cdot\left[\begin{array}{l}
1.65382390 \\
4.83262217
\end{array}\right], \quad C_{1 \mathrm{P}}=\left[\begin{array}{rr}
0.14353033 & 1.18520854 \\
1 & 0
\end{array}\right], \\
D_{1 \mathrm{PY}}=10^{3} \cdot\left[\begin{array}{r}
1.65382390 \\
0
\end{array}\right], \quad D_{22 \mathrm{PY}}=-3297.4252, \\
C_{2 \mathrm{PY}}=[-1.74280115
\end{array}-2.36309110\right], \quad D_{2 \mathrm{PY}}=0.30400789, ~ \$
$$

and finally the controller parameters,

$$
A_{\mathrm{cmp}}=0, \quad B_{\mathrm{cmp}}=\left[\begin{array}{ll}
0.06254887 & 0.05328328
\end{array}\right]
$$

and

$$
C_{\mathrm{cmp}}=0, \quad D_{\mathrm{cmp}}=\left[\begin{array}{ll}
6.55844429 & 4.79141397
\end{array}\right]
$$

The poles of the closed-loop system comprising the given plant and the above controller are given by 0 and $0.4878 \pm j 0.1199$. Clearly, they are stable. The plot of the singular values of the closed-loop transfer matrix from $w$ to $h$ in Figure 9.3.1 shows that $\left\|T_{h w}\left(\Sigma \times \Sigma_{\mathrm{cmp}}\right)\right\|_{\infty}$ is indeed less than the given $\gamma$. ${ }^{\text {® }}$


Figure 9.3.1: Singular values of $T_{h w}$ under reduced order output feedback.

## Chapter 10

## Discrete-time $H_{\infty}$ Almost Disturbance Decoupling

### 10.1. Introduction

In THIS CHAPTER, we consider the problem of $H_{\infty}$ almost disturbance decoupling for general discrete-time plants whose subsystems are allowed to have invariant zeros on the unit circle of the complex plane. The stability region of a discrete-time system considered in this chapter is defined as usual as the open unit disc. In contrast to the continuous-time case, the problem of almost disturbance decoupling for general discrete-time systems is less studied in the literature. In 1996, Chen, Guo and Lin [17] gave a set of solvability conditions for the $H_{\infty}$-ADDPMS for the special case when a given plant whose subsystems do not have invariant zeros on the unit circle. Only very recently, has the necessary and sufficient conditions under which the $H_{\infty}$-ADDPMS for general discrete-time systems been derived by Chen, He and Chen [18]. Solutions to such a general problem have just been reported by Lin and Chen [65]. The results of [18] and [65] form the core of this chapter.

To be more specific, we consider the following standard linear time-invariant discrete-time system $\Sigma$ characterized by

$$
\Sigma:\left\{\begin{align*}
x(k+1) & =A x(k)+B u(k)  \tag{10.1.1}\\
y(k) & =C_{1} x(k) \\
& +D_{1} w(k), \\
h(k) & =C_{2} x(k)+D_{2} u(k)+D_{22} w(k),
\end{align*}\right.
$$

where $x \in \mathbb{R}^{n}$ is the state, $u \in \mathbb{R}^{m}$ is the control input, $y \in \mathbb{R}^{\ell}$ is the measurement, $w \in \mathbb{R}^{q}$ is the disturbance and $h \in \mathbb{R}^{p}$ is the output to be controlled.

As usual, we denote $\Sigma_{\mathrm{P}}$ and $\Sigma_{\mathrm{Q}}$ as the subsystems characterized by matrix quadruples $\left(A, B, C_{2}, D_{2}\right)$ and $\left(A, E, C_{1}, D_{1}\right)$, respectively. The following dynamic feedback control laws are investigated:

$$
\Sigma_{\mathrm{cmp}}:\left\{\begin{align*}
x_{c}(k+1) & =A_{\mathrm{cmp}} x_{c}(k)+B_{\mathrm{cmp}} y(k)  \tag{10.1.2}\\
u(k) & =C_{\mathrm{cmp}} x_{c}(k)+D_{\mathrm{cmp}} y(k)
\end{align*}\right.
$$

The controller $\Sigma_{\mathrm{cmp}}$ of (10.1.2) is said to be internally stabilizing when applied to the system $\Sigma$, if the following matrix is asymptotically stable:

$$
A_{\mathrm{cl}}:=\left[\begin{array}{cc}
A+B D_{\mathrm{cmp}} C_{1} & B C_{\mathrm{cmp}}  \tag{10.1.3}\\
B_{\mathrm{cmp}} C_{1} & A_{\mathrm{cmp}}
\end{array}\right]
$$

i.e., all its eigenvalues lie inside the open unit disc of the complex plane. Denote by $T_{h w}$ the corresponding closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$. Then, the solvability of the $H_{\infty}$ almost disturbance decoupling problem for general discrete-time systems can be defined as follows.

Definition 10.1.1. The general $H_{\infty}$ almost disturbance decoupling problem with measurement feedback and with internal stability ( $H_{\infty}$-ADDPMS) for (10.1.1) is said to be solvable if, for any given positive scalar $\gamma>0$, there exists at least one controller of the form (10.1.2) such that,

1. in the absence of disturbance, the closed-loop system comprising the system (10.1.1) and the controller (10.1.2) is asymptotically stable, i.e., the matrix $A_{\mathrm{cl}}$ as given by (10.1.3) is asymptotically stable;
2. the closed-loop system has an $\mathcal{L}_{2}$-gain, from the disturbance $w$ to the controlled output $h$, that is less than or equal to $\gamma$, i.e.,

$$
\begin{equation*}
\|h\|_{2} \leq \gamma\|w\|_{2}, \quad \forall w \in \mathcal{L}_{2} \text { and for }\left(x(0), x_{c}(0)\right)=(0,0) \tag{10.1.4}
\end{equation*}
$$

Equivalently, the $H_{\infty}$-norm of the closed-loop transfer matrix from $w$ to $h, T_{h w}$, is less than or equal to $\gamma$, i.e., $\left\|T_{h w}\right\|_{\infty} \leq \gamma$.
The problem of $H_{\infty}$ almost disturbance decoupling with state feedback or with full information feedback can be defined in a similar and obvious way. The goal of this chapter is to identify the solvability conditions for these problems and to construct their solutions, if they are existent. The rest of this chapter is organized as follows: In Section 10.2, we give solvability conditions under which the proposed $H_{\infty}$-ADDPMS for general discrete-time systems is solvable. Sections 10.3 and 10.4 give constructive algorithms that would yield solutions to the general discrete-time $H_{\infty}$-ADDPMS, if such solutions exist. All proofs of the main results of this chapter are given in Section 10.5 for the sake of clarity of presentation.

### 10.2. Solvability Conditions

We give in this section the solvability conditions for the general $H_{\infty}$ almost disturbance decoupling problems with internal stability for the following three cases: the full information feedback, the full state feedback and the measurement feedback. These conditions are characterized in terms of some well defined geometric subspaces. We also develop a numerical algorithm that will check these conditions without actually computing any geometric subspaces. The proofs of the main results of this section are given in Section 10.5.

Let us first examine the full information case. We have the following result.
Theorem 10.2.1. Consider the given discrete-time linear time-invariant system $\Sigma$ of (10.1.1) with the measurement output being

$$
\begin{equation*}
y=\binom{x}{w}, \quad \text { or } \quad C_{1}=\binom{I}{0}, \quad D_{1}=\binom{0}{I} \tag{10.2.1}
\end{equation*}
$$

i.e., all the state variables and the disturbances (full information) are measurable and available for feedback. The $H_{\infty}$ almost disturbance decoupling problem with full information feedback and with internal stability for the given system is solvable if and only if the following conditions are satisfied:
(a) $(A, B)$ is stabilizable;
(b) $\operatorname{Im}\left(D_{22}\right) \subset \operatorname{Im}\left(D_{2}\right)$, i.e., $D_{22}+D_{2} S=0$, where $S=-\left(D_{2}^{\prime} D_{2}\right)^{\dagger} D_{2}^{\prime} D_{22}$;
(c) $\operatorname{Im}(E+B S) \subset\left\{\mathcal{V}^{\odot}\left(\Sigma_{\mathrm{P}}\right)+B \operatorname{Ker}\left(D_{2}\right)\right\} \cap\left\{\bigcap_{|\lambda|=1} \mathcal{S}_{\lambda}\left(\Sigma_{\mathrm{P}}\right)\right\}$.

Proof. See Subsection 10.5.A.
因

The result for the general measurement feedback case is given in the next.
Theorem 10.2.2. Consider the given discrete-time linear time-invariant system $\Sigma$ of (10.1.1). The $H_{\infty}$ almost disturbance decoupling problem with measurement feedback and with internal stability ( $H_{\infty}$-ADDPMS) for (10.1.1) is solvable if and only if the following conditions are satisfied:
(a) $(A, B)$ is stabilizable;
(b) $\left(A, C_{1}\right)$ is detectable;
(c) $D_{22}+D_{2} S D_{1}=0$, where $S=-\left(D_{2}^{\prime} D_{2}\right)^{\dagger} D_{2}^{\prime} D_{22} D_{1}^{\prime}\left(D_{1} D_{1}^{\prime}\right)^{\dagger}$;
(d) $\operatorname{Im}\left(E+B S D_{1}\right) \subset\left\{\mathcal{V}^{\odot}\left(\Sigma_{\mathrm{P}}\right)+B \operatorname{Ker}\left(D_{2}\right)\right\} \cap\left\{\bigcap_{|\lambda|=1} \mathcal{S}_{\lambda}\left(\Sigma_{\mathrm{P}}\right)\right\}$;
(e) $\operatorname{Ker}\left(C_{2}+D_{2} S C_{1}\right) \supset\left\{\mathcal{S}^{\odot}\left(\Sigma_{Q}\right) \cap C_{1}^{-1}\left\{\operatorname{Im}\left(D_{1}\right)\right\}\right\} \cup\left\{\bigcup_{|\lambda|=1} \mathcal{V}_{\lambda}\left(\Sigma_{Q}\right)\right\}$;
(f) $\mathcal{S}^{\ominus}\left(\Sigma_{Q}\right) \subset \mathcal{V}^{\ominus}\left(\Sigma_{P}\right)$.

Proof. See Subsection 10.5.B.
The following remarks are in order.
Remark 10.2.1. Note that if $\Sigma_{P}$ is of minimum phase and right invertible with no infinite zeros, and $\Sigma_{Q}$ is of minimum phase and left invertible with no infinite zero, then Conditions (d) to (f) of Theorem 10.2.2 are automatically satisfied. Hence, the solvability conditions of the $H_{\infty}$-ADDPMS for such a case reduce to:
(a) $(A, B)$ is stabilizable;
(b) $\left(A, C_{1}\right)$ is detectable; and
(c) $D_{22}+D_{2} S D_{1}=0$, where $S=-\left(D_{2}^{\prime} D_{2}\right)^{\dagger} D_{2}^{\prime} D_{22} D_{1}^{\prime}\left(D_{1} D_{1}^{\prime}\right)^{\dagger}$

Remark 10.2.2. For special case when all the states of the system (10.1.1) are measurable and available for feedback, i.e., $y=x$, it can be easily derived from Theorem 10.2.2 that the $H_{\infty}$ almost disturbance decoupling problem with full state feedback and with internal stability for such a system is solvable if and only if the following conditions are satisfied:
(a) $(A, B)$ is stabilizable;
(b) $D_{22}=0$; and
(c) $\operatorname{Im}(E) \subset \mathcal{V}^{\circ}\left(\Sigma_{\mathrm{P}}\right) \cap\left\{\bigcap_{|\lambda|=1} \mathcal{S}_{\lambda}\left(\Sigma_{\mathrm{P}}\right)\right\}$.

Next, we proceed to develop a numerical algorithm which verifies the solvability conditions of Theorem 10.2.2 without computing any geometric subspaces of $\Sigma_{\mathrm{P}}$ or $\Sigma_{\mathrm{Q}}$.

Step 10.2.0: Let $S=-\left(D_{2}^{\prime} D_{2}\right)^{\dagger} D_{2}^{\prime} D_{22} D_{1}^{\prime}\left(D_{1} D_{1}^{\prime}\right)^{\dagger}$. If $D_{22}+D_{2} S D_{1} \neq 0$, the $H_{\infty}$-ADDPMS for (10.1.1) is not solvable and the algorithm stops here. Otherwise, go to the next step.

Step 10.2.1: Compute the special coordinate basis of $\Sigma_{\mathrm{P}}$, i.e., the quadruple ( $A, B, C_{2}, D_{2}$ ). For easy reference, we append a subscript ' p ' to all submatrices and transformations in the SCB associated with $\Sigma_{\mathrm{P}}$, e.g., $\Gamma_{\text {sP }}$ is the state transformation of the SCB of $\Sigma_{\mathrm{P}}, B_{d \mathrm{P}}$ is replacing the submatrix $B_{d}$, and $A_{\text {aap }}^{0}$ is associated with invariant zero dynamics of $\Sigma_{\mathrm{P}}$ on the unit circle.

Step 10.2.2: Next, we denote the set of eigenvalues of $A_{a a \Gamma}^{0}$ with a non-negative imaginary part as $\left\{\omega_{\mathrm{P} 1}, \omega_{\mathrm{P} 2}, \cdots, \omega_{\mathrm{P} k_{\mathrm{P}}}\right\}$ and for $i=1,2, \cdots, k_{\mathrm{P}}$, choose complex matrices $V_{i \mathrm{P}}$, whose columns form a basis for the eigenspace $\left\{x \in \mathbb{C}^{n_{a \mathrm{P}}^{0}} \mid x^{\mathrm{H}}\left(\omega_{\mathrm{P} i} I-A_{a a \mathrm{P}}^{0}\right)=0\right\}$, where $n_{a \mathrm{P}}^{0}$ is the dimension of $\mathcal{X}_{a \mathrm{P}}^{0}$. Then, let

$$
V_{\mathrm{P}}:=\left[\begin{array}{llll}
V_{1 \mathrm{P}} & V_{2 \mathrm{P}} & \cdots & V_{k_{\mathrm{P}}} \tag{10.2.2}
\end{array}\right] .
$$

We also compute $n_{x \mathrm{P}}:=\operatorname{dim}\left(\mathcal{X}_{a \mathrm{P}}^{+}\right)+\operatorname{dim}\left(\mathcal{X}_{b \mathrm{P}}\right)+\operatorname{dim}\left(\mathcal{X}_{d \mathrm{P}}\right)$, and

$$
\Gamma_{s \mathrm{P}}^{-1}\left(E+B S D_{1}\right):=\left[\begin{array}{c}
E_{c \mathrm{P}}  \tag{10.2.3}\\
E_{a \mathrm{P}}^{-} \\
E_{a \mathrm{P}}^{0} \\
E_{a \mathrm{P}}^{+} \\
E_{b \mathrm{P}} \\
E_{d \mathrm{P}}
\end{array}\right]
$$

Step 10.2.3: Let $\Sigma_{\mathrm{Q}}^{\star}$ be the dual system of $\Sigma_{\mathbf{Q}}$ and be characterized by a quadruple $\left(A^{\prime}, C_{1}^{\prime}, E^{\prime}, D_{1}^{\prime}\right)$. We compute the special coordinate basis of $\Sigma_{Q}^{\star}$. Again, for easy reference, we append a subscript ' $Q$ ' to all submatrices and transformations in the SCB associated with $\Sigma_{\mathrm{Q}}^{\star}$, e.g., $\Gamma_{s Q}$ is the state transformation of the SCB of $\Sigma_{\mathrm{Q}}^{\star}, B_{d Q}$ is replacing the submatrix $B_{d}$, and $A_{a a \mathrm{Q}}^{0}$ is associated with invariant zero dynamics of $\Sigma_{\mathrm{Q}}^{\star}$ on the unit circle.

Step 10.2.4: Similarly, we denote the set of eigenvalues of $A_{a a Q}^{0}$ with a nonnegative imaginary part as $\left\{\omega_{\mathrm{Q} 1}, \omega_{\mathrm{Q} 2}, \cdots, \omega_{\mathrm{Q}} k_{\mathrm{Q}}\right\}$ and for $i=1,2, \cdots, k_{\mathrm{Q}}$, choose complex matrices $V_{i Q}$, whose columns form a basis for the eigenspace $\left\{x \in \mathbb{C}^{n_{a Q}^{0}} \mid x^{\mathrm{H}}\left(\omega_{\mathrm{Q} i} I-A_{a a Q}^{0}\right)=0\right\}$, where $n_{a \mathrm{Q}}^{0}$ is the dimension of $\mathcal{X}_{a Q}^{0}$. Then, let

$$
V_{\mathrm{Q}}:=\left[\begin{array}{llll}
V_{1 \mathrm{Q}} & V_{2 \mathrm{Q}} & \cdots & V_{k_{\mathrm{QQ}}} \tag{10.2.4}
\end{array}\right]
$$

We next compute $n_{x Q}:=\operatorname{dim}\left(\mathcal{X}_{a Q}^{+}\right)+\operatorname{dim}\left(\mathcal{X}_{b Q}\right)+\operatorname{dim}\left(\mathcal{X}_{d Q}\right)$, and

$$
\Gamma_{s Q}^{-1}\left(C_{2}+D_{2} S C_{1}\right)^{\prime}:=\left[\begin{array}{c}
E_{c Q}  \tag{10.2.5}\\
E_{a Q}^{-} \\
E_{a Q}^{0} \\
E_{a Q}^{+} \\
E_{b Q} \\
E_{d Q}
\end{array}\right]
$$

Step 10.2.5: Finally, compute

$$
\Gamma_{s \mathrm{P}}^{-1}\left(\Gamma_{s \mathrm{Q}}^{-1}\right)^{\prime}=\left[\begin{array}{ll}
\star & \star  \tag{10.2.6}\\
\star & \Gamma
\end{array}\right]
$$

where $\Gamma$ is a $n_{x \mathrm{P}} \times n_{x Q}$ constant matrix.
®

The following proposition summaries the result of the above algorithm. It also gives a set of necessary and sufficient conditions, in terms of sub-matrices associated with the SCB's of $\Sigma_{\mathbf{P}}$ and $\Sigma_{\mathbf{Q}}$, for the solvability of the $H_{\infty}$-ADDPMS for the general discrete-time system $\Sigma$ of (10.1.1).

Proposition 10.2.1. Consider the given discrete-time linear time-invariant system $\Sigma$ of (10.1.1). The $H_{\infty}$ almost disturbance decoupling problem with measurement feedback and with internal stability ( $H_{\infty}$-ADDPMS) for (10.1.1) is solvable if and only if the following conditions are satisfied:
(a) $(A, B)$ is stabilizable;
(b) $\left(A, C_{1}\right)$ is detectable;
(c) $D_{22}-D_{2}\left(D_{2}^{\prime} D_{2}\right)^{\dagger} D_{2}^{\prime} D_{22} D_{1}^{\prime}\left(D_{1} D_{1}^{\prime}\right)^{\dagger} D_{1}=0$;
(d) $V_{\mathrm{P}}^{\mathrm{H}} E_{a \mathrm{P}}^{0}=0, E_{a \mathrm{P}}^{+}=0, E_{b \mathrm{P}}=0, \operatorname{Im}\left(E_{d \mathrm{P}}\right) \subset \operatorname{Im}\left(B_{d \mathrm{P}}\right)$;
(e) $V_{\mathrm{Q}}^{\mathrm{H}} E_{a \mathrm{Q}}^{0}=0, E_{a \mathrm{Q}}^{+}=0, E_{b \mathrm{Q}}=0, \operatorname{Im}\left(E_{d Q}\right) \subset \operatorname{Im}\left(B_{d \mathrm{Q}}\right)$; and
(f) $\Gamma=0$.

Note that all the matrices in (d)-(f) are well-defined in Steps 10.2.0 to 10.2.5 of the algorithm.

The above result can be directly verified using the properties of the special coordinate basis of Chapter 2 and the result of Theorem 10.2.2 (see also Chapter 7 for a similar result for continuous-time systems).

### 10.3. Solutions to State and Full Information Feedback Cases

In this section, we consider feedback control law design for the general $H_{\infty}$ almost disturbance decoupling problem with internal stability as well as with both full state feedback and full information feedback, where internal stability is with respect to the open unit disc. More specifically, we will first present a
design procedure that constructs a family of parameterized static state feedback control laws,

$$
\begin{equation*}
u(k)=F(\varepsilon) x(k) \tag{10.3.1}
\end{equation*}
$$

which solves the general $H_{\infty}$-ADDPMS for the following system,

$$
\left\{\begin{align*}
x(k+1) & =A x(k)+B u(k)+E w(k)  \tag{10.3.2}\\
y(k) & =x(k) \\
h(k) & =C_{2} x(k)+D_{2} u(k)+D_{22} w(k)
\end{align*}\right.
$$

That is, under this family of state feedback control laws, the resulting closedloop system is asymptotically stable for sufficiently small $\varepsilon$ and the $H_{\infty}$-norm of the closed-loop transfer matrix from $w$ to $h, T_{h w}(z, \varepsilon)$, tends to zero as $\varepsilon$ tends to zero, where

$$
\begin{equation*}
T_{h w}(z, \varepsilon)=\left[C_{2}+D_{2} F(\varepsilon)\right][z I-A-B F(\varepsilon)]^{-1} E+D_{22} \tag{10.3.3}
\end{equation*}
$$

We have the following algorithm for constructing such an $F(\varepsilon)$.
Step 10.S.1: (Decomposition of $\Sigma_{P}$ ). Transform the subsystem $\Sigma_{P}$, i.e., the matrix quadruple ( $A, B, C_{2}, D_{2}$ ), into the special coordinate basis (SCB) as given by Theorem 2.3.1. Denote the state, output and input transformation matrices as $\Gamma_{s \mathrm{P}}, \Gamma_{o \mathrm{P}}$ and $\Gamma_{i \mathrm{P}}$, respectively.

Step 10.S.2: (Gain matrix for the subsystem associated with $\mathcal{X}_{c}$ ). Let $F_{c}$ be any constant matrix subject to the constraint that

$$
\begin{equation*}
A_{c c}^{c}=A_{c c}-B_{c} F_{c} \tag{10.3.4}
\end{equation*}
$$

is a stable matrix. Note that the existence of such an $F_{c}$ is guaranteed by the property of the special coordinate basis, i.e., $\left(A_{c c}, B_{c}\right)$ is controllable.

Step 10.S.3: (Gain matrix for the subsystem associated with $\mathcal{X}_{a}^{+}, \mathcal{X}_{b}$ and $\mathcal{X}_{d}$ ). Let

$$
F_{a b d}:=\left[\begin{array}{ccccc}
0 & 0 & F_{a 0}^{+} & F_{b 0} & F_{d 0}  \tag{10.3.5}\\
E_{d a}^{-} & E_{d a}^{0} & F_{a d}^{+} & F_{b d} & F_{d d}
\end{array}\right]
$$

where

$$
F_{a b d}^{+}:=\left[\begin{array}{lll}
F_{a 0}^{+} & F_{b 0} & F_{d 0}  \tag{10.3.6}\\
F_{a d}^{+} & F_{b d} & F_{d d}
\end{array}\right]
$$

is any constant matrix subject to the constraint that

$$
A_{a b d}^{+c}:=\left[\begin{array}{ccc}
A_{a a}^{+} & L_{a b}^{+} C_{b} & L_{a d}^{+} C_{d}  \tag{10.3.7}\\
0 & A_{b b} & L_{b d} C_{d} \\
B_{d} E_{d a}^{+} & B_{d} E_{d b} & A_{d d}
\end{array}\right]-\left[\begin{array}{cc}
B_{0 a}^{+} & 0 \\
B_{0 b} & 0 \\
B_{0 d} & B_{d}
\end{array}\right] F_{a b d}^{+}
$$

is an asymptotically stable matrix. Again, the existence of such an $F_{a b d}^{+}$ is guaranteed by the property of the special coordinate basis.

Step 10.S.4: (Gain matrix for the subsystem associated with $A_{a a}^{0}$ ). The construction of this gain matrix is carried out in the following sub-steps.

Step 10.S.4.1: (Preliminary coordinate transformation). Noting that

$$
A_{\mathrm{con}}:=\left[\begin{array}{ccc}
A_{a a} & L_{a b} C_{b} & L_{a d} C_{d} \\
0 & A_{b b} & L_{b d} C_{d} \\
B_{d} E_{d a} & B_{d} E_{d b} & A_{d d}
\end{array}\right], B_{\mathrm{con}}:=\left[\begin{array}{cc}
B_{0 a} & 0 \\
B_{0 b} & 0 \\
B_{0 d} & B_{d}
\end{array}\right]
$$

we have

$$
A_{\mathrm{con}}-B_{\mathrm{con}} F_{a b d}=\left[\begin{array}{ccc}
A_{a a}^{-} & 0 & A_{a b d}^{-}  \tag{10.3.8}\\
0 & A_{a a}^{0} & A_{a b d}^{0} \\
0 & 0 & A_{a b d}^{+c}
\end{array}\right], \quad B_{\mathrm{con}}=\left[\begin{array}{cc}
B_{0 a}^{-} & 0 \\
B_{0 a}^{0} & 0 \\
B_{0 a b d}^{+} & \tilde{B}_{d}
\end{array}\right]
$$

where

$$
\begin{gather*}
B_{0 a b d}^{+}=\left[\begin{array}{c}
B_{0 a}^{+} \\
B_{0 b} \\
B_{0 d}
\end{array}\right], \quad \tilde{B}_{d}=\left[\begin{array}{c}
0 \\
0 \\
B_{d}
\end{array}\right],  \tag{10.3.9}\\
A_{a b d}^{0}=\left[\begin{array}{lll}
0 & L_{a b}^{0} C_{b} & L_{a d}^{0} C_{d}
\end{array}\right]-\left[\begin{array}{ll}
B_{0 a}^{0} & 0
\end{array}\right] F_{a b d}^{+}, \tag{10.3.10}
\end{gather*}
$$

and

$$
A_{a b d}^{-}=\left[\begin{array}{lll}
0 & L_{a b}^{-} C_{b} & L_{a d}^{-} C_{d}
\end{array}\right]-\left[\begin{array}{ll}
B_{0 a}^{-} & 0 \tag{10.3.11}
\end{array}\right] F_{a b d}^{+}
$$

Clearly, the pair ( $A_{\text {con }}-B_{\text {con }} F_{a b d}, B_{\text {con }}$ ) remains stabilizable. Construct the following nonsingular transformation matrix,

$$
\Gamma_{a b d}=\left[\begin{array}{ccc}
I_{n_{\bar{a}}} & 0 & 0  \tag{10.3.12}\\
0 & 0 & I_{n_{a}^{+}+n_{b}+n_{d}} \\
0 & I_{n_{a}^{0}} & T_{a}^{0}
\end{array}\right]^{-1}
$$

where $T_{a}^{0}$ is the unique solution to the following Lyapunov equation,

$$
\begin{equation*}
A_{a a}^{0} T_{a}^{0}-T_{a}^{0} A_{a b d}^{+c}=A_{a b d}^{0} \tag{10.3.13}
\end{equation*}
$$

We note here that such a unique solution to the above Lyapunov equation always exists since all the eigenvalues of $A_{a a}^{0}$ are on the unit circle and all the eigenvalues of $A_{a b d}^{+c}$ are on the open unit disc. It is now easy to verify that

$$
\Gamma_{a b d}^{-1}\left(A_{\mathrm{con}}-B_{\mathrm{con}} F_{a b d}\right) \Gamma_{a b d}=\left[\begin{array}{ccc}
A_{a a}^{-} & A_{a a b}^{-} & 0  \tag{10.3.14}\\
0 & A_{a b d}^{+c} & 0 \\
0 & 0 & A_{a a}^{0}
\end{array}\right]
$$

and

$$
\Gamma_{a b d}^{-1} B_{\mathrm{con}}=\left[\begin{array}{cc}
B_{0 a}^{-} & 0  \tag{10.3.15}\\
B_{0 a b d}^{+} & \tilde{B}_{d} \\
B_{0 a}^{0}+T_{a}^{0} B_{0 a b d}^{+} & T_{a}^{0} \tilde{B}_{d}
\end{array}\right]
$$

Hence, the matrix pair $\left(A_{a a}^{0}, B_{a}^{0}\right)$ is controllable, where

$$
\begin{equation*}
B_{a}^{0}=\left[B_{0 a}^{0}+T_{a}^{0} B_{0 a b d}^{+} \quad T_{a}^{0} \tilde{B}_{d}\right] \tag{10.3.16}
\end{equation*}
$$

Step 10.S.4.2: (Further coordinate transformation). Use the results of Chapter 2 to find nonsingular transformation matrices $\Gamma_{s a}^{0}$ and $\Gamma_{i a}^{0}$ such that $\left(A_{a a}^{0}, B_{a}^{0}\right)$ can be transformed into the block diagonal controllability canonical form,

$$
\left(\Gamma_{s a}^{0}\right)^{-1} A_{a a}^{0} \Gamma_{s a}^{0}=\left[\begin{array}{cccc}
A_{1} & 0 & \cdots & 0  \tag{10.3.17}\\
0 & A_{2} & \cdots & 0 \\
\vdots & \vdots & \ddots & \vdots \\
0 & 0 & \cdots & A_{l}
\end{array}\right]
$$

and

$$
\left(\Gamma_{s a}^{0}\right)^{-1} B_{a}^{0} \Gamma_{i a}^{0}=\left[\begin{array}{ccccc}
B_{1} & B_{12} & \cdots & B_{1 l} & \star  \tag{10.3.18}\\
0 & B_{2} & \cdots & B_{2 l} & \star \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & B_{l} & \star
\end{array}\right]
$$

where $l$ is an integer and for $i=1,2, \cdots, l$,

$$
A_{i}=\left[\begin{array}{ccccc}
0 & 1 & 0 & \cdots & 0 \\
0 & 0 & 1 & \cdots & 0 \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
0 & 0 & 0 & \cdots & 1 \\
-a_{n_{i}}^{i} & -a_{n_{i}-1}^{i} & -a_{n_{i}-2}^{i} & \cdots & -a_{1}^{i}
\end{array}\right], \quad B_{i}=\left[\begin{array}{c}
0 \\
0 \\
\vdots \\
0 \\
1
\end{array}\right] .
$$

We note that all the eigenvalues of $A_{i}$ are on the unit circle. Here, the $\star$ 's represent sub-matrices of less interest.

Step 10.S.4.3: (Subsystem design). For each $\left(A_{i}, B_{i}\right)$, let $F_{i}(\varepsilon) \in \mathbb{R}^{1 \times n_{i}}$ be the state feedback gain such that

$$
\lambda\left\{A_{i}+B_{i} F_{i}(\varepsilon)\right\}=\left\{(1-\varepsilon) e^{j \theta_{i 1}}, \cdots,(1-\varepsilon) e^{j \theta_{i n_{i}}}\right\}
$$

where $e^{j \theta_{i \ell}}, \ell=1,2, \cdots, n_{i}$, are the eigenvalues of $A_{i}$. Clearly, all the eigenvalues of $A_{i}+B_{i} F_{i}(\varepsilon)$ are on the open unit disc and $F_{i}(\varepsilon)$ is unique.

Step 10.S.4.4: (Composition of gain matrix for subsystem associated with $\mathcal{X}_{a}^{0}$ ). Let

$$
F_{a}^{0}(\varepsilon):=\Gamma_{i a}^{0}\left[\begin{array}{ccccc}
F_{1}(\varepsilon) & 0 & \cdots & 0 & 0  \tag{10.3.19}\\
0 & F_{2}(\varepsilon) & \cdots & 0 & 0 \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
0 & 0 & \cdots & F_{l-1}(\varepsilon) & 0 \\
0 & 0 & \cdots & 0 & F_{l}(\varepsilon) \\
0 & 0 & \cdots & 0 & 0
\end{array}\right]\left(\Gamma_{s a}^{0}\right)^{-1},
$$

where $\varepsilon \in(0,1]$ is a design parameter whose value is to be specified later. For future use, we partition

$$
F_{a}^{0}(\varepsilon)=\left[\begin{array}{l}
F_{a 0}^{0}(\varepsilon)  \tag{10.3.20}\\
F_{a d}^{0}(\varepsilon)
\end{array}\right],
$$

and

$$
F_{a}^{0}(\varepsilon) T_{a}^{0}=\left[\begin{array}{lll}
F_{a 0+}^{0}(\varepsilon) & F_{a 0 b}^{0}(\varepsilon) & F_{a 0 d}^{0}(\varepsilon)  \tag{10.3.21}\\
F_{a d+}^{0}(\varepsilon) & F_{a d b}^{0}(\varepsilon) & F_{a d d}^{0}(\varepsilon)
\end{array}\right] .
$$

Step 10.S.5: (Composition of parameterized gain matrix $F(\varepsilon)$ ). In this step, various gains calculated in Steps 10.S. 3 to 10.S. 5 are put together to form a composite state feedback gain matrix $F(\varepsilon)$. It is given by

$$
\begin{equation*}
F(\varepsilon):=-\Gamma_{i \mathrm{P}}\left[F_{0}+F_{\star}(\varepsilon)\right] \Gamma_{s \mathrm{P}}^{-1}, \tag{10.3.22}
\end{equation*}
$$

where

$$
F_{0}=\left[\begin{array}{cccccc}
C_{0 a}^{-} & C_{0 a}^{0} & C_{0 a}^{+}+F_{a 0}^{+} & C_{0 b}+F_{b 0} & C_{0 c} & C_{0 d}+F_{d 0}  \tag{10.3.23}\\
E_{d a}^{-} & E_{d a}^{0} & F_{a d}^{+} & F_{b d} & E_{d c} & F_{d d} \\
E_{c a}^{-} & E_{c a}^{0} & E_{c a}^{+} & 0 & F_{c} & 0
\end{array}\right],
$$

and

$$
F_{\star}(\varepsilon)=\left[\begin{array}{cccccc}
0 & F_{a 0}^{0}(\varepsilon) & F_{a 0+}^{0}(\varepsilon) & F_{a 0 b}^{0}(\varepsilon) & 0 & F_{a 0 d}^{0}(\varepsilon)  \tag{10.3.24}\\
0 & F_{a d}^{0}(\varepsilon) & F_{a d+}^{0}(\varepsilon) & F_{a d b}^{0}(\varepsilon) & 0 & F_{a d d}^{0}(\varepsilon) \\
0 & 0 & 0 & 0 & 0 & 0
\end{array}\right] .
$$

This completes the construction of the parameterized state feedback gain matrix $F(\varepsilon)$.

We have the following theorem.
Theorem 10.3.1. Consider the given system (10.3.2) in which all the states are available for feedback. Assume that the problem of $H_{\infty}$ almost disturbance
decoupling with internal stability for (10.3.2) is solvable, i.e., the solvability conditions of Remark 10.2.2 are satisfied. Then, the closed-loop system comprising (10.3.2) and the full state feedback control law,

$$
\begin{equation*}
u(k)=F(\varepsilon) x(k) \tag{10.3.25}
\end{equation*}
$$

with $F(\varepsilon)$ given by (10.3.22), has the following properties: For any given $\gamma>0$, there exists a positive scalar $\varepsilon^{*}>0$ such that for all $0<\varepsilon \leq \varepsilon^{*}$,

1. the closed-loop system is asymptotically stable, i.e., $\lambda\{A+B F(\varepsilon)\}$ are on the open unit disc; and
2. the $H_{\infty}$-norm of the closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$ is less than $\gamma$, i.e., $\left\|T_{h w}(z, \varepsilon)\right\|_{\infty}<\gamma$.

Hence, by Definition 10.1.1, the control law of (10.3.25) solves the $H_{\infty}$-ADDPMS for (10.3.2).

Proof. The proof of this theorem is somewhat similar to that of its continuoustime counterpart, i.e., Theorem 7.3.1. We refer interested readers to [65] for further details.

Next, we proceed to design a parameterized control law,

$$
\begin{equation*}
u(k)=F_{x}(\varepsilon) x(k)+F_{w} w(k) \tag{10.3.26}
\end{equation*}
$$

which solves the $H_{\infty}$ almost disturbance decoupling problem with internal stability for the following full information system,

$$
\left\{\begin{align*}
& x(k+1)=A x(k)+B u(k)+E w(k)  \tag{10.3.27}\\
& y(k)=\binom{I}{0} x(k) \\
& h(k)\left.=C_{2} x(k)+D_{2} u(k)+\begin{array}{c}
0 \\
I
\end{array}\right) w(k) \\
& D_{22} w(k)
\end{align*}\right.
$$

That is, under the above full information feedback control law, the resulting closed-loop system is asymptotically stable for sufficiently small $\varepsilon$ and the $H_{\infty^{-}}$ norm of the closed-loop transfer matrix from $w$ to $h, T_{h w}(z, \varepsilon)$, tends to zero as $\varepsilon$ tends to zero, where

$$
T_{h w}(z, \varepsilon)=\left[C_{2}+D_{2} F_{x}(\varepsilon)\right]\left[z I-A-B F_{x}(\varepsilon)\right]^{-1}\left(E+B F_{w}\right)+\left(D_{22}+D_{2} F_{w}\right)
$$

The following is a step-by-step algorithm for constructing $F_{x}(\varepsilon)$ and $F_{w}$.

Step 10.F.1: (Computation of $S$ ). Compute

$$
\begin{equation*}
S=-\left(D_{2}^{\prime} D_{2}\right)^{\dagger} D_{2}^{\prime} D_{22} \tag{10.3.28}
\end{equation*}
$$

Step 10.F.2: (Computation of $F_{x}(\varepsilon)$ ). Follow Steps 10.S. 1 to $10 . S .5$ of the previous algorithm to yield a gain matrix $F(\varepsilon)$. Then, let

$$
\begin{equation*}
F_{x}(\varepsilon)=F(\varepsilon) \tag{10.3.29}
\end{equation*}
$$

Also, we need to retain the transformation matrices $\Gamma_{s P}$ and $\Gamma_{i \mathrm{P}}$, as well as the sub-matrix $B_{d}$ of the SCB of $\Sigma_{\mathrm{P}}$ in order to compute $F_{w}$ in the next step.

Step 10.F.3: (Construction of gain matrix $F_{w}$ ). Let

$$
\Gamma_{s \mathrm{P}}^{-1}(E+B S)=\left[\begin{array}{c}
E_{a}^{-}  \tag{10.3.30}\\
E_{a}^{0} \\
E_{a}^{+} \\
E_{b} \\
E_{c} \\
E_{d}
\end{array}\right]
$$

Then, the gain matrix $F_{w}$ is given by

$$
F_{w}=-\Gamma_{i \mathrm{P}}\left[\begin{array}{c}
0  \tag{10.3.31}\\
\left(B_{d}^{\prime} B_{d}\right)^{-1} B_{d}^{\prime} E_{d} \\
0
\end{array}\right]+S
$$

It is interesting to note that the first portion of matrix $F_{w}$ is used to clean up the disturbance associated with $E_{d}$ and in the range space of $B_{d}$, while the second portion is used to reject disturbance entering into the system through $D_{22}$.

We have the following result.
Theorem 10.3.2. Consider the given system (10.3.27) in which all the states and the disturbances are available for feedback. Assume that the problem of $H_{\infty}$ almost disturbance decoupling with internal stability for (10.3.27) is solvable, i.e., the solvability conditions of Theorem 10.2 .1 are satisfied. Then, the closed-loop system comprising (10.3.27) and the full information feedback control law,

$$
\begin{equation*}
u(k)=F_{x}(\varepsilon) x(k)+F_{w} w(k) \tag{10.3.32}
\end{equation*}
$$

with $F_{x}(\varepsilon)$ and $F_{w}$ being given by (10.3.29) and (10.3.31), respectively, has the following properties: For any given $\gamma>0$, there exists a positive scalar $\varepsilon^{*}>0$ such that for all $0<\varepsilon \leq \varepsilon^{*}$,

1. the closed-loop system is asymptotically stable, i.e., $\lambda\left\{A+B F_{x}(\varepsilon)\right\}$ are on the open unit disc; and
2. the $H_{\infty}$-norm of the closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$ is less than $\gamma$, i.e., $\left\|T_{h w}(z, \varepsilon)\right\|_{\infty}<\gamma$.

Hence, by Definition 10.1.1, the control law of (10.3.32) solves the $H_{\infty}$-ADDPMS for (10.3.27).

Proof. See Subsection 10.5.C.
因

We illustrate the results of this section with the following example.

Example 10.3.1. Consider a discrete-time system characterized by (10.1.1) with

$$
A=\left[\begin{array}{rrrrr}
1 & 1 & 1 & 1 & 0  \tag{10.3.33}\\
0 & 1 & 1 & 1 & 0 \\
0 & 0 & 0.1 & 1 & 0 \\
0 & 0 & 0 & 0 & 1 \\
0.1 & 0.1 & 0.1 & 0.1 & 0.1
\end{array}\right], \quad B=\left[\begin{array}{l}
0 \\
0 \\
0 \\
0 \\
1
\end{array}\right], \quad E=\left[\begin{array}{rr}
0 & 1 \\
0 & 0 \\
0 & 0 \\
0 & 0 \\
\alpha_{e} & 0
\end{array}\right]
$$

where $\alpha_{e}$ is a scalar, and

$$
C_{2}=\left[\begin{array}{lllll}
0 & 0 & 0 & 1 & 0  \tag{10.3.34}\\
0 & 0 & 1 & 0 & 0
\end{array}\right], \quad D_{2}=\left[\begin{array}{l}
0 \\
0
\end{array}\right], \quad D_{22}=\left[\begin{array}{ll}
0 & 0 \\
0 & 0
\end{array}\right] .
$$

We will consider both the state feedback case and the full information feedback case in this example. Using the toolbox of Chen [12], we can verify that $(A, B)$ is controllable and $\Sigma_{\mathrm{P}}$, i.e., $\left(A, B, C_{2}, D_{2}\right)$, is left invertible with two invariant zeros at $z=1$ and one infinite zero of order 2 . Moreover,

$$
\mathcal{V}^{\odot}\left(\Sigma_{\mathrm{P}}\right)=\operatorname{Im}\left\{\left[\begin{array}{ll}
1 & 0  \tag{10.3.35}\\
0 & 1 \\
0 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right]\right\}, \quad B \operatorname{Ker}\left(D_{2}\right)=\operatorname{Im}\left\{\left[\begin{array}{l}
0 \\
0 \\
0 \\
0 \\
1
\end{array}\right]\right\}
$$

and

$$
\bigcap_{|\lambda|=1} \mathcal{S}_{\lambda}\left(\Sigma_{\mathrm{P}}\right)=\operatorname{Im}\left\{\left[\begin{array}{lll}
1 & 0 & 0  \tag{10.3.36}\\
0 & 0 & 0 \\
0 & 0 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{array}\right]\right\} .
$$

Also, we have

$$
\mathcal{V}^{\odot}\left(\Sigma_{\mathrm{P}}\right) \cap\left\{\bigcap_{|\lambda|=1} \mathcal{S}_{\lambda}\left(\Sigma_{\mathrm{P}}\right)\right\}=\operatorname{Im}\left\{\left[\begin{array}{l}
1 \\
0 \\
0 \\
0 \\
0
\end{array}\right]\right\}
$$

and

$$
\left\{\mathcal{V}^{\odot}\left(\Sigma_{\mathrm{P}}\right)+B \operatorname{Ker}\left(D_{2}\right)\right\} \cap\left\{\bigcap_{|\lambda|=1} \mathcal{S}_{\lambda}\left(\Sigma_{\mathrm{P}}\right)\right\}=\operatorname{Im}\left\{\left[\begin{array}{ll}
1 & 0 \\
0 & 0 \\
0 & 0 \\
0 & 0 \\
0 & 1
\end{array}\right]\right\}
$$

It is clear to see now that the $H_{\infty}$ almost disturbance decoupling problem with internal stability ( $H_{\infty}$-ADDPS) using state feedback for the given system is solvable if and only if $\alpha_{e}=0$ and the $H_{\infty}$-ADDPS using full information feedback for the given system is always solvable. Following the algorithms of this section, we obtain the following parameterized gain matrices,

$$
F_{x}(\varepsilon)=\left[\begin{array}{c}
-0.526316(\varepsilon-1)^{2}-1.052632(\varepsilon-1)-0.626316  \tag{10.3.37}\\
-0.775623(\varepsilon-1)^{2}-2.603878(\varepsilon-1)-1.928255 \\
-0.798061(\varepsilon-1)^{2}-2.763490(\varepsilon-1)-2.066429 \\
-(\varepsilon-1)^{2}-4.2(\varepsilon-1)-3.31 \\
-2(\varepsilon-1)-2.2
\end{array}\right]
$$

which places the eigenvalues of $A+B F_{x}(\varepsilon)$ around at $0,0,0,1-\varepsilon$ and $1-\varepsilon$, and

$$
F_{w}=\left[\begin{array}{ll}
-\alpha_{e} & 0 \tag{10.3.38}
\end{array}\right]
$$

The maximum singular value plots of the corresponding closed-loop transfer matrix $T_{h w}(z, \varepsilon)$ in Figure 10.3 .1 clearly show that the $H_{\infty}$-ADDPS using full information feedback (or state feedback when $\alpha_{e}=0$ ) is attained as $\varepsilon$ tends smaller and smaller.

### 10.4. Solutions to Measurement Feedback Case

We present in this section the designs of both full order and reduced order output feedback controllers that solve the general $H_{\infty}$-ADDPMS for the given system (10.1.1). Here, by full order controller, we mean that the order of the controller is exactly the same as the given system (10.1.1), i.e, is equal to $n$. A reduced order controller, on the other hand, refers to a controller whose dynamical order is less than $n$.


Figure 10.3.1: Max. singular values of $T_{h w}$ - Full information case.

### 10.4.1. Full Order Output Feedback

The following is a step-by-step algorithm for constructing a parameterized full order output feedback controller that solves the $H_{\infty}$-ADDPMS for (10.1.1).

Step 10.F.C.1: (Computation of $N$ ). Utilize the properties of the SCB to compute two constant matrices $X$ and $Y$ such that $\mathcal{V}^{\odot}\left(\Sigma_{\mathrm{P}}\right)=\operatorname{Ker}(X)$ and $\mathcal{S}^{\odot}\left(\Sigma_{\mathrm{Q}}\right)=\operatorname{Im}(Y)$. Then, compute

$$
\begin{align*}
& N=-\left(B^{\prime} X^{\prime} X B+D_{2}^{\prime} D_{2}\right)^{\dagger}\left[\begin{array}{ll}
B^{\prime} & X^{\prime} \\
D_{2}^{\prime}
\end{array}\right]\left[\begin{array}{cc}
X A Y & X E \\
C_{2} Y & D_{22}
\end{array}\right] \\
& \times\left[\begin{array}{c}
Y^{\prime} C_{1}^{\prime} \\
D_{1}^{\prime}
\end{array}\right]\left(C_{1} Y Y^{\prime} C_{1}^{\prime}+D_{1} D_{1}^{\prime}\right)^{\dagger} \tag{10.4.1}
\end{align*}
$$

Step 10.F.C.2: (Construction of the gain matrix $F_{\mathrm{P}}(\varepsilon)$ ). Define an auxiliary system

$$
\left\{\begin{align*}
x(k+1) & =\tilde{A} x(k)+B u(k)+\tilde{E} w(k)  \tag{10.4.2}\\
y(k) & =x(k) \\
h(k) & =\tilde{C}_{2} x(k)+D_{2} u(k)+0 w(k)
\end{align*}\right.
$$

where

$$
\begin{equation*}
\tilde{A}:=A+B N C_{1}, \tag{10.4.3}
\end{equation*}
$$

$$
\begin{align*}
\tilde{E} & :=E+B N D_{1},  \tag{10.4.4}\\
\tilde{C}_{2} & :=C_{2}+D_{2} N C_{1} \tag{10.4.5}
\end{align*}
$$

and then perform Steps 10.S. 1 to 10.S.5 of the previous section to the above system (10.4.2) to obtain a parameterized gain matrix $F(\varepsilon)$. We let $F_{\mathrm{P}}(\varepsilon)=F(\varepsilon)$.

Step 10.F.C.3: (Construction of the gain matrix $K_{\mathrm{Q}}(\varepsilon)$ ). Define another auxiliary system

$$
\left\{\begin{align*}
x(k+1) & =\tilde{A}^{\prime} x(k)+C_{1}^{\prime} u(k)+\tilde{C}_{2}^{\prime} w(k),  \tag{10.4.6}\\
y(k) & =x(k) \\
h(k) & =\tilde{E}^{\prime} x(k)+D_{1}^{\prime} u(k)+0 \quad w(k),
\end{align*}\right.
$$

and then perform Steps 10.S. 1 to 10.S. 6 of the previous section to the above system to get the parameterized gain matrix $F(\varepsilon)$. Similarly, we let $K_{\mathrm{Q}}(\varepsilon)=F(\varepsilon)^{\prime}$.

Step 10.F.C.4: (Construction of the full order controller $\Sigma_{\mathrm{FC}}(\varepsilon)$ ). Finally, the parameterized full order output feedback controller is given by

$$
\Sigma_{\mathrm{FC}}(\varepsilon):\left\{\begin{array}{cl}
x_{c}(k+1) & =A_{\mathrm{FC}}(\varepsilon) x_{c}(k)+B_{\mathrm{FC}}(\varepsilon) y(k)  \tag{10.4.7}\\
u(k) & =C_{\mathrm{FC}}(\varepsilon) x_{c}(k)+D_{\mathrm{FC}}(\varepsilon) y(k)
\end{array}\right.
$$

where

$$
\left.\begin{array}{l}
A_{\mathrm{FC}}(\varepsilon):=A+B N C_{1}+B F_{\mathrm{P}}(\varepsilon)+K_{\mathrm{Q}}(\varepsilon) C_{1}  \tag{10.4.8}\\
B_{\mathrm{FC}}(\varepsilon):=-K_{\mathrm{Q}}(\varepsilon) \\
C_{\mathrm{FC}}(\varepsilon):=F_{\mathrm{P}}(\varepsilon) \\
D_{\mathrm{FC}}(\varepsilon):=N
\end{array}\right\}
$$

We have the following theorem.
Theorem 10.4.1. Consider the given system $\Sigma$ of (10.1.1). Assume that the problem of $H_{\infty}$ almost disturbance decoupling with internal stability for (10.1.1) is solvable, i.e., the solvability conditions of Theorem 10.2.2 are satisfied. Then, the closed-loop system comprising (10.1.1) and the full order measurement feedback controller (10.4.7) has the following properties: For any given $\gamma>0$, there exists a positive scalar $\varepsilon^{*}>0$ such that for all $0<\varepsilon \leq \varepsilon^{*}$,

1. the closed-loop system is asymptotically stable; and
2. the $H_{\infty}$-norm of the closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$ is less than $\gamma$, i.e., $\left\|T_{h w}(z, \varepsilon)\right\|_{\infty}<\gamma$.

Hence, by Definition 10.1.1, the control law of (10.4.7) solves the $H_{\infty}$-ADDPMS for (10.1.1).

Proof. See Subsection 10.5.D.因

We illustrate the above result in the following example.
Example 10.4.1. We now consider a discrete-time system characterized by (10.1.1) with $A, B, E, C_{2}, D_{2}$ and $D_{22}$ being given as in Example 10.3.1, and

$$
C_{1}=\left[\begin{array}{rrrrr}
0.5 & 0.1 & 0.5 & 0.2 & 0.1  \tag{10.4.9}\\
1 & 0 & 0 & 0 & 0
\end{array}\right], \quad D_{1}=\left[\begin{array}{ll}
1 & 0 \\
0 & 0
\end{array}\right] .
$$

For simplicity, we let $\alpha_{e}=1$ in matrix $E$. Using the toolbox of Chen [12] again, one can verify that $\left(A, C_{1}\right)$ is observable and $\Sigma_{2}$, i.e., $\left(A, E, C_{1}, D_{1}\right)$, is invertible with one infinite zero of order one and four invariant zeros at -0.6554 , $0.3777 \pm j 0.6726$, and 1 . Moreover,

$$
\mathcal{S}^{\odot}\left(\Sigma_{\mathrm{Q}}\right)=\operatorname{Im}\left\{\left[\begin{array}{l}
1  \tag{10.4.10}\\
0 \\
0 \\
0 \\
0
\end{array}\right]\right\}, \quad C_{1}^{-1}\left\{\operatorname{Im}\left(D_{1}\right)\right\}=\operatorname{Im}\left\{\left[\begin{array}{l}
1 \\
0 \\
0 \\
0 \\
0
\end{array}\right]\right\}
$$

and

$$
\bigcup_{|\lambda|=1} \mathcal{V}_{\lambda}\left(\Sigma_{Q}\right)=\operatorname{Im}\left\{\left[\begin{array}{l}
0  \tag{10.4.11}\\
1 \\
0 \\
0 \\
0
\end{array}\right]\right\}
$$

Hence,

$$
\left\{\mathcal{S}^{\odot}\left(\Sigma_{\mathrm{Q}}\right) \cap C_{1}^{-1}\left\{\operatorname{Im}\left(D_{1}\right)\right\}\right\} \cup\left\{\bigcup_{|\lambda|=1} \mathcal{V}_{\lambda}\left(\Sigma_{\mathrm{Q}}\right)\right\}=\operatorname{Im}\left\{\left[\begin{array}{ll}
1 & 0  \tag{10.4.12}\\
0 & 1 \\
0 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right]\right\}
$$

It is ready to see now that all conditions in Theorem 10.2.2 are satisfied. Hence, the $H_{\infty}$-ADDPMS for the given system is solvable. Following the algorithm of this subsection, we obtain a full order output feedback controller of the form (10.4.7) with

$$
N=\left[\begin{array}{ll}
-1 & 0.4 \tag{10.4.13}
\end{array}\right]
$$



Figure 10.4.1: Max. singular values of $T_{h w}$ - Full order output feedback.

$$
F_{\mathrm{P}}(\varepsilon)=\left[\begin{array}{c}
-0.526316(\varepsilon-1)^{2}-1.052632(\varepsilon-1)-0.526316  \tag{10.4.14}\\
-0.775623(\varepsilon-1)^{2}-2.603878(\varepsilon-1)-1.828255 \\
-0.798061(\varepsilon-1)^{2}-2.763490(\varepsilon-1)-1.566429 \\
-(\varepsilon-1)^{2}-4.2(\varepsilon-1)-3.11 \\
-2(\varepsilon-1)-2.1
\end{array}\right]^{\prime}
$$

which places the eigenvalues of $\tilde{A}+B F_{\mathrm{P}}(\varepsilon)$ around at $0,0,0,1-\varepsilon$ and $1-\varepsilon$, and

$$
K_{\mathrm{Q}}(\varepsilon)=\left[\begin{array}{cc}
-10 & 4  \tag{10.4.15}\\
-10 \varepsilon & 5 \varepsilon \\
0 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right]
$$

which places the eigenvalues of $\tilde{A}+K_{\mathrm{Q}}(\varepsilon) C_{1}$ at $-0.6554,0.3777 \pm j 0.6726,0$ and $1-\varepsilon$. The maximum singular value plots of the corresponding closedloop transfer matrix $T_{h w}(z, \varepsilon)$ in Figure 10.4.1 show that the $H_{\infty}$-ADDPMS is attained as $\varepsilon$ tends to zero.

### 10.4.2. Reduced Order Output Feedback

In this subsection, we follow the procedure of Chapter 7 to design a reduced order output feedback controller. We will show that such a controller structure with appropriately chosen gain matrices also solves the general $H_{\infty}$-ADDPMS
for the discrete-time system (10.1.1). First of all, without loss of generality but for simplicity of presentation, we assume that the matrices $C_{1}$ and $D_{1}$ are already in the form,

$$
C_{1}=\left[\begin{array}{cc}
0 & C_{1,02}  \tag{10.4.16}\\
I_{k} & 0
\end{array}\right] \quad \text { and } \quad D_{1}=\left[\begin{array}{c}
D_{1,0} \\
0
\end{array}\right]
$$

where $k=\ell-\operatorname{rank}\left(D_{1}\right)$ and $D_{1,0}$ is of full rank. Next, we follow Steps 10.F.C. 1 and 10.F.C. 2 of the previous subsection to compute the constant matrix $N$, and form the following system,
where $\tilde{A}, \tilde{E}$ and $\tilde{C}_{2}$ are defined as in (10.4.3)-(10.4.5). Then, partition (10.4.17) as follows,

$$
\left\{\begin{aligned}
\binom{x_{1}(k+1)}{x_{2}(k+1)} & =\left[\begin{array}{ll}
A_{11} & A_{12} \\
A_{21} & A_{22}
\end{array}\right]\binom{x_{1}(k)}{x_{2}(k)}+\left[\begin{array}{l}
B_{1} \\
B_{2}
\end{array}\right] u(k)+\left[\begin{array}{l}
E_{1} \\
E_{2}
\end{array}\right] w(k), \\
\binom{y_{0}(k)}{y_{1}(k)} & =\left[\begin{array}{cc}
0 & C_{1,02} \\
I_{k} & 0
\end{array}\right]\binom{x_{1}(k)}{x_{2}(k)} \\
h(k) & =\left[\begin{array}{ll}
C_{2,1} & C_{2,2}
\end{array}\right]\binom{x_{1}(k)}{x_{2}(k)}+\begin{array}{ll}
D_{2} & u(k)+0 \\
h(k),
\end{array}
\end{aligned}\right.
$$

where the state $x$ of (10.4.17) is partitioned to two parts, $x_{1}$ and $x_{2}$; and $y$ is partitioned to $y_{0}$ and $y_{1}$ with $y_{1} \equiv x_{1}$. Thus, one needs to estimate only the state $x_{2}$ in the reduced order controller design. Next, define an auxiliary subsystem $\Sigma_{Q R}$ characterized by a matrix quadruple ( $A_{\mathrm{R}}, E_{\mathrm{R}}, C_{\mathrm{R}}, D_{\mathrm{R}}$ ), where

$$
\left(A_{\mathrm{R}}, E_{\mathrm{R}}, C_{\mathrm{R}}, D_{\mathrm{R}}\right)=\left(A_{22}, E_{2},\left[\begin{array}{c}
C_{1,02}  \tag{10.4.18}\\
A_{12}
\end{array}\right],\left[\begin{array}{c}
D_{1,0} \\
E_{1}
\end{array}\right]\right) .
$$

The following is a step-by-step algorithm that constructs the reduced order output feedback controller for the general discrete-time $H_{\infty}$-ADDPMS.

Step 10.R.C.1: (Construction of the gain matrix $F_{\mathrm{P}}(\varepsilon)$ ). Define an auxiliary system

$$
\left\{\begin{align*}
x(k+1) & =\tilde{A} x(k)+B u(k)+\tilde{E} w(k),  \tag{10.4.19}\\
y(k) & =\tilde{x}^{2}(k) \\
h(k) & =\tilde{C}_{2} x(k)+D_{2} u(k)+0 w(k),
\end{align*}\right.
$$

and then perform Steps 10.S. 1 to 10.S. 5 of the previous section to the above system to obtain a parameterized gain matrix $F(\varepsilon)$. Furthermore, we let $F_{\mathrm{P}}(\varepsilon)=F(\varepsilon)$.

Step 10.R.C.2: (Construction of the gain matrix $K_{\mathrm{R}}(\varepsilon)$ ). Define another auxiliary system

$$
\left\{\begin{align*}
x(k+1) & =A_{\mathrm{R}}^{\prime} x(k)+C_{\mathrm{R}}^{\prime} u(k)+C_{2,2}^{\prime} w(k),  \tag{10.4.20}\\
y(k) & =x(k) \\
h(k) & =E_{\mathrm{R}}^{\prime} x(k)+D_{\mathrm{R}}^{\prime} u(k)+0 \quad w(k)
\end{align*}\right.
$$

and then perform Steps 10.S. 1 to 10.S.5 of the previous section to the above system to obtain a parameterized gain matrix $F(\varepsilon)$. Similarly, we let $K_{\mathrm{R}}(\varepsilon)=F(\varepsilon)^{\prime}$.

Step 10.R.C.3: (Construction of the reduced order controller $\Sigma_{\mathrm{RC}}(\varepsilon)$ ). Let us partition $F_{\mathrm{P}}(\varepsilon)$ and $K_{\mathrm{R}}(\varepsilon)$ as,

$$
F_{\mathrm{P}}(\varepsilon)=\left[\begin{array}{ll}
F_{\mathrm{P} 1}(\varepsilon) & F_{\mathrm{P} 2}(\varepsilon)
\end{array}\right] \quad \text { and } \quad K_{\mathrm{R}}(\varepsilon)=\left[\begin{array}{ll}
K_{\mathrm{R} 0}(\varepsilon) & K_{\mathrm{R} 1}(\varepsilon) \tag{10.4.21}
\end{array}\right]
$$

in conformity with the partitions of $x=\binom{x_{1}}{x_{2}}$ and $y=\binom{y_{0}}{y_{1}}$, respectively. Then define

$$
\begin{equation*}
G_{\mathrm{R}}(\varepsilon)=\left[-K_{\mathrm{R} 0}(\varepsilon), \quad A_{21}+K_{\mathrm{R} 1}(\varepsilon) A_{11}-\left(A_{\mathrm{R}}+K_{\mathrm{R}}(\varepsilon) C_{\mathrm{R}}\right) K_{\mathrm{R} 1}(\varepsilon)\right] \tag{10.4.22}
\end{equation*}
$$

Finally, the parameterized reduced order output feedback controller is given by

$$
\Sigma_{\mathrm{RC}}(\varepsilon):\left\{\begin{align*}
x_{c}(k+1) & =A_{\mathrm{RC}}(\varepsilon) x_{c}(k)+B_{\mathrm{RC}}(\varepsilon) y(k)  \tag{10.4.23}\\
u(k) & =C_{\mathrm{RC}}(\varepsilon) x_{c}(k)+D_{\mathrm{RC}}(\varepsilon) y(k),
\end{align*}\right.
$$

where

$$
\begin{align*}
& A_{\mathrm{RC}}(\varepsilon):=A_{\mathrm{R}}+B_{2} F_{\mathrm{P} 2}(\varepsilon)+K_{\mathrm{R}}(\varepsilon) C_{\mathrm{R}}+K_{\mathrm{R} 1}(\varepsilon) B_{1} F_{\mathrm{P} 2}(\varepsilon) \\
& B_{\mathrm{RC}}(\varepsilon):=G_{\mathrm{R}}(\varepsilon)+\left[B_{2}+K_{\mathrm{R} 1}(\varepsilon) B_{1}\right]\left[0, F_{\mathrm{P} 1}(\varepsilon)-F_{\mathrm{P} 2}(\varepsilon) K_{\mathrm{R} 1}(\varepsilon)\right] \\
& C_{\mathrm{RC}}(\varepsilon):=F_{\mathrm{P} 2}(\varepsilon) \\
& D_{\mathrm{RC}}(\varepsilon):=\left[0, F_{\mathrm{P} 1}(\varepsilon)-F_{\mathrm{P} 2}(\varepsilon) K_{\mathrm{R} 1}(\varepsilon)\right]+N . \tag{10.4.24}
\end{align*}
$$

We have the following theorem.
Theorem 10.4.2. Consider the given system $\Sigma$ of (10.1.1). Assume that the problem of $H_{\infty}$ almost disturbance decoupling with internal stability for (10.1.1) is solvable, i.e., the solvability conditions of Theorem 10.2 .2 are satisfied. Then, the closed-loop system comprising (10.1.1) and the reduced order measurement feedback controller (10.4.23) has the following properties: For any given $\gamma>0$, there exists a positive scalar $\varepsilon^{*}>0$ such that for all $0<\varepsilon \leq \varepsilon^{*}$,

1. the closed-loop system is asymptotically stable; and
2. the $H_{\infty}$-norm of the closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$ is less than $\gamma$, i.e., $\left\|T_{h w}(z, \varepsilon)\right\|_{\infty}<\gamma$.

Hence, by Definition 10.1.1, the control law of (10.4.23) solves the $H_{\infty}$-ADDPMS for (10.1.1).

Proof. See Subsection 10.5.E.
We illustrate the above result in the following example.
Example 10.4.2. We again consider the given system as in Example 10.4.1. In what follows, we will construct a reduced order output feedback controller. We first partition

$$
\begin{gather*}
\tilde{A}=\left[\begin{array}{l|l}
A_{11} & A_{12} \\
\hline A_{21} & A_{22}
\end{array}\right]=\left[\begin{array}{r|rrrr}
1 & 1 & 1 & 1 & 0 \\
\hline 0 & 1 & 1 & 1 & 0 \\
0 & 0 & 0.1 & 1 & 0 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & -0.4 & -0.1 & 0
\end{array}\right],  \tag{10.4.25}\\
B=\left[\begin{array}{l}
B_{1} \\
\hline B_{2}
\end{array}\right]=\left[\begin{array}{l}
0 \\
\hline 0 \\
0 \\
0 \\
1
\end{array}\right], \tilde{E}=\left[\frac{E_{1}}{E_{2}}\right]=\left[\begin{array}{ll}
0 & 1 \\
\hline 0 & 0 \\
0 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right], \tag{10.4.26}
\end{gather*}
$$

and $A_{\mathrm{R}}=A_{22}, E_{\mathrm{R}}=E_{2}$, and

$$
C_{\mathrm{R}}=\left[\begin{array}{rrcc}
0.1 & 0.5 & 0.2 & 0.1  \tag{10.4.27}\\
1 & 1 & 1 & 0
\end{array}\right], \quad D_{\mathrm{R}}=\left[\begin{array}{ll}
1 & 0 \\
0 & 1
\end{array}\right]
$$

Following our algorithm, we obtain

$$
\begin{align*}
F_{\mathrm{P}}(\varepsilon)^{\prime} & =\left[\frac{F_{\mathrm{P} 1}(\varepsilon)^{\prime}}{F_{\mathrm{P} 2}(\varepsilon)^{\prime}}\right] \\
& =\left[\begin{array}{c}
-0.526316(\varepsilon-1)^{2}-1.052632(\varepsilon-1)-0.526316 \\
\hline-0.775623(\varepsilon-1)^{2}-2.603878(\varepsilon-1)-1.828255 \\
-0.798061(\varepsilon-1)^{2}-2.763490(\varepsilon-1)-1.566429 \\
-(\varepsilon-1)^{2}-4.2(\varepsilon-1)-3.11 \\
-2(\varepsilon-1)-2.1
\end{array}\right] \tag{10.4.28}
\end{align*}
$$

and

$$
K_{\mathrm{R}}(\varepsilon)=\left[K_{\mathrm{R} 0}(\varepsilon) \mid K_{\mathrm{R} 1}(\varepsilon)\right]=\left[\begin{array}{c|r}
0 & -\varepsilon  \tag{10.4.29}\\
0 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right]
$$



Figure 10.4.2: Max. singular values of $T_{h w}$ - Reduced order output feedback.
which places the eigenvalues of $A_{\mathrm{R}}+K_{\mathrm{R}}(\varepsilon) C_{\mathrm{R}}$ at $-0.6554,0.3777 \pm j 0.6726$, and $1-\varepsilon$. Also, we obtain a reduced order output feedback controller of the form (10.4.23) with all sub-matrices as defined in (10.4.24). The maximum singular value plots of the corresponding closed-loop transfer matrix $T_{h w}(z, \varepsilon)$ in Figure 10.4 .2 show that the $H_{\infty}$-ADDPMS is attained as $\varepsilon$ tends to zero. 国

### 10.5. Proofs of Main Results

### 10.5.A. Proof of Theorem $\mathbf{1 0 . 2}$. 1

We show the result of Theorem 10.2 .1 , i.e., the solvability conditions of the $H_{\infty}$-ADDPMS for the following full information system,

We first define the following auxiliary continuous-time system,

$$
\check{\Sigma}_{\mathrm{FI}}: \begin{cases}\dot{\tilde{x}}= & \check{A}  \tag{10.5.2}\\
\left.\check{y}+\check{B} \check{u}+\begin{array}{c}
\check{E} \\
\check{y} \\
=\binom{I}{0} \check{x} \\
\check{z} \\
\check{z} \\
\check{C}_{2} \\
I
\end{array}\right) \check{x}+\check{D}_{2} \check{u}+\check{D}_{22} & \check{w}\end{cases}
$$

where $\check{A}, \check{B}, \check{E}, \check{C}_{2}, \check{D}_{2}$ and $\check{D}_{22}$ are defined as

$$
\left.\begin{array}{rl}
\check{A} & =\left(A+B F_{0}+I\right)^{-1}\left(A+B F_{0}-I\right),  \tag{10.5.3}\\
\check{B} & =\sqrt{2}\left(A+B F_{0}+I\right)^{-1} B, \\
\check{E} & =\sqrt{2}\left(A+B F_{0}+I\right)^{-1} E, \\
\check{C}_{2} & =\sqrt{2}\left(C_{2}+D_{2} F_{0}\right)\left(A+B F_{0}+I\right)^{-1}, \\
\check{D}_{2} & =D_{2}-\left(C_{2}+D_{2} F_{0}\right)\left(A+B F_{0}+I\right)^{-1} B, \\
\check{D}_{22} & =D_{22}-\left(C_{2}+D_{2} F_{0}\right)\left(A+B F_{0}+I\right)^{-1} E,
\end{array}\right\}
$$

and where $F_{0}$ is chosen such that $A+B F_{0}$ has no eigenvalue at -1 . This can always be done provided that $(A, B)$ is stabilizable. For future use, we denote $\check{\Sigma}_{\mathrm{p}}$ as the subsystem characterized by ( $\check{A}, \check{B}, \breve{C}_{2}, \check{D}_{2}$ ). It was shown in Glover [45] (see also Chapter 4) that the infimum of $H_{\infty}$ optimization for the discrete-time system (10.5.1) is equivalent to that of $H_{\infty}$ optimization for the auxiliary continuous-time system (10.5.2). Thus, as a direct consequence, the $H_{\infty}$-ADDPMS for the discrete-time system (10.5.1) is solvable if and only if the $H_{\infty}$-ADDPMS for the continuous-time system (10.5.2) is solvable. Following the results of Scherer [95,96], one can show that the $H_{\infty}$-ADDPMS for (10.5.2) is solvable if and only if the following conditions are satisfied:
(a) $(\check{A}, \check{B})$ is stabilizable;
(b) there exists a matrix $\check{S}$ such that $\check{D}_{22}+\check{D}_{2} \check{S}=0$; and
(c) $\operatorname{Im}(\check{E}+\check{B} \check{S}) \subset \mathcal{S}^{+}\left(\check{\Sigma}_{\mathrm{P}}\right) \cap\left\{\cap_{\lambda \in \mathbb{C}^{0}} \mathcal{S}_{\lambda}\left(\check{\Sigma}_{\mathrm{P}}\right)\right\}$.

It is simple to show that $(A, B)$ is stabilizable if and only if $(\check{A}, \check{B})$ is stabilizable. Hence, it is sufficient to show Theorem 10.2 . 1 by showing that the following two statements are equivalent:

1. The first statement:
(a) There exists a $S$ such that $D_{22}+D_{2} S=0$;
(b) $\operatorname{Im}(E+B S) \subset\left\{\mathcal{V}^{\circ}\left(\Sigma_{\mathrm{P}}\right)+B \operatorname{Ker}\left(D_{2}\right)\right\} \cap\left\{\bigcap_{|\lambda|=1} \mathcal{S}_{\lambda}\left(\Sigma_{\mathrm{P}}\right)\right\}$.
2. The second statement:
(a) There exists a $\check{S}$ such that $\check{D}_{22}+\check{D}_{2} \check{S}=0$;
(b) $\operatorname{Im}(\check{E}+\check{B} \check{S}) \subset \mathcal{S}^{+}\left(\check{\Sigma}_{\mathrm{P}}\right) \cap\left\{\bigcap_{\lambda \in \mathbf{C}^{0}} \mathcal{S}_{\lambda}\left(\check{\Sigma}_{\mathrm{P}}\right)\right\}$.

Statement $1 \Rightarrow$ Statement 2: It is without loss of any generality to assume that matrix $D_{22}$ in (10.5.1) is equal to 0 . Also, by the definitions of the geometric subspaces $\mathcal{V}^{\mathrm{x}}, \mathcal{S}^{\mathrm{x}}, \mathcal{V}_{\lambda}$ and $\mathcal{S}_{\lambda}$, it is simple to verify that they are all invariant under any state feedback, output injection laws, and non-singular input as well as non-singular output transformations. Hereafter, we will assume that the subsystem $\Sigma_{\mathbf{P}}$, i.e., the quadruple ( $A, B, C_{2}, D_{2}$ ), is in the form of the special coordinate basis of Theorem 2.3.1. For easy reference in future development, we further assume that the state space of $\Sigma_{\mathrm{P}}$ has been decomposed as follows:

$$
\begin{equation*}
\mathcal{X}=\mathcal{X}_{a}^{0 *} \oplus \mathcal{X}_{a}^{-} \oplus \mathcal{X}_{c} \oplus \mathcal{X}_{a}^{+} \oplus \mathcal{X}_{b} \oplus \mathcal{X}_{d} \oplus \mathcal{X}_{a}^{01} \tag{10.5.4}
\end{equation*}
$$

where $\mathcal{X}_{a}^{01}$ is corresponding the zero dynamics of $\Sigma_{P}$ associated with the invariant zero at $z=-1$ and $\mathcal{X}_{a}^{0 *}$ is corresponding to the zero dynamics of $\Sigma_{\mathbf{P}}$ associated with the rest invariant zeros on the unit circle. More specifically, we let
$A=\left[\begin{array}{ccccccc}A_{a a}^{0 *} & 0 & 0 & 0 & L_{a b}^{0 *} C_{b} & L_{a d}^{0 *} C_{d} & 0 \\ 0 & A_{a a}^{-} & 0 & 0 & L_{a b}^{-} C_{b} & L_{a d}^{-} C_{d} & 0 \\ B_{c} E_{c a}^{0 *} & B_{c} E_{c a}^{-} & A_{c c} & B_{c} E_{c a}^{+} & L_{c b} C_{b} & L_{c d} C_{d} & B_{c} E_{c a}^{01} \\ 0 & 0 & 0 & A_{a a}^{+} & L_{a b}^{+} C_{b} & L_{a d}^{+} C_{d} & 0 \\ 0 & 0 & 0 & 0 & A_{b b} & L_{b d} C_{d} & 0 \\ B_{d} E_{d a}^{0 *} & B_{d} E_{d a}^{-} & B_{d} E_{d c} & B_{d} E_{d a}^{+} & B_{d} E_{d b} & A_{d d} & B_{d} E_{d a}^{01} \\ 0 & 0 & 0 & 0 & L_{a b}^{01} C_{b} & L_{a d}^{01} C_{d} & A_{a a}^{01}\end{array}\right]+B_{0} C_{2,0}$,

$$
\begin{gather*}
B=\left[\begin{array}{ll}
B_{0} & B_{1}
\end{array}\right]=\left[\begin{array}{ccc}
B_{0 a}^{0 *} & 0 & 0 \\
B_{0 a}^{-} & 0 & 0 \\
B_{0 c} & 0 & B_{c} \\
B_{0 a}^{+} & 0 & 0 \\
B_{0 b} & 0 & 0 \\
B_{0 d} & B_{d} & 0 \\
B_{0 a}^{01} & 0 & 0
\end{array}\right], E=\left[\begin{array}{c}
E_{a}^{0 *} \\
E_{a}^{-} \\
E_{c} \\
E_{a}^{+} \\
E_{b} \\
E_{d} \\
E_{a}^{01}
\end{array}\right],  \tag{10.5.6}\\
D_{2}=\left[\begin{array}{ccc}
I & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right],
\end{gather*}
$$

and

$$
C_{2}=\left[\begin{array}{l}
C_{2,0}  \tag{10.5.8}\\
C_{2,1}
\end{array}\right]=\left[\begin{array}{ccccccc}
C_{0 a}^{0 *} & C_{0 a}^{-} & C_{0 c} & C_{0 a}^{+} & C_{0 b} & C_{0 d} & C_{0 a}^{01} \\
0 & 0 & 0 & 0 & 0 & C_{d} & 0 \\
0 & 0 & 0 & 0 & C_{b} & 0 & 0
\end{array}\right]
$$

where $A_{a a}^{01}$ has all its eigenvalues at -1 and $A_{a a}^{0 *}$ has all its eigenvalues on the unit circle, but excluding the point -1 . Then, the condition in Statement 1(b) is equivalent to that

$$
\begin{equation*}
E_{a}^{+}=0, \quad E_{b}=0, \quad E_{a}^{01}=\left(I+A_{a a}^{01}\right) X_{a}^{01}, \quad E_{d}=B_{d} X_{d} \tag{10.5.9}
\end{equation*}
$$

for some appropriately dimensional $X_{a}^{01}$ and $X_{d}$, and

$$
\begin{equation*}
E_{a}^{0 *}=Y_{a a}^{0 *} X_{a}^{0 *}, \tag{10.5.10}
\end{equation*}
$$

where $Y_{a a}^{0 *}$ is a matrix whose columns span $\cap_{\alpha \in \lambda\left(A_{a a}^{0 *}\right)} \operatorname{Im}\left(\alpha I-A_{a a}^{0 *}\right)$ and $X_{a}^{0 *}$ is an appropriately dimensional matrix.

Let us now choose $F_{0}$ as,

$$
F_{0}=-\left[\begin{array}{ccccccc}
C_{0 a}^{0 *} & C_{0 a}^{-} & C_{0 c} & C_{0 a}^{+} & C_{0 b} & C_{0 d} & C_{0 a}^{01}  \tag{10.5.11}\\
E_{d a}^{0 *} & E_{d a}^{-} & E_{d c} & E_{d a}^{+} & E_{d b} & 0 & E_{d a}^{01}-E_{d a}^{01} \\
E_{c a}^{0 *} & E_{c a}^{-} & 0 & E_{c a}^{+} & 0 & 0 & 0
\end{array}\right] .
$$

Then, we have

$$
\hat{A}=A+B F_{0}=\left[\begin{array}{ccccccc}
A_{a a}^{0 *} & 0 & 0 & 0 & L_{a b}^{0 *} C_{b} & L_{a d}^{0 *} C_{d} & 0  \tag{10.5.12}\\
0 & A_{a a}^{-} & 0 & 0 & L_{a b}^{-} C_{b} & L_{a d}^{-} C_{d} & 0 \\
0 & 0 & A_{c c} & 0 & L_{c b} C_{b} & L_{c d} C_{d} & 0 \\
0 & 0 & 0 & A_{a a}^{+} & L_{a b}^{+} C_{b} & L_{a d}^{+} C_{d} & 0 \\
0 & 0 & 0 & 0 & A_{b b} & L_{b d} C_{d} & 0 \\
0 & 0 & 0 & 0 & 0 & A_{d d} & B_{d} \hat{E}_{d a}^{01} \\
0 & 0 & 0 & 0 & L_{a b}^{01} C_{b} & L_{a d}^{01} C_{d} & A_{a a}^{01}
\end{array}\right]
$$

and

$$
\hat{C}_{2}=C_{2}+D_{2} F_{0}=\left[\begin{array}{ccccccc}
0 & 0 & 0 & 0 & 0 & 0 & 0  \tag{10.5.13}\\
0 & 0 & 0 & 0 & 0 & C_{d} & 0 \\
0 & 0 & 0 & 0 & C_{b} & 0 & 0
\end{array}\right]
$$

For simplicity, we further assume that $A_{c c}, A_{b b}$ and $A_{d d}$ have no eigenvalue at -1 . Otherwise, some additional pre-state feedback will relocate them to
somewhere else. Also, $\hat{E}_{d a}^{01}$ is chosen such that $\hat{A}$ has no eigenvalue at -1 . Next, it can be computed that

$$
\left(A+B F_{0}+I\right)^{-1}=\left[\begin{array}{cccccccc}
\left(I+A_{a a}^{0 *}\right)^{-1} & 0 & 0 & 0 & X_{15} & X_{16} & X_{17} \\
0 & \left(I+A_{a a}^{-}\right)^{-1} & 0 & 0 & X_{25} & X_{26} & X_{27} \\
0 & & 0 & \left(I+A_{c c}\right)^{-1} & 0 & X_{35} & X_{36} & X_{37} \\
0 & 0 & 0 & 0 & \left(I+A_{a a}^{+}\right)^{-1} & X_{45} & X_{46} & X_{47} \\
0 & 0 & 0 & 0 & X_{55} & X_{56} & X_{57} \\
0 & 0 & 0 & 0 & X_{65} & X_{66} & X_{67} \\
0 & 0 & 0 & 0 & X_{75} & X_{76} & X_{77}
\end{array}\right]
$$

where

$$
\begin{align*}
X_{55}= & \left(I+A_{b b}\right)^{-1}\left\{I-L_{b d} C_{d}\left(I+A_{d d}\right)^{-1} B_{d} \hat{E}_{d a}^{01} \Delta^{-1} L_{a b}^{01} C_{b}\left(I+A_{b b}\right)^{-1}\right\},(10.5 .15) \\
X_{56}= & -\left(I+A_{b b}\right)^{-1} L_{b d}\left\{I+L_{b d} C_{d}\left(I+A_{d d}\right)^{-1} B_{d} \hat{E}_{d a}^{01} \Delta^{-1}\right. \\
& \left.\quad \times\left[L_{a d}^{01}-L_{a b}^{01} C_{b}\left(I+A_{b b}\right)^{-1} L_{b d}\right]\right\} C_{d}\left(I+A_{d d}\right)^{-1}, \\
X_{57}= & \left(I+A_{b b}\right)^{-1} L_{b d} C_{d}\left(I+A_{d d}\right)^{-1} B_{d} \hat{E}_{d a}^{01} \Delta^{-1}  \tag{10.5.17}\\
X_{65}= & \left(I+A_{d d}\right)^{-1} B_{d} \hat{E}_{d a}^{01} \Delta^{-1} L_{a b}^{01} C_{b}\left(I+A_{b b}\right)^{-1}  \tag{10.5.18}\\
X_{66}= & \left(I+A_{d d}\right)^{-1} \times \\
& \left\{B_{d} \hat{E}_{d a}^{01} \Delta^{-1}\left[L_{a d}^{01}-L_{a b}^{01} C_{b}\left(I+A_{b b}\right)^{-1} L_{b d}\right] C_{d}\left(I+A_{d d}\right)^{-1}+I\right\}  \tag{10.5.19}\\
X_{67}= & -\left(I+A_{d d}\right)^{-1} B_{d} \hat{E}_{d a}^{01} \Delta^{-1}  \tag{10.5.20}\\
X_{75}= & -\Delta^{-1} L_{a b}^{01} C_{b}\left(I+A_{b b}\right)^{-1}  \tag{10.5.21}\\
X_{76}= & \Delta^{-1}\left[L_{a b}^{01} C_{b}\left(I+A_{b b}\right)^{-1} L_{b d}-L_{a d}^{01}\right] C_{d}\left(I+A_{d d}\right)^{-1}  \tag{10.5.22}\\
X_{77}= & \Delta^{-1},  \tag{10.5.23}\\
X_{15}= & -\left(I+A_{a a}^{0 *}\right)^{-1}\left(L_{a b}^{0 *} C_{b} X_{55}+L_{a d}^{0 *} C_{d} X_{65}\right)  \tag{10.5.24}\\
X_{16}= & -\left(I+A_{a a}^{0 *}\right)^{-1}\left(L_{a b}^{0 *} C_{b} X_{56}+L_{a d}^{0 *} C_{d} X_{66}\right)  \tag{10.5.25}\\
X_{17}= & -\left(I+A_{a a}^{0 *}\right)^{-1}\left(L_{a b}^{0 *} C_{b} X_{57}+L_{a d}^{0 *} C_{d} X_{67}\right)  \tag{10.5.26}\\
X_{25}= & -\left(I+A_{a a}^{-}\right)^{-1}\left(L_{a b}^{-} C_{b} X_{55}+L_{a d}^{-} C_{d} X_{65}\right) \tag{10.5.27}
\end{align*}
$$

$$
\begin{align*}
& X_{26}=-\left(I+A_{a a}^{-}\right)^{-1}\left(L_{a b}^{-} C_{b} X_{56}+L_{a d}^{-} C_{d} X_{66}\right),  \tag{10.5.28}\\
& X_{27}=-\left(I+A_{a a}^{-}\right)^{-1}\left(L_{a b}^{-} C_{b} X_{57}+L_{a d}^{-} C_{d} X_{67}\right),  \tag{10.5.29}\\
& X_{35}=-\left(I+A_{c c}\right)^{-1}\left(L_{c b} C_{b} X_{55}+L_{c d} C_{d} X_{65}\right)  \tag{10.5.30}\\
& X_{36}=-\left(I+A_{c c}\right)^{-1}\left(L_{c b} C_{b} X_{56}+L_{c d} C_{d} X_{66}\right)  \tag{10.5.31}\\
& X_{37}=-\left(I+A_{c c}\right)^{-1}\left(L_{c b} C_{b} X_{57}+L_{c d} C_{d} X_{67}\right)  \tag{10.5.32}\\
& X_{45}=-\left(I+A_{a a}^{+}\right)^{-1}\left(L_{a b}^{+} C_{b} X_{55}+L_{a d}^{+} C_{d} X_{65}\right)  \tag{10.5.33}\\
& X_{46}=-\left(I+A_{a a}^{+}\right)^{-1}\left(L_{a b}^{+} C_{b} X_{56}+L_{a d}^{+} C_{d} X_{66}\right)  \tag{10.5.34}\\
& X_{47}=-\left(I+A_{a a}^{+}\right)^{-1}\left(L_{a b}^{+} C_{b} X_{57}+L_{a d}^{+} C_{d} X_{67}\right), \tag{10.5.35}
\end{align*}
$$

and where

$$
\begin{equation*}
\Delta=I+A_{a a}^{01}+\left[L_{a b}^{01} C_{b}\left(I+A_{b b}\right)^{-1} L_{b d}-L_{a d}^{01}\right] C_{d}\left(I+A_{d d}\right)^{-1} B_{d} \hat{E}_{d a}^{01} \tag{10.5.36}
\end{equation*}
$$

Furthermore, we have

$$
\check{B}=\sqrt{2}\left[\begin{array}{rcc}
\left(I+A_{a a}^{0 *}\right)^{-1} B_{0 a}^{0 *}+X_{15} B_{0 b}+X_{16} B_{0 d}+X_{17} B_{0 a}^{01} & X_{16} B_{d} & 0  \tag{10.5.37}\\
\left(I+A_{a a}^{-}\right)^{-1} B_{0 a}^{-}+X_{25} B_{0 b}+X_{26} B_{0 d}+X_{27} B_{0 a}^{01} & X_{26} B_{d} & 0 \\
\left(I+A_{c c}\right)^{-1} B_{0 c}+X_{35} B_{0 b}+X_{36} B_{0 d}+X_{37} B_{0 a}^{01} & X_{36} B_{d} & X_{c c} \\
\left(I+A_{a a}^{+}\right)^{-1} B_{0 a}^{+}+X_{45} B_{0 b}+X_{46} B_{0 d}+X_{47} B_{0 a}^{01} & X_{46} B_{d} & 0 \\
X_{55} B_{0 b}+X_{56} B_{0 d}+X_{57} B_{0 a}^{01} & X_{56} B_{d} & 0 \\
X_{65} B_{0 b}+X_{66} B_{0 d}+X_{67} B_{0 a}^{01} & X_{66} B_{d} & 0 \\
X_{75} B_{0 b}+X_{76} B_{0 d}+X_{77} B_{0 a}^{01} & X_{76} B_{d} & 0
\end{array}\right],
$$

where $X_{c c}=\left(I+A_{c c}\right)^{-1} B_{c}$,

$$
\left[\begin{array}{r}
\left(I+A_{a a}^{0 *}\right)^{-1} Y_{a a}^{0 *} X_{a}^{0 *}+X_{16} B_{d} X_{d}+X_{17}\left(I+A_{a a}^{01}\right) X_{a}^{01}  \tag{10.5.38}\\
\left(I+A_{a a}^{-}\right)^{-1} E_{a}^{-}+X_{26} B_{d} X_{d}+X_{27}\left(I+A_{a a}^{01}\right) X_{a}^{01} \\
\left(I+A_{c c}\right)^{-1} E_{c}+X_{36} B_{d} X_{d}+X_{37}\left(I+A_{a a}^{01}\right) X_{a}^{01} \\
X_{46} B_{d} X_{d}+X_{47}\left(I+A_{a a}^{01}\right) X_{a}^{01} \\
X_{56} B_{d} X_{d}+X_{57}\left(I+A_{a a}^{01}\right) X_{a}^{01} \\
X_{66} B_{d} X_{d}+X_{67}\left(I+A_{a a}^{01}\right) X_{a}^{01} \\
X_{76} B_{d} X_{d}+X_{77}\left(I+A_{a a}^{01}\right) X_{a}^{01}
\end{array}\right],
$$

$$
\check{D}_{2}=\left[\begin{array}{ccc}
I & 0 & 0  \tag{10.5.39}\\
-C_{d}\left(X_{65} B_{0 b}+X_{66} B_{0 d}+X_{67} B_{0 a}^{01}\right) & -C_{d} X_{66} B_{d} & 0 \\
-C_{b}\left(X_{55} B_{0 b}+X_{56} B_{0 d}+X_{57} B_{0 a}^{01}\right) & -C_{b} X_{56} B_{d} & 0
\end{array}\right]
$$

and

$$
\check{D}_{22}=\left[\begin{array}{c}
0  \tag{10.5.40}\\
-C_{d}\left[X_{66} B_{d} X_{d}+X_{67}\left(I+A_{a a}^{01}\right) X_{a}^{01}\right] \\
-C_{b}\left[X_{56} B_{d} X_{d}+X_{57}\left(I+A_{a a}^{01}\right) X_{a}^{01}\right]
\end{array}\right]
$$

Next, let us define

$$
\check{S}:=\left[\begin{array}{c}
0  \tag{10.5.41}\\
-X_{d}+\hat{E}_{d a}^{01} X_{a}^{01} \\
0
\end{array}\right] .
$$

Noting that

$$
\begin{equation*}
I+A_{a a}^{01}=\Delta-\left[L_{a b}^{01} C_{b}\left(I+A_{b b}\right)^{-1} L_{b d}-L_{a d}^{01}\right] C_{d}\left(I+A_{d d}\right)^{-1} B_{d} \hat{E}_{d a}^{01} \tag{10.5.42}
\end{equation*}
$$

it is straightforward to verify that

$$
\check{D}_{22}+\check{D}_{2} \check{S}=\left[\begin{array}{c}
0  \tag{10.5.43}\\
-C_{d}\left[X_{67}\left(I+A_{a a}^{01}\right) X_{a}^{01}+X_{66} B_{d} \hat{E}_{d a}^{01} X_{a}^{01}\right] \\
-C_{b}\left[X_{57}\left(I+A_{a a}^{01}\right) X_{a}^{01}+X_{56} B_{d} \hat{E}_{d a}^{01} X_{a}^{01}\right]
\end{array}\right]=0
$$

which shows that Statement 2(a) holds, and

$$
\begin{align*}
& =\sqrt{2}\left[\begin{array}{c}
\left(I+A_{a a}^{0 *}\right)^{-1} Y_{a a}^{0 *} X_{a}^{0 *}+X_{16} B_{d} \hat{E}_{d a}^{01} X_{a}^{01}+X_{17}\left(I+A_{a a}^{01}\right) X_{a}^{01} \\
\left(I+A_{a a}^{-}\right)^{-1} E_{a}^{-}+X_{26} B_{d} \hat{E}_{d a}^{01} X_{a}^{01}+X_{27}\left(I+A_{a a}^{01}\right) X_{a}^{01} \\
\left(I+A_{c c}\right)^{-1} E_{c}+X_{36} B_{d} \hat{E}_{d a}^{01} X_{a}^{01}+X_{37}\left(I+A_{a a}^{01}\right) X_{a}^{01} \\
X_{46} B_{d} \hat{E}_{d a}^{01} X_{a}^{01}+X_{47}\left(I+A_{a a}^{01}\right) X_{a}^{01} \\
X_{56} B_{d} \hat{E}_{d a}^{01} X_{a}^{01}+X_{57}\left(I+A_{a a}^{01}\right) X_{a}^{01} \\
X_{66} B_{d} \hat{E}_{d a}^{01} X_{a}^{01}+X_{67}\left(I+A_{a a}^{01}\right) X_{a}^{01} \\
X_{76} B_{d} \hat{E}_{d a}^{01} X_{a}^{01}+X_{77}\left(I+A_{a a}^{01}\right) X_{a}^{01}
\end{array}\right] \\
& =\sqrt{2}\left[\begin{array}{c} 
\\
\star \\
\star \\
0 \\
0 \\
\star
\end{array}\right], \tag{10.5.44}
\end{align*}
$$

where $\star$ 's are matrices of not much interest. Let the state space of $\check{\Sigma}_{\mathbf{P}}$, i.e., the matrix quadruple $\left(\check{A}, \check{B}, \check{C}_{2}, \check{D}_{2}\right)$, be decomposed as follows:

$$
\begin{equation*}
\check{\mathcal{X}}=\check{\mathcal{X}}_{a}^{0} \oplus \check{\mathcal{X}}_{a}^{-} \oplus \check{\mathcal{X}}_{c} \oplus \check{\mathcal{X}}_{a}^{+*} \oplus \check{\mathcal{X}}_{b} \oplus \check{\mathcal{X}}_{a}^{+1} \oplus \check{\mathcal{X}}_{d} \tag{10.5.45}
\end{equation*}
$$

where $\check{\mathcal{X}}_{a}^{0}, \check{\mathcal{X}}_{a}^{-}, \check{\mathcal{X}}_{c}, \check{\mathcal{X}}_{b}$ and $\check{\mathcal{X}}_{d}$ are the usual subspaces defined in the special coordinate basis of $\check{\Sigma}_{\mathbf{P}}$, while $\check{\mathcal{X}}_{a}^{+1}$ is corresponding to the zero dynamics of $\check{\Sigma}_{\mathbf{P}}$ associated with the invariant zero at $s=1$, and $\check{\mathcal{X}}_{a}^{+*}$ is corresponding to the zero dynamics of $\check{\Sigma}_{\mathrm{P}}$ associated with the rest unstable invariant zeros (excluding the point $s=1$ ). It was shown in Chapter 4, i.e., (4.1.100), that $\check{\mathcal{X}}$ of $\check{\Sigma}_{\mathrm{p}}$ and $\mathcal{X}$ of $\Sigma_{\mathrm{P}}$ are related by

$$
\begin{equation*}
\check{\mathcal{X}}_{a}^{0}=\mathcal{X}_{a}^{0 *}, \quad \check{\mathcal{X}}_{a}^{-}=\mathcal{X}_{a}^{-}, \quad \check{\mathcal{X}}_{c}=\mathcal{X}_{c}, \quad \check{\mathcal{X}}_{a}^{+*}=\mathcal{X}_{a}^{+} \tag{10.5.46}
\end{equation*}
$$

and

$$
\begin{equation*}
\check{\mathcal{X}}_{b}=\mathcal{X}_{b}, \quad \check{\mathcal{X}}_{a}^{+1}=\mathcal{X}_{d}, \quad \check{\mathcal{X}}_{d}=\mathcal{X}_{a}^{01} \tag{10.5.47}
\end{equation*}
$$

Moreover, the zero dynamics of $\check{\Sigma}_{\mathrm{p}}$ corresponding to the imaginary axis invariant zeros are fully characterized by the eigenstructure of the following matrix,

$$
\begin{equation*}
\check{A}_{a a}^{0}:=\left(A_{a a}^{0 *}+I\right)^{-1}\left(A_{a a}^{0 *}-I\right) \tag{10.5.48}
\end{equation*}
$$

Noting (10.5.10), it is ready to verify that

$$
\begin{equation*}
\operatorname{Im}\left\{\left(I+A_{a a}^{0 *}\right)^{-1} Y_{a a}^{0 *}\right\}=\bigcap_{\beta \in \lambda\left(\tilde{A}_{a a}^{0}\right)} \operatorname{Im}\left\{\beta I-\check{A}_{a a}^{0}\right\} \tag{10.5.49}
\end{equation*}
$$

It is now straightforward to see from (10.5.44) and the properties of the special coordinate basis that

$$
\begin{equation*}
\operatorname{Im}(\check{E}+\check{B} \check{S}) \subset \mathcal{S}^{+}\left(\check{\Sigma}_{\mathrm{P}}\right) \cap\left\{\cap_{\lambda \in \mathbb{C}^{0}} \mathcal{S}_{\lambda}\left(\check{\Sigma}_{\mathrm{P}}\right)\right\} \tag{10.5.50}
\end{equation*}
$$

i.e., Statement 2(b) holds.

Statement $2 \Rightarrow$ Statement 1: It follows by reversing the above arguments using the well-known bilinear transformation and the results of Chapter 4. Thus, it is omitted. This completes the proof of Theorem 10.2.1.

### 10.5.B. Proof of Theorem $\mathbf{1 0 . 2}$.2

For simplicity of presentation, we assume throughout this proof that matrix $A$ has no eigenvalue at -1 . Then, we define the following auxiliary continuoustime system,

$$
\check{\Sigma}:\left\{\begin{array}{l}
\dot{\bar{x}}=\check{A} \check{x}+\check{B} \check{u}+\check{E} \check{w}  \tag{10.5.51}\\
\check{y}=\check{C}_{1} \check{x} \\
\check{z}=\check{C}_{2} \check{x}+\check{D}_{2} \check{\sim}+\check{D}_{22} \check{w}
\end{array}\right.
$$

where $\check{A}, \check{B}, \check{E}, \check{C}_{1}, \check{D}_{1}, \check{C}_{2}, \check{D}_{2}$ and $\check{D}_{22}$ are defined as

$$
\begin{align*}
\check{A} & =(A+I)^{-1}(A-I) \\
\check{B} & =\sqrt{2}(A+I)^{-1} B \\
\check{E} & =\sqrt{2}(A+I)^{-1} E \\
\check{C}_{1} & =\sqrt{2} C_{1}(A+I)^{-1}  \tag{10.5.52}\\
\check{D}_{1} & =D_{1}-C_{1}(A+I)^{-1} E \\
\check{C}_{2} & =\sqrt{2} C_{2}(A+I)^{-1} \\
\check{D}_{2} & =D_{2}-C_{2}(A+I)^{-1} B \\
\check{D}_{22} & =D_{22}-C_{2}(A+I)^{-1} E
\end{align*}
$$

For easy reference later on, we let $\check{\Sigma}_{\mathrm{P}}$ denote the subsystem characterized by ( $\check{A}, \check{B}, \check{C}_{2}, \check{D}_{2}$ ) and $\check{\Sigma}_{Q}$ denote the subsystem characterized by characterized by ( $\check{A}, \check{E}, \check{C}_{1}, \check{D}_{1}$ ), respectively. Following the result of Glover [45], one can show that the following two statements are equivalent:

1. The $H_{\infty}$-ADDPMS for the originally given discrete-time system $\Sigma$ of (10.1.1) is solvable.
2. The $H_{\infty}$-ADDPMS for the auxiliary continuous-time system $\check{\Sigma}$ of (10.5.51) is solvable.

It was shown in Scherer $[95,96]$ that the second statement above is also equivalent to the following conditions (see also Theorem 7.2.1):
(a) $(\check{A}, \check{B})$ is stabilizable.
(b) $\left(\check{A}, \check{C}_{1}\right)$ is detectable.
(c) $\check{D}_{22}+\check{D}_{2} \check{S} \check{D}_{1}=0$, where $\check{S}=-\left(\check{D}_{2}^{\prime} \check{D}_{2}\right)^{\dagger} \check{D}_{2}^{\prime} \check{D}_{22} \check{D}_{1}^{\prime}\left(\check{D}_{1} \check{D}_{1}^{\prime}\right)^{\dagger}$.
(d) $\operatorname{Im}\left(\check{E}+\check{B} \check{S} \check{D}_{1}\right) \subset \mathcal{S}^{+}\left(\check{\Sigma}_{\mathrm{P}}\right) \cap\left\{\bigcap_{\lambda \in \mathbb{C}^{0}} \mathcal{S}_{\lambda}\left(\check{\Sigma}_{\mathrm{P}}\right)\right\}$.
(e) $\operatorname{Ker}\left(\check{C}_{2}+\check{D}_{2} \check{S}_{\check{C}_{1}}\right) \supset \mathcal{V}^{+}\left(\check{\Sigma}_{Q}\right) \cup\left\{\bigcup_{\lambda \in \mathbb{C}^{0}} \mathcal{V}_{\lambda}\left(\check{\Sigma}_{Q}\right)\right\}$.
(f) $\mathcal{V}^{+}\left(\check{\Sigma}_{Q}\right) \subset \mathcal{S}^{+}\left(\check{\Sigma}_{P}\right)$.

First, it is simple to check that the triple $\left(\check{A}, \check{B}, \check{C}_{1}\right)$ is stabilizable and detectable if and only if the triple $(A, B, C)$ is stabilizable and detectable. Next, following the proof in Subsection 10.5.A, we have the following equivalent statements:

## 1. Statement I:

(a) $D_{22}+D_{2} S D_{1}=0$, where $S=-\left(D_{2}^{\prime} D_{2}\right)^{\dagger} D_{2}^{\prime} D_{22} D_{1}^{\prime}\left(D_{1} D_{1}^{\prime}\right)^{\dagger}$;
(b) $\operatorname{Im}(E+B S) \subset\left\{\mathcal{V}^{\odot}\left(\Sigma_{\mathrm{P}}\right)+B \operatorname{Ker}\left(D_{2}\right)\right\} \cap\left\{\bigcap_{|\lambda|=1} \mathcal{S}_{\lambda}\left(\Sigma_{\mathrm{P}}\right)\right\}$.
2. Statement II:
(a) $\check{D}_{22}+\check{D}_{2} \check{S}_{1} \check{D}_{1}=0$, where $\check{S}=-\left(\check{D}_{2}^{\prime} \check{D}_{2}\right)^{\dagger} \check{D}_{2}^{\prime} \check{D}_{22} \check{D}_{1}^{\prime}\left(\check{D}_{1} \check{D}_{1}^{\prime}\right)^{\dagger}$;
(b) $\operatorname{Im}\left(\check{E}+\check{B} \check{S} \check{D}_{1}\right) \subset \mathcal{S}^{+}\left(\check{\Sigma}_{\mathrm{P}}\right) \cap\left\{\bigcap_{\lambda \in \mathbb{C}^{0}} \mathcal{S}_{\lambda}\left(\check{\Sigma}_{\mathrm{P}}\right)\right\}$.

Dualizing the arguments of Subsection 10.5.A, we can show that the following two statements are also equivalent:

1. Statement A:
(a) $D_{22}+D_{2} S D_{1}=0$, where $S=-\left(D_{2}^{\prime} D_{2}\right)^{\dagger} D_{2}^{\prime} D_{22} D_{1}^{\prime}\left(D_{1} D_{1}^{\prime}\right)^{\dagger}$;
(b) $\operatorname{Ker}\left(C_{2}+D_{2} S C_{1}\right) \supset\left\{\mathcal{S}^{\odot}\left(\Sigma_{Q}\right) \cap C_{1}^{-1}\left\{\operatorname{Im}\left(D_{1}\right)\right\}\right\} \cup\left\{\bigcup_{|\lambda|=1} \mathcal{V}_{\lambda}\left(\Sigma_{\mathrm{Q}}\right)\right\}$.
2. Statement B:
(a) $\check{D}_{22}+\check{D}_{2} \check{S} \check{D}_{1}=0$, where $\check{S}=-\left(\check{D}_{2}^{\prime} \check{D}_{2}\right)^{\dagger} \check{D}_{2}^{\prime} \check{D}_{22} \check{D}_{1}^{\prime}\left(\check{D}_{1} \check{D}_{1}^{\prime}\right)^{\dagger}$;
(b) $\operatorname{Ker}\left(\check{C}_{2}+\check{D}_{2} \check{S} \check{C}_{1}\right) \supset \mathcal{V}^{+}\left(\check{\Sigma}_{Q}\right) \cup\left\{\bigcup_{\lambda \in \mathbb{C}^{0}} \mathcal{V}_{\lambda}\left(\check{\Sigma}_{Q}\right)\right\}$.

Finally, it was shown in Chapter 4 that

$$
\begin{equation*}
\mathcal{V}^{\odot}\left(\Sigma_{\mathrm{P}}\right)=\mathcal{S}^{+}\left(\check{\Sigma}_{\mathrm{P}}\right), \quad \mathcal{S}^{\odot}\left(\Sigma_{\mathrm{P}}\right)=\mathcal{V}^{+}\left(\check{\Sigma}_{\mathrm{P}}\right) \tag{10.5.53}
\end{equation*}
$$

and

$$
\begin{equation*}
\mathcal{V}^{\odot}\left(\Sigma_{\mathbf{Q}}\right)=\mathcal{S}^{+}\left(\check{\Sigma}_{\mathbf{Q}}\right), \quad \mathcal{S}^{\odot}\left(\Sigma_{\mathbf{Q}}\right)=\mathcal{V}^{+}\left(\check{\Sigma}_{\mathbf{Q}}\right) \tag{10.5.54}
\end{equation*}
$$

Hence, the following two statements are equivalent:

1. $\mathcal{S}^{\odot}\left(\Sigma_{Q}\right) \subset \mathcal{V}^{\odot}\left(\Sigma_{P}\right)$.
2. $\mathcal{V}^{+}\left(\check{\Sigma}_{\mathrm{Q}}\right) \subset \mathcal{S}^{+}\left(\check{\Sigma}_{\mathrm{P}}\right)$.

Thus, the result of Theorem 10.2.2 follows.

### 10.5.C. Proof of Theorem 10.3.2

Without loss of any generality, but for simplicity of presentation, we assume that the matrix quadruple ( $A, B, C_{2}, D_{2}$ ) is in the form of the special coordinate basis of Theorem 2.3.1. It is simple to verify that if Condition (b) of Theorem 10.2.1 holds, we have

$$
D_{22}+D_{2} F_{w}=D_{22}+D_{2} S-\left[\begin{array}{lll}
I & 0 & 0  \tag{10.5.55}\\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right]\left[\begin{array}{c}
0 \\
\left(B_{d}^{\prime} B_{d}\right)^{-1} B_{d}^{\prime} E_{d} \\
0
\end{array}\right]=0
$$

Also, Condition (c) of Theorem 10.2.1 implies that

$$
E+B S=\left[\begin{array}{c}
E_{a}^{-}  \tag{10.5.56}\\
E_{a}^{0} \\
0 \\
0 \\
E_{c} \\
B_{d} X_{d}
\end{array}\right]
$$

with an appropriately dimensional $X_{d}$, and

$$
\begin{equation*}
E_{a}^{0}=Y_{a}^{0} X_{a}^{0} \tag{10.5.57}
\end{equation*}
$$

where $Y_{a}^{0}$ is a matrix whose columns span $\cap_{\alpha \in \lambda\left(A_{a a}^{0}\right)} \operatorname{Im}\left(\alpha I-A_{a a}^{0}\right)$ and $X_{a}^{0}$ is an appropriately dimensional matrix. Next, it is simple to verify that

$$
E+B F_{w}=\left[\begin{array}{c}
E_{a}^{-}  \tag{10.5.58}\\
E_{a}^{0} \\
0 \\
0 \\
E_{c} \\
B_{d} X_{d}-B_{d}\left(B_{d}^{\prime} B_{d}\right)^{-1} B_{d}^{\prime} B_{d} X_{d}
\end{array}\right]=\left[\begin{array}{c}
E_{a}^{-} \\
E_{a}^{0} \\
0 \\
0 \\
E_{c} \\
0
\end{array}\right] .
$$

Hence, we have

$$
\begin{equation*}
\operatorname{Im}\left(E+B F_{w}\right) \subset \mathcal{V}^{\odot}\left(\Sigma_{\mathbf{P}}\right) \cap\left\{\cap_{|\lambda|=1} \mathcal{S}_{\lambda}\left(\Sigma_{\mathbf{P}}\right)\right\} \tag{10.5.59}
\end{equation*}
$$

and the result follows from Theorem 10.3.1.

### 10.5.D. Proof of Theorem 10.4.1

We are to examine the result of Theorem 10.4.1. Let us first apply a pre-output feedback control law,

$$
\begin{equation*}
u=S y+\hat{u} \tag{10.5.60}
\end{equation*}
$$

with $S=-\left(D_{2}^{\prime} D_{2}\right)^{\dagger} D_{2}^{\prime} D_{22} D_{1}^{\prime}\left(D_{1} D_{1}^{\prime}\right)^{\dagger}$, to the given system $\Sigma$ of (10.1.1). Under Condition (c) of Theorem 10.2.2, we have $D_{22}+\hat{D}_{2} S D_{1}=0$. We also have a new system,

$$
\left\{\begin{array}{clllcl}
x(k+1) & = & \left(A+B S C_{1}\right) & x(k)+B \hat{u}(k)+\left(E+B S D_{1}\right) & w(k)  \tag{10.5.61}\\
y(k) & = & C_{1} & x(k) & + & D_{1} \\
h(k) & = & \left(C_{2}+D_{2} S D_{1}\right) & x(k)+D_{2} \hat{u}(k)+ & 0 & w(k)
\end{array}\right.
$$

We denote $\hat{\Sigma}_{\mathrm{P}}$ and $\hat{\Sigma}_{\mathbf{Q}}$ the subsystems characterized by the matrix quadruples $\left(A+B S C_{1}, B, C_{2}+D_{2} S C_{1}, D_{2}\right)$ and $\left(A+B S C_{1}, E+B S D_{1}, C_{1}, D_{1}\right)$, respectively. Recalling the definitions of $\mathcal{V}^{\odot}$ and $\mathcal{S}^{\odot}$, which are invariant under any state feedback and output injection laws, we have $\mathcal{V}^{\ominus}\left(\Sigma_{\mathrm{P}}\right)=\mathcal{V}^{\odot}\left(\hat{\Sigma}_{\mathrm{P}}\right)$, $\mathcal{S}^{\odot}\left(\Sigma_{Q}\right)=\mathcal{V}^{\odot}\left(\hat{\Sigma}_{Q}\right)$, and

$$
\left[\begin{array}{c}
A+B S C_{1}  \tag{10.5.62}\\
C_{2}+D_{2} S C_{1}
\end{array}\right] \mathcal{V}^{\odot}\left(\Sigma_{\mathrm{P}}\right) \subset\left(\mathcal{V}^{\ominus}\left(\Sigma_{\mathrm{P}}\right) \oplus\{0\}\right)+\operatorname{Im}\left\{\left[\begin{array}{c}
B \\
D_{2}
\end{array}\right]\right\}
$$

as well as

$$
\left[A+B S C_{1} \quad E+B S D_{1}\right]\left\{\left(\mathcal{S}^{\odot}\left(\Sigma_{\mathrm{Q}}\right) \oplus \mathbb{R}^{q}\right) \cap \operatorname{Ker}\left\{\left[\begin{array}{ll}
C_{1} & D_{1} \tag{10.5.63}
\end{array}\right]\right\}\right\} \subset \mathcal{S}^{\odot}\left(\Sigma_{\mathrm{Q}}\right)
$$

Furthermore, it can be easily verified that Condition (d) of Theorem 10.2.2 implies

$$
\operatorname{Im}\left\{\left[\begin{array}{c}
E+B S D_{1}  \tag{10.5.64}\\
0
\end{array}\right]\right\} \subset\left(\mathcal{V}^{\odot}\left(\Sigma_{\mathrm{P}}\right) \oplus\{0\}\right)+\operatorname{Im}\left\{\left[\begin{array}{c}
B \\
D_{2}
\end{array}\right]\right\}
$$

and that Condition (e) of Theorem 10.2.2 implies

$$
\left(\mathcal{S}^{\odot}\left(\Sigma_{Q}\right) \oplus \mathbb{R}^{q}\right) \cap \operatorname{Ker}\left\{\left[\begin{array}{ll}
C_{1} & D_{1}
\end{array}\right]\right\} \subset \operatorname{Ker}\left\{\left[\begin{array}{ll}
C_{2} & 0 \tag{10.5.65}
\end{array}\right]\right\}
$$

Next, it is ready to show that (10.5.62) and (10.5.64) together with Condition (f) of Theorem 10.2.2 imply that

$$
\left[\begin{array}{cc}
A+B S C_{1} & E+B S D_{1}  \tag{10.5.66}\\
C_{2}+D_{2} S C_{1} & 0
\end{array}\right]\left(\mathcal{S}^{\odot}\left(\Sigma_{Q}\right) \oplus \mathbb{R}^{q}\right) \subset\left(\mathcal{V}^{\odot}\left(\Sigma_{\mathrm{P}}\right) \oplus\{0\}\right)+\operatorname{Im}\left\{\left[\begin{array}{c}
B \\
D_{2}
\end{array}\right]\right\}
$$

and that (10.5.63) and (10.5.65) together with Condition (f) of Theorem 10.2.2 imply that

$$
\begin{align*}
{\left[\begin{array}{cc}
A+B S C_{1} & E+B S D_{1} \\
C_{2}+D_{2} S C_{1} & 0
\end{array}\right]\left\{\left(\mathcal{S}^{\odot}\left(\Sigma_{Q}\right) \oplus \mathbb{R}^{q}\right)\right.} & \left.\cap \operatorname{Ker}\left\{\left[\begin{array}{ll}
C_{1} & D_{1}
\end{array}\right]\right\}\right\} \\
& \subset\left(\mathcal{V}^{\ominus}\left(\Sigma_{\mathrm{P}}\right) \oplus\{0\}\right) \tag{10.5.67}
\end{align*}
$$

Finally, (10.5.66) and (10.5.67) imply that there exists a matrix $\tilde{N}$, which satisfies the following condition,

$$
\begin{array}{r}
\left(\left[\begin{array}{cc}
A+B S C_{1} & E+B S D_{1} \\
C_{2}+D_{2} S D_{1} & 0
\end{array}\right]+\left[\begin{array}{c}
B \\
D_{2}
\end{array}\right] \tilde{N}\left[\begin{array}{ll}
C_{1} & D_{1}
\end{array}\right]\right)\left(\mathcal{S}^{\odot}\left(\Sigma_{Q}\right) \oplus \mathbb{R}^{q}\right) \\
 \tag{10.5.68}\\
\subset\left(\mathcal{V}^{\odot}\left(\Sigma_{\mathrm{P}}\right) \oplus\{0\}\right)
\end{array}
$$

It is simple to verify that matrix $\tilde{N}:=N-S$, where $N$ is given as in (10.4.1), is one of the solutions to (10.5.68). Following the result of [101], one can show that matrix $N$ of (10.4.1) or $N=\tilde{N}+S$ with $\tilde{N}$ being any solution of (10.5.68) has the following properties:

$$
\begin{gather*}
\operatorname{Im}\left(E+B N D_{1}\right) \subset \mathcal{V}^{\odot}\left(\Sigma_{\mathbf{P}}\right)  \tag{10.5.69}\\
\operatorname{Ker}\left(C_{2}+D_{2} N C_{1}\right) \supset \mathcal{S}^{\odot}\left(\Sigma_{\mathbf{Q}}\right) \tag{10.5.70}
\end{gather*}
$$

and

$$
\begin{equation*}
\left(A+B N C_{1}\right) \mathcal{S}^{\odot}\left(\Sigma_{\mathrm{Q}}\right) \subset \mathcal{V}^{\odot}\left(\Sigma_{\mathrm{P}}\right), \quad D_{2} N D_{1}=0 \tag{10.5.71}
\end{equation*}
$$

Noting that $D_{2} N D_{1}=0$, it can be further showed using the compact form of the special coordinate basis that

$$
\begin{equation*}
\operatorname{Im}\left(E+B N D_{1}\right) \subset \mathcal{V}^{\odot}\left(\Sigma_{\mathrm{P}}\right) \cap\left\{\cap_{|\lambda|=1} \mathcal{S}_{\lambda}\left(\Sigma_{\mathrm{P}}\right)\right\} \tag{10.5.72}
\end{equation*}
$$

and

$$
\begin{equation*}
\operatorname{Ker}\left(C_{2}+D_{2} N C_{1}\right) \supset \mathcal{S}^{\odot}\left(\Sigma_{\mathbf{Q}}\right) \cup\left\{\cup_{|\lambda|=1} \mathcal{V}_{\lambda}\left(\Sigma_{\mathbf{Q}}\right)\right\} \tag{10.5.73}
\end{equation*}
$$

Now, let us apply the following pre-output feedback law, $u=N y+\tilde{u}$, to the system (10.1.1). We obtain

$$
\left\{\begin{align*}
x(k+1) & =\tilde{A} x(k)+B \tilde{u}(k)+\tilde{E} w(k),  \tag{10.5.74}\\
y(k) & =C_{1} x(k) \\
h(k) & =\tilde{C}_{2} x(k)+D_{2} \tilde{u}(k)+0 w(k),
\end{align*}\right.
$$

where $\tilde{A}, \tilde{E}$ and $\tilde{C}_{2}$ are as defined in (10.4.3) to (10.4.5). Clearly, it is sufficient to prove Theorem 10.4 .1 by showing the following controller

$$
\tilde{\Sigma}_{\mathrm{FC}}(\varepsilon):\left\{\begin{align*}
x_{c}(k+1) & =A_{\mathrm{FC}}(\varepsilon) x_{c}(k)+B_{\mathrm{FC}}(\varepsilon) y(k),  \tag{10.5.75}\\
\tilde{u}(k) & =C_{\mathrm{FC}}(\varepsilon) x_{c}(k)+0 \quad y(k)
\end{align*}\right.
$$

with $A_{\mathrm{FC}}(\varepsilon), B_{\mathrm{FC}}(\varepsilon)$ and $C_{\mathrm{FC}}(\varepsilon)$ being given as in (10.4.8), solves the $H_{\infty^{-}}$ ADDPMS for (10.5.74). For simplicity of presentation, we denote $\tilde{\Sigma}_{\mathrm{P}}$ the subsystem,

$$
\begin{equation*}
\left(\tilde{A}, B, \tilde{C}_{2}, D_{2}\right):=\left(A+B N C_{1}, B, C_{2}+D_{2} N C_{1}, D_{2}\right) \tag{10.5.76}
\end{equation*}
$$

and denote $\tilde{\Sigma}_{Q}$ the subsystem,

$$
\begin{equation*}
\left(\tilde{A}, \tilde{E}, C_{1}, D_{1}\right):=\left(A+B N C_{1}, E+B N D_{1}, C_{1}, D_{1}\right) \tag{10.5.77}
\end{equation*}
$$

It is simple to see that ( $\tilde{A}, B, C_{1}$ ) remains stabilizable and detectable. Also, it is trivial to show the stability of the closed-loop system comprising the given plant (10.5.74) and the controller (10.5.75). The closed-loop eigenvalues are given by $\lambda\left\{\tilde{A}+B F_{\mathrm{P}}(\varepsilon)\right\}$, which are in $\mathbb{C}^{\odot}$ for sufficiently small $\varepsilon$ as shown in Theorem 10.3.1, and $\lambda\left\{\tilde{A}+K_{\mathrm{Q}}(\varepsilon) C_{1}\right\}$, which can be dually shown to be in $\mathbb{C}^{\odot}$ for sufficiently small $\varepsilon$ as well. In what follows, we will show that the controller (10.5.75) achieves the $H_{\infty}$-ADDPMS for (10.5.74), under all the conditions of Theorem 10.2.2. By (10.5.71)-(10.5.73), and the fact that $\mathcal{V}^{\odot}, \mathcal{S}^{\odot}, \mathcal{V}_{\lambda}$ as well as $\mathcal{S}_{\lambda}$ are all invariant under any state feedback and output injection laws, we have that Conditions (d) to (f) of Theorem 10.2.2 are equivalent to the following conditions:
$(\tilde{d}) . \operatorname{Im}(\tilde{E}) \subset \mathcal{V}^{\odot}\left(\tilde{\Sigma}_{\mathbf{P}}\right) \cap\left\{\bigcap_{|\lambda|=1} \mathcal{S}_{\lambda}\left(\tilde{\Sigma}_{\mathrm{P}}\right)\right\} ;$
(ẽ). $\operatorname{Ker}\left(\tilde{C}_{2}\right) \supset \mathcal{S}^{\odot}\left(\tilde{\Sigma}_{Q}\right) \cup\left\{\bigcup_{|\lambda|=1} \mathcal{V}_{\lambda}\left(\tilde{\Sigma}_{\mathrm{Q}}\right)\right\} ;$
$(\tilde{f}) . \mathcal{S}^{\odot}\left(\tilde{\Sigma}_{\mathbf{Q}}\right) \subset \mathcal{V}^{\odot}\left(\tilde{\Sigma}_{\mathbf{P}}\right) ;$ and
( $\tilde{g}) . \tilde{A} \mathcal{S}^{\odot}\left(\tilde{\Sigma}_{Q}\right) \subset \mathcal{V}^{\odot}\left(\tilde{\Sigma}_{\mathrm{P}}\right)$.
Next, without of loss any generality but for simplicity of presentation, hereafter we assume throughout the rest of the proof that the subsystem $\tilde{\Sigma}_{\mathrm{P}}$, i.e., the quadruple ( $\tilde{A}, B, \tilde{C}_{2}, D_{2}$ ), has already been transformed into the special coordinate basis as given in Theorem 2.3.1. To be more specific, we have

$$
\begin{gather*}
\tilde{A}=B_{0} C_{2,0}+\left[\begin{array}{cccccc}
A_{a a}^{-a} & 0 & 0 & L_{a b}^{-} C_{b} & 0 & L_{a d}^{-} C_{d} \\
0 & A_{a a}^{0} & 0 & L_{a b}^{0} C_{b} & 0 & L_{a d}^{0} C_{d} \\
0 & 0 & A_{a a}^{+} & L_{a b}^{+} C_{b} & 0 & L_{a d}^{+} C_{d} \\
0 & 0 & 0 & A_{b b} & 0 & L_{b d} C_{d} \\
B_{c} E_{c a}^{-} & B_{c} E_{c a}^{0} & B_{c} E_{c a}^{+} & L_{c b} C_{b} & A_{c c} & L_{c d} C_{d} \\
B_{d} E_{d a}^{-} & B_{d} E_{d a}^{0} & B_{d} E_{d a}^{+} & B_{d} E_{d b} & B_{d} E_{d c} & A_{d d}
\end{array}\right] \\
:=B_{0} C_{2,0}+\bar{A},  \tag{10.5.78}\\
B=\left[\begin{array}{ccc}
B_{0 a}^{-} & 0 & 0 \\
B_{0 a}^{0} & 0 & 0 \\
B_{0 a}^{+} & 0 & 0 \\
B_{0 b} & 0 & 0 \\
B_{0 c} & 0 & B_{c} \\
B_{0 d} & B_{d} & 0
\end{array}\right], \quad B_{0}=\left[\begin{array}{c}
B_{0 a}^{-} \\
B_{0 a}^{0} \\
B_{0 a}^{+} \\
B_{0 b} \\
B_{0 c} \\
B_{0 d}
\end{array}\right], \tag{10.5.79}
\end{gather*}
$$

$$
\begin{gather*}
\tilde{C}_{2}=\left[\begin{array}{cccccc}
C_{0 a}^{-} & C_{0 a}^{0} & C_{0 a}^{+} & C_{0 b} & C_{0 c} & C_{0 d} \\
0 & 0 & 0 & 0 & 0 & C_{d} \\
0 & 0 & 0 & C_{b} & 0 & 0
\end{array}\right],  \tag{10.5.80}\\
C_{2,0}=\left[\begin{array}{llllll}
C_{0 a}^{-} & C_{0 a}^{0} & C_{0 a}^{+} & C_{0 b} & C_{0 c} & C_{0 d}
\end{array}\right], \quad D_{2}=\left[\begin{array}{lll}
I & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right], \tag{10.5.81}
\end{gather*}
$$

and

$$
\mathcal{V}^{\odot}\left(\tilde{\Sigma}_{\mathbf{P}}\right)=\operatorname{Im}\left\{\left[\begin{array}{ccc}
I & 0 & 0  \tag{10.5.82}\\
0 & I & 0 \\
0 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & I \\
0 & 0 & 0
\end{array}\right]\right\}
$$

It is simple to note that Condition ( $\tilde{d}$ ) implies that

$$
\tilde{E}=\left[\begin{array}{c}
E_{a}^{-}  \tag{10.5.83}\\
E_{a}^{0} \\
0 \\
0 \\
E_{c} \\
0
\end{array}\right]
$$

Next, for any $\zeta \in \mathcal{V}_{\lambda}\left(\tilde{\Sigma}_{Q}\right)$ with $\lambda \in \mathbb{C}^{\text {o }}$, we partition $\zeta$ as follows,

$$
\zeta=\left(\begin{array}{c}
\zeta_{a}^{-}  \tag{10.5.84}\\
\zeta_{a}^{0} \\
\zeta_{a}^{+} \\
\zeta_{b} \\
\zeta_{c} \\
\zeta_{d}
\end{array}\right)
$$

Then, Condition ( $\tilde{e}$ ) implies that $\tilde{C}_{2} \zeta=0$, or equivalently

$$
\begin{equation*}
C_{2,0} \zeta=0, \quad C_{b} \zeta_{b}=0 \quad \text { and } \quad C_{d} \zeta_{d}=0 \tag{10.5.85}
\end{equation*}
$$

By Definition 2.3.3, we have

$$
\left[\begin{array}{cc}
\tilde{A}-\lambda I & \tilde{E}  \tag{10.5.86}\\
C_{1} & D_{1}
\end{array}\right]\binom{\zeta}{\eta}=0
$$

for some appropriate vector $\eta$. Clearly, (10.5.86) and (10.5.83) imply that

$$
(\tilde{A}-\lambda I) \zeta=-\tilde{E} \eta=\left(\begin{array}{c}
\star  \tag{10.5.87}\\
\star \\
0 \\
0 \\
\star \\
0
\end{array}\right),
$$

where $\star$ 's are some vectors of not much interests. Note that (10.5.85) implies

$$
\begin{align*}
&(\tilde{A}-\lambda I) \zeta=\left(B_{0} C_{2,0}+\bar{A}-\lambda I\right) \zeta=(\bar{A}-\lambda I) \zeta \\
& \star \\
& \star  \tag{10.5.88}\\
&=\left[\begin{array}{c}
\left(A_{a a}^{+}-\lambda I\right) \zeta_{a}^{+}+L_{a b}^{+} C_{b} \zeta_{b}+L_{a d}^{+} C_{d} \zeta_{d} \\
\left(A_{b b}-\lambda I\right) \zeta_{b}+L_{b d} C_{d} \zeta_{d} \\
\star \\
\left(A_{d d}-\lambda I\right) \zeta_{d}+B_{d} \zeta_{x}
\end{array}\right] \\
& \star \\
& \star \\
& {\left[\begin{array}{c}
\left(A_{a a}^{+}-\lambda I\right) \zeta_{a}^{+} \\
\left(A_{b b}-\lambda I\right) \zeta_{b} \\
\star \\
\left(A_{d d}-\lambda I\right) \zeta_{d}+B_{d} \zeta_{x}
\end{array}\right] }
\end{align*}
$$

where

$$
\begin{equation*}
\zeta_{x}=E_{d a}^{-} \zeta_{a}^{-}+E_{d a}^{0} \zeta_{a}^{0}+E_{d a}^{+} \zeta_{a}^{+}+E_{d b} \zeta_{b}+E_{d c} \zeta_{c} \tag{10.5.89}
\end{equation*}
$$

(10.5.87) and (10.5.88) imply

$$
\begin{equation*}
\left(A_{a a}^{+}-\lambda I\right) \zeta_{a}^{+}=0, \quad\left(A_{b b}-\lambda I\right) \zeta_{b}=0 \tag{10.5.90}
\end{equation*}
$$

and

$$
\begin{equation*}
\left(A_{d d}-\lambda I\right) \zeta_{d}+B_{d} \zeta_{x}=0 \tag{10.5.91}
\end{equation*}
$$

Since $A_{a a}^{+}$has all its eigenvalues in $\mathbb{C}^{\otimes},\left(A_{a a}^{+}-\lambda I\right) \zeta_{a}^{+}=0$ implies that $\zeta_{a}^{+}=0$. Similarly, since $\left(A_{b b}, C_{b}\right)$ is completely observable, $\left(A_{b b}-\lambda I\right) \zeta_{b}=0$ and $C_{b} \zeta_{b}=0$ imply $\zeta_{b}=0$. Also, (10.5.91) and $C_{d} \zeta_{d}=0$ imply that

$$
\left[\begin{array}{cc}
A_{d d}-\lambda I & B_{d}  \tag{10.5.92}\\
C_{d} & 0
\end{array}\right]\binom{\zeta_{d}}{\zeta_{x}}=0
$$

Because ( $A_{d d}, B_{d}, C_{d}$ ) is invertible and is free of invariant zeros, (10.5.92) implies that $\zeta_{d}=0$ and $\zeta_{x}=0$. Thus, we have

$$
\zeta \in \operatorname{Ker}\left\{B_{d}\left[\begin{array}{llllll}
E_{d a}^{-} & E_{d a}^{0} & E_{d a}^{+} & E_{d b} & E_{d c} & 0 \tag{10.5.93}
\end{array}\right]\right\}
$$

and hence

$$
\mathcal{V}_{\lambda}\left(\tilde{\Sigma}_{\mathrm{q}}\right) \subset \operatorname{Ker}\left\{\begin{array}{llllll}
B_{d}\left[\begin{array}{lllll}
E_{d a}^{-} & E_{d a}^{0} & E_{d a}^{+} & E_{d b} & E_{d c}
\end{array}\right] \tag{10.5.94}
\end{array}\right]
$$

Moreover, $\zeta$ has the following property,

$$
\zeta=\left(\begin{array}{c}
\zeta_{a}^{-}  \tag{10.5.95}\\
\zeta_{a}^{0} \\
0 \\
0 \\
\zeta_{c} \\
0
\end{array}\right) \in \mathcal{V}^{\odot}\left(\tilde{\Sigma}_{\mathbf{P}}\right)
$$

Obviously, (10.5.95) together with Condition ( $\tilde{f}$ ) imply

$$
\begin{equation*}
\mathcal{V}^{\odot}\left(\tilde{\Sigma}_{\mathrm{P}}\right) \supset \mathcal{S}^{\odot}\left(\tilde{\Sigma}_{Q}\right) \cup\left\{U_{\lambda \in \mathbb{C}^{0}} \mathcal{V}_{\lambda}\left(\tilde{\Sigma}_{\mathbf{Q}}\right)\right\} \tag{10.5.96}
\end{equation*}
$$

Similarly, for any $\xi \in \mathcal{S}^{\odot}\left(\tilde{\Sigma}_{Q}\right)$, Conditions ( $\left.\tilde{e}\right)$ and $(\tilde{g})$ imply that $\tilde{C}_{2} \xi=0$ and

$$
\tilde{A} \xi=\left(\begin{array}{c}
\star  \tag{10.5.97}\\
\star \\
0 \\
0 \\
\star \\
0
\end{array}\right)
$$

Now, it is straightforward to show that

$$
\xi \in \operatorname{Ker}\left\{B_{d}\left[\begin{array}{llllll}
E_{d a}^{-} & E_{d a}^{0} & E_{d a}^{+} & E_{d b} & E_{d c} & 0 \tag{10.5.98}
\end{array}\right]\right\}
$$

and hence

$$
\mathcal{S}^{\odot}\left(\tilde{\Sigma}_{Q}\right) \subset \operatorname{Ker}\left\{\begin{array}{llllll}
\left.B_{d}\left[\begin{array}{lllll}
E_{d a}^{-} & E_{d a}^{0} & E_{d a}^{+} & E_{d b} & E_{d c}
\end{array}\right]\right\} . . . ~ \tag{10.5.99}
\end{array}\right.
$$

(10.5.94) and (10.5.99) imply that
$\operatorname{Ker}\left\{B_{d}\left[\begin{array}{llllll}E_{d a}^{-} & E_{d a}^{0} & E_{d a}^{+} & E_{d b} & E_{d c} & 0\end{array}\right]\right\} \supset \mathcal{S}^{\odot}\left(\tilde{\Sigma}_{\mathbf{Q}}\right) \cup\left\{\cup_{\lambda \in \mathbb{C}^{0}} \mathcal{V}_{\lambda}\left(\tilde{\Sigma}_{\mathbf{Q}}\right)\right\}$.
Next, we partition $\tilde{A}-z I$ as follows,

$$
\begin{equation*}
\tilde{A}-z I=X_{1}+X_{2} C_{2}+X_{3}+X_{4}+X_{5} \tag{10.5.101}
\end{equation*}
$$

where

$$
X_{1}:=\left[\begin{array}{cccccc}
A_{a a}^{-}-z I & 0 & 0 & L_{a b}^{-} C_{b} & 0 & L_{a d}^{-} C_{d}  \tag{10.5.102}\\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
B_{c} E_{c a}^{-} & B_{c} E_{c a}^{0} & B_{c} E_{c a}^{+} & L_{c b} C_{b} & A_{c c}-z I & L_{c d} C_{d} \\
0 & 0 & 0 & 0 & 0 & 0
\end{array}\right]
$$

$$
\begin{gather*}
X_{2}=\left[\begin{array}{ccc}
B_{0 a}^{-} & 0 & 0 \\
B_{0 a}^{0} & L_{a d}^{0} & L_{a b}^{0} \\
B_{0 a}^{+} & L_{a d}^{+} & L_{a b}^{+} \\
B_{0 b} & L_{b d} & 0 \\
B_{0 c} & 0 & 0 \\
B_{0 d} & 0 & 0
\end{array}\right],  \tag{10.5.103}\\
X_{3}=\left[\begin{array}{cccccc}
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & A_{a a}^{+}-z I & 0 & 0 & 0 \\
0 & 0 & 0 & A_{b b}-z I & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & A_{d d}-z I
\end{array}\right],  \tag{10.5.104}\\
X_{4}=\left[\begin{array}{cccccc}
0 & 0 & 0 & 0 & 0 & 0 \\
0 & A_{a a}^{0}-z I & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0
\end{array}\right], \tag{10.5.105}
\end{gather*}
$$

and

$$
X_{5}=\left[\begin{array}{cccccc}
0 & 0 & 0 & 0 & 0 & 0  \tag{10.5.106}\\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
B_{d} E_{d a}^{-} & B_{d} E_{d a}^{0} & B_{d} E_{d a}^{+} & B_{d} E_{d b} & B_{d} E_{d c} & 0
\end{array}\right] .
$$

It is simple to see that

$$
\begin{align*}
\operatorname{Im}\left(X_{1}\right) \subset \mathcal{V}^{\ominus}\left(\tilde{\Sigma}_{\mathrm{P}}\right) \cap\left\{\cap_{|\lambda|=1} \mathcal{S}_{\lambda}\left(\tilde{\Sigma}_{\mathrm{P}}\right)\right\}  \tag{10.5.107}\\
\operatorname{Ker}\left(X_{3}\right) \supset \mathcal{V}^{\odot}\left(\tilde{\Sigma}_{\mathrm{P}}\right) \supset \mathcal{S}^{\odot}\left(\tilde{\Sigma}_{\mathrm{Q}}\right) \cup\left\{\cup_{|\lambda|=1} \mathcal{V}_{\lambda}\left(\tilde{\Sigma}_{\mathrm{Q}}\right)\right\} . \tag{10.5.108}
\end{align*}
$$

Also, (10.5.100) implies that

$$
\begin{equation*}
\operatorname{Ker}\left(X_{5}\right) \supset \mathcal{S}^{\odot}\left(\tilde{\Sigma}_{\mathrm{Q}}\right) \cup\left\{\cup_{|\lambda|=1} \mathcal{V}_{\lambda}\left(\tilde{\Sigma}_{\mathrm{Q}}\right)\right\} \tag{10.5.109}
\end{equation*}
$$

It follows from the proof of Theorem 10.3 .1 that as $\varepsilon \rightarrow 0$

$$
\begin{equation*}
\left\|\left[\tilde{C}_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[z I-\tilde{A}-B F_{\mathrm{P}}(\varepsilon)\right]^{-1}\right\|_{\infty}<\kappa_{\mathrm{P}} \tag{10.5.110}
\end{equation*}
$$

where $\kappa_{\mathrm{P}}$ is a finite positive constant and is independent of $\varepsilon$. Moreover, under Condition ( $\tilde{d}$ ), we have

$$
\begin{equation*}
\left[\tilde{C}_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[z I-\tilde{A}-B F_{\mathrm{P}}(\varepsilon)\right]^{-1} \tilde{E} \rightarrow 0 \tag{10.5.111}
\end{equation*}
$$

and

$$
\begin{equation*}
\left[\tilde{C}_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[z I-\tilde{A}-B F_{\mathrm{P}}(\varepsilon)\right]^{-1} X_{1} \rightarrow 0 \tag{10.5.112}
\end{equation*}
$$

pointwise in $z$ as $\varepsilon \rightarrow 0$. It was proved in [65] that

$$
\begin{equation*}
\left[\tilde{C}_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[z I-\tilde{A}-B F_{\mathrm{P}}(\varepsilon)\right]^{-1} X_{4} \rightarrow 0 \tag{10.5.113}
\end{equation*}
$$

pointwise in $z$ as $\varepsilon \rightarrow 0$. Dually, one can show that

$$
\begin{equation*}
\left\|\left[z I-\tilde{A}-K_{\mathrm{Q}}(\varepsilon) C_{1}\right]^{-1}\left[\tilde{E}+K_{\mathrm{Q}}(\varepsilon) D_{1}\right]\right\|_{\infty}<\kappa_{\mathrm{Q}} \tag{10.5.114}
\end{equation*}
$$

where $\kappa_{\mathrm{Q}}$ is a finite positive constant and is independent of $\varepsilon$. If Condition ( $\tilde{e}$ ) is satisfied, the following results hold,

$$
\begin{align*}
& \tilde{C}_{2}\left[z I-\tilde{A}-K_{\mathrm{Q}}(\varepsilon) C_{1}\right]^{-1}\left[\tilde{E}+K_{\mathrm{Q}}(\varepsilon) D_{1}\right] \rightarrow 0  \tag{10.5.115}\\
& X_{3}\left[z I-\tilde{A}-K_{\mathrm{Q}}(\varepsilon) C_{1}\right]^{-1}\left[\tilde{E}+K_{\mathrm{Q}}(\varepsilon) D_{1}\right] \rightarrow 0 \tag{10.5.116}
\end{align*}
$$

and

$$
\begin{equation*}
X_{5}\left[z I-\tilde{A}-K_{\mathrm{Q}}(\varepsilon) C_{1}\right]^{-1}\left[\tilde{E}+K_{\mathrm{Q}}(\varepsilon) D_{1}\right] \rightarrow 0 \tag{10.5.117}
\end{equation*}
$$

pointwise in $z$ as $\varepsilon \rightarrow 0$.
Finally, it is simple to verify that the closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$ of the closed-loop system comprising the system (10.5.74) and the controller (10.5.75) is given by

$$
\begin{aligned}
T_{h w}(z, \varepsilon)= & {\left[\tilde{C}_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[z I-\tilde{A}-B F_{\mathrm{P}}(\varepsilon)\right]^{-1} \tilde{E} } \\
& +\tilde{C}_{2}\left[z I-\tilde{A}-K_{\mathrm{Q}}(\varepsilon) C_{1}\right]^{-1}\left[\tilde{E}+K_{\mathrm{Q}}(\varepsilon) D_{1}\right]+\left[\tilde{C}_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right] \\
& \cdot\left[z I-\tilde{A}-B F_{\mathrm{P}}(\varepsilon)\right]^{-1}(\tilde{A}-z I)\left[z I-\tilde{A}-K_{\mathrm{Q}}(\varepsilon) C_{1}\right]^{-1}\left[\tilde{E}+K_{\mathrm{Q}}(\varepsilon) D_{1}\right] .
\end{aligned}
$$

Using (10.5.101), we can rewrite $T_{h w}(z, \varepsilon)$ as

$$
\begin{aligned}
T_{h w}(z, \varepsilon)= & {\left[\tilde{C}_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[z I-\tilde{A}-B F_{\mathrm{P}}(\varepsilon)\right]^{-1} \tilde{E} } \\
& +\tilde{C}_{2}\left[z I-\tilde{A}-K_{\mathrm{Q}}(\varepsilon) C_{1}\right]^{-1}\left[\tilde{E}+K_{\mathrm{Q}}(\varepsilon) D_{1}\right] \\
& +\left[\tilde{C}_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[z I-\tilde{A}-B F_{\mathrm{P}}(\varepsilon)\right]^{-1}\left(X_{1}+X_{2} \tilde{C}_{2}+X_{3}+X_{4}+X_{5}\right) \\
& \cdot\left[z I-\tilde{A}-K_{\mathrm{Q}}(\varepsilon) C_{1}\right]^{-1}\left[\tilde{E}+K_{\mathrm{Q}}(\varepsilon) D_{1}\right] .
\end{aligned}
$$

Following (10.5.110) to (10.5.117), and some simple manipulations, it is straightforward to show that as $\varepsilon \rightarrow 0, T_{h w}(z, \varepsilon) \rightarrow 0$, pointwise in $z$, which is equivalent to $\left\|T_{h w}\right\|_{\infty} \rightarrow 0$ as $\varepsilon \rightarrow 0$. Hence, the full order output feedback controller (10.4.7) solves the $H_{\infty}$-ADDPMS for the given plant (10.1.1), provided that all the conditions of Theorem 10.2.2 are satisfied.

### 10.5.E. Proof of Theorem $\mathbf{1 0 . 4 . 2}$

It is sufficient to show Theorem 10.4 .2 by showing that the following controller,

$$
\tilde{\Sigma}_{\mathrm{RC}}(\varepsilon):\left\{\begin{array}{cl}
x_{c}(k+1) & =A_{\mathrm{RC}}(\varepsilon) x_{c}(k)+B_{\mathrm{RC}}(\varepsilon) y(k),  \tag{10.5.118}\\
\tilde{u}(k) & =C_{\mathrm{RC}}(\varepsilon) x_{c}(k)+\tilde{D}_{\mathrm{RC}}(\varepsilon) y(k),
\end{array}\right.
$$

with $A_{\mathrm{RC}}(\varepsilon), B_{\mathrm{RC}}(\varepsilon), C_{\mathrm{RC}}(\varepsilon)$ being given as in (10.4.24), and

$$
\begin{equation*}
\tilde{D}_{\mathrm{RC}}(\varepsilon)=\left[0, \quad F_{\mathrm{P} 1}(\varepsilon)-F_{\mathrm{P} 2}(\varepsilon) K_{\mathrm{R} 1}(\varepsilon)\right] \tag{10.5.119}
\end{equation*}
$$

solves the $H_{\infty}$-ADDPMS for (10.5.74). Again, it is trivial to show the stability of the closed-loop system comprising with (10.5.74) and the controller (10.5.118) as the closed-loop poles are given by $\lambda\left\{\tilde{A}+B F_{\mathrm{P}}(\varepsilon)\right\}$ and $\lambda\left\{A_{\mathrm{R}}+K_{\mathrm{R}}(\varepsilon) C_{\mathrm{R}}\right\}$, which are asymptotically stable for sufficiently small $\varepsilon$. Next, it is easy to compute the corresponding closed-loop transfer matrix from the disturbance $w$ to the controlled output $h$,

$$
\begin{aligned}
T_{h w}(z, \varepsilon)= & {\left[\tilde{C}_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[z I-\tilde{A}-B F_{\mathrm{P}}(\varepsilon)\right]^{-1} \tilde{E} } \\
& +\left[\tilde{C}_{2}+D_{2} F_{\mathrm{P}}(\varepsilon)\right]\left[z I-\tilde{A}-B F_{\mathrm{P}}(\varepsilon)\right]^{-1}(\tilde{A}-z I)\binom{0}{I_{n-k}} \\
& \cdot\left[z I-\tilde{A}_{\mathrm{R}}-K_{\mathrm{R}}(\varepsilon) C_{\mathrm{R}}\right]^{-1}\left[E_{\mathrm{R}}+K_{\mathrm{R}}(\varepsilon) D_{\mathrm{R}}\right] \\
& +\tilde{C}_{2}\binom{0}{I_{n-k}}\left[z I-\tilde{A}_{\mathrm{R}}-K_{\mathrm{R}}(\varepsilon) C_{\mathrm{R}}\right]^{-1}\left[E_{\mathrm{R}}+K_{\mathrm{R}}(\varepsilon) D_{\mathrm{R}}\right]
\end{aligned}
$$

Following the result of Chen [10] (i.e., Proposition 2.2.1), one can show that

$$
\begin{equation*}
\binom{0}{I_{n-k}} \mathcal{S}^{\odot}\left(\Sigma_{\mathrm{QR}}\right)=\mathcal{S}^{\odot}\left(\tilde{\Sigma}_{\mathrm{Q}}\right) \cap C_{1}^{-1}\left\{\operatorname{Im}\left(D_{1}\right)\right\} \tag{10.5.120}
\end{equation*}
$$

and

$$
\begin{equation*}
\binom{0}{I_{n-k}} \cup_{|\lambda|=1} \mathcal{V}_{\lambda}\left(\Sigma_{Q R}\right)=\cup_{|\lambda|=1} \mathcal{V}_{\lambda}\left(\tilde{\Sigma}_{Q}\right) \tag{10.5.121}
\end{equation*}
$$

Hence, we have

$$
\begin{align*}
\binom{0}{I_{n-k}} & \left(\mathcal{S}^{\odot}\left(\Sigma_{\mathbf{Q R}}\right) \cup\left\{\cup_{|\lambda|=1} \mathcal{V}_{\lambda}\left(\Sigma_{\mathrm{QR}}\right)\right\}\right) \\
& =\left\{\mathcal{S}^{\odot}\left(\tilde{\Sigma}_{\mathbf{Q}}\right) \cap C_{1}^{-1}\left\{\operatorname{Im}\left(D_{1}\right)\right\}\right\} \cup\left\{\cup_{|\lambda|=1} \mathcal{V}_{\lambda}\left(\tilde{\Sigma}_{\mathbf{Q}}\right)\right\} \\
& \subset \mathcal{S}^{\odot}\left(\tilde{\Sigma}_{\mathbf{Q}}\right) \cup\left\{\cup_{|\lambda|=1} \mathcal{V}_{\lambda}\left(\tilde{\Sigma}_{\mathbf{Q}}\right)\right\} \tag{10.5.122}
\end{align*}
$$

The rest of the proof follows from the same lines as those of Theorem 10.4.1. 因

## Chapter 11

## A Piezoelectric Actuator System Design

### 11.1. Introduction

We present in this chapter a case study on a piezoelectric bimorph actuator control system design using an $H_{\infty}$ optimization approach. This work was originally reported in Chen et al [21].

Piezoelectricity is a fundamental process in electromechanical energy conversion. It relates electric polarization to mechanical stress/strain in piezoelectric materials. Under the direct piezoelectric effect, an electric charge can be observed when the materials are deformed. The converse or the reciprocal piezoelectric effect is when the application of an electric field can cause mechanical stress/strain in the piezo materials. There are numerous piezoelectric materials available today with PZT (Lead Zirconate Titanate), PLZT (Lanthanum modified Lead Zirconate Titanate), and PVDF (Piezoelectric Polymeric Polyvinylidene Fluoride) to name a few (see Low and Guo [66]).

Piezoelectric structures are widely used in applications that require electrical to mechanical energy conversion coupled with size limitations, precision, and speed of operation. Typical examples are micro-sensors, micro-positioners, speakers, medical diagnostics, shutters and impact print hammers. In most applications, bimorph or stack piezoelectric structures are used because of the relatively high stress/strain to input electric field ratio (see Low and Guo [66]).

The present work is motivated by the possibility of applying piezoelectric micro-actuators in magnetic recording. The exponential growth of area densities seen in magnetic disk drives means that data tracks and data bits are being placed at closer proximity than ever before. The 25,000 TPI (tracks-per-inch)
track densities envisaged at the turn of the century mean that the positioning of the read/write ( $\mathrm{R} / \mathrm{W}$ ) heads could only tolerate at most 1 to 2 micro-inch error in track following. The closed loop positioning servo will also be required to have a bandwidth in excess of 1 to 2 kHz to be able to maintain this accuracy at the high spindle speeds required for channel data transfer rates, which will be in excess of $200 \mathrm{Mbits} / \mathrm{s}$. Such a performance is clearly out of reach with the present voice coil motor (VCM) actuators used in disk drive access systems.

A dual actuator was successfully demonstrated by Tsuchiura et al of Hitachi [107]. In [107], a fine positioner based on a piezoelectric structure was mounted at the end of a primary VCM stage to form the dual actuator. The higher bandwidth of the fine positioner allowed the R/W heads to be accurately positioned. There have been other instances where electromagnetic (see Miu and Tai [69]) and electrostatic (see Fan et al [41]) micro-actuators have been used for fine positioning of $\mathrm{R} / \mathrm{W}$ heads.

The focus of this chapter is to concentrate on the control issues involved in dealing with the nonlinear hysteresis behaviour displayed by most piezoelectric actuators. More specifically, we consider a robust controller design for a piezoelectric bimorph actuator as depicted in Figure 11.1.1. A scaled up model of this piezoelectric actuator, which is targeted for use in the secondary stage of a future dual actuator for magnetic recording, was actually built and modelled by Low and Guo [66]. It has two pairs of bimorph beams which are subjected to bipolar excitation. The dynamics of the actuator were identified in [66] as a second order linear model coupled with a hysteresis. The linear model is given by

$$
\begin{equation*}
m \ddot{x}_{1}+b \dot{x}_{1}+k x_{1}=k(d u-z) \tag{11.1.1}
\end{equation*}
$$

where $m, b, k$ and $d$ are the tangent mass, damping, stiffness and effective piezoelectric coefficients, while $u$ is the input voltage that generates excitation forces to the actuator system. The variable $x_{1}$ is the displacement of the actuator and it is also the only measurement we can have in this system. It should be noted that the working range of the displacement of this actuator is within $\pm 1 \mu \mathrm{~m}$. The variable $z$ is from the hysteretic nonlinear dynamics [66] and is governed by

$$
\begin{equation*}
\dot{z}=\alpha d \dot{u}-\beta|\dot{u}| z-\gamma \dot{u}|z|, \tag{11.1.2}
\end{equation*}
$$



1-base; 2-piezoelectric bimorph beams; 3-moving plate; and 4-guides
Figure 11.1.1: Structure of the piezoelectric bimorph actuator.


Figure 11.1.2: Piezoelectric bimorph actuator plant with controller.
where $\alpha, \beta$ and $\gamma$ are some constants that control the shapes of the hysteresis. For the actuator system that we are considering in this paper, the above coefficients are identified as follows:

$$
\left.\begin{array}{rlc}
m= & 0.01595 \mathrm{~kg}  \tag{11.1.3}\\
b= & 1.169 \mathrm{Ns} / \mathrm{m} \\
k= & 4385 \mathrm{~N} / \mathrm{m} \\
d= & 8.209 \times 10^{-7} \mathrm{~m} / \mathrm{V} \\
\alpha= & 0.4297, \\
\beta= & 0.03438 \\
\gamma= & -0.002865
\end{array}\right\}
$$

For a more detailed description of this piezoelectric actuator system and the identifications of the above parameters, we refer interested readers to the work of Low and Guo [66]. Our goal in this chapter is to design a robust controller, as in Figure 11.1.2, that meets the following design specifications:

1. The steady state tracking errors of the displacement should be less than $1 \%$ for any input reference signals that have frequencies ranging from 0 to 30 Hz , as the actuator is to be used to track certain color noise type of signals in disk drive systems.
2. The $1 \%$ settling time should be as fast as possible (we are able to achieve a $1 \%$ settling time of less than 0.003 seconds in our design).
3. The control input signal $u(t)$ should not exceed 112.5 volts because of the physical limitations of the piezoelectric materials.

Our approach is as follows: we will first use the stochastic equivalent linearization method proposed in Chang [8] to obtain a linearized model for the
nonlinear hysteretic dynamics. Then we reformulate our design into an $H_{\infty}$ almost disturbance decoupling problem in which the disturbance inputs are the reference input and the error between the hysteretic dynamics and that of its linearized model, while the controlled output is simply the double integration of the tracking error. Thus, our task becomes to design a controller such that when it is applied to the piezoelectric actuator, the overall system is asymptotically stable, and the controlled output, which corresponds to the tacking error, is as small as possible and decays as fast as possible.

The outline of this chapter is as follows: In Section 11.2, a first order linearized model is obtained for the nonlinear hysteresis using the stochastic equivalent linearization method. A simulation result is also given to show the match between the nonlinear and linearized models. In Section 11.3, we formulate our controller design into a standard almost disturbance decoupling problem by properly defining the disturbance input and the controlled output. Two integrators are augmented into the original plant to enhance the performance of the overall system. Then a robust controller that is explicitly parameterized by a certain tuning parameter and that solves the proposed almost disturbance decoupling problem, is carried out using a so-called asymptotic time-scale and eigenstructure assignment technique. In Section 11.4, we present the final controller and simulation results of our overall control system using MATLAB SIMULINK. We also obtain an explicit relationship between the peak values of the control signal and the tuning parameter of the controller, as well as an explicit linear relationship of the maximum trackable frequency, i.e, the corresponding tracking error can be settled to $1 \%$, vs the tuning parameter of the controller. The simulation results of this section clearly show that all the design specifications are met and the overall performance is very satisfactory.

### 11.2. Linearization of the Nonlinear Hysteretic Dynamics

We will proceed to linearize the nonlinear hysteretic dynamics of (11.1.2) in this section. As pointed out in Chang [8], there are basically three methods available in the literature to linearize the hysteretic type of nonlinear systems. These are i) the Fokker-Planck equation approach (see for example Caughey [34]), ii) the perturbation techniques (see for example Crandall [36] and Lyon [67]) and iii) the stochastic linearization approach. All of them have certain advantages and limitations. However, the stochastic linearization technique has the widest range of applications compared to the other methods. This method is based on the concept of replacing the nonlinear system with an "equivalent" linear system in such a way that the "difference" between these two systems is
minimized in a certain sense. The technique was initiated by Booton [6]. In this chapter, we will just follow the stochastic linearization method given in Chang [8] to obtain a linear model of the following form

$$
\begin{equation*}
\dot{z}=k_{1} \dot{u}+k_{2} z \tag{11.2.1}
\end{equation*}
$$

for the hysteretic dynamics of (11.1.2), where $k_{1}$ and $k_{2}$ are the linearization coefficients and are to be determined. The procedure is quite straightforward and proceeds as follows: First we introduce a so-called "difference" function $e$ between $\dot{z}$ of (11.1.2) and $\dot{z}$ of (11.2.1),

$$
\begin{equation*}
e\left(k_{1}, k_{2}\right)=\alpha d \dot{u}-\beta|\dot{u}| z-\gamma \dot{u}|z|-\left(k_{1} \dot{u}+k_{2} z\right) \tag{11.2.2}
\end{equation*}
$$

Then minimizing $\boldsymbol{E}\left[e^{2}\right]$, where $\boldsymbol{E}$ is the expectation operator, with respect to $k_{1}$ and $k_{2}$, we obtain

$$
\begin{equation*}
\frac{\partial \boldsymbol{E}\left[e^{2}\right]}{\partial k_{1}}=\frac{\partial \boldsymbol{E}\left[e^{2}\right]}{\partial k_{2}}=0 \tag{11.2.3}
\end{equation*}
$$

from which the stochastic linearization coefficients $k_{1}$ and $k_{2}$ are determined. It turns out that if $h$ and $\dot{u}$ are of zero means and jointly Gaussian, then $k_{1}$ and $k_{2}$ can be easily obtained. Let us assume that $h$ and $\dot{u}$ have a joint probability density function

$$
\begin{equation*}
f_{\dot{u} z}(\dot{u}, z)=\frac{1}{2 \pi \sigma_{\dot{u}} \sigma_{z} \sqrt{1-\rho_{\dot{u} z}^{2}}} \exp \left\{-\frac{\sigma_{\dot{u}}^{2} z^{2}-2 \sigma_{\dot{u}} \sigma_{z} \rho_{\dot{u} z} \dot{u} z+\sigma_{z}^{2} \dot{u}^{2}}{2 \sigma_{\dot{u}}^{2} \sigma_{z}^{2}\left(1-\rho_{\dot{u} z}^{2}\right)}\right\} \tag{11.2.4}
\end{equation*}
$$

where $\rho_{\dot{u} z}$ is the normalized covariance of $\dot{u}$ and $z$, and $\sigma_{\dot{u}}$ and $\sigma_{z}$ are the standard deviation of $\dot{u}$ and $z$, respectively. Then the linearization coefficients $k_{1}$ and $k_{2}$ can be expressed as follows:

$$
\begin{equation*}
k_{1}=\alpha d-\beta c_{1}-\gamma c_{2} \tag{11.2.5}
\end{equation*}
$$

and

$$
\begin{equation*}
k_{2}=-\beta c_{3}-\gamma c_{4}, \tag{11.2.6}
\end{equation*}
$$

where $c_{1}, c_{2}, c_{3}$ and $c_{4}$ are given by

$$
\begin{align*}
& c_{1}=0.79788456 \sigma_{z} \cos \left[\tan ^{-1}\left(\frac{\sqrt{1-\rho_{\dot{u} z}^{2}}}{\rho_{\dot{u} z}}\right)\right]  \tag{11.2.7}\\
& c_{2}=0.79788456 \sigma_{z}, \quad c_{4}=0.79788456 \rho_{\dot{u} z} \sigma_{\dot{u}} \tag{11.2.8}
\end{align*}
$$

and

$$
\begin{equation*}
c_{3}=0.79788456 \sigma_{\dot{u}}\left\{1-\rho_{\dot{u} z}^{2}+\rho_{\dot{u} z} \cos \left[\tan ^{-1}\left(\frac{\sqrt{1-\rho_{\dot{u} z}^{2}}}{\rho_{\dot{u} z}}\right)\right]\right\} \tag{11.2.9}
\end{equation*}
$$

After a few iterations, we found that a sinusoidal excitation $\dot{u}$ with frequencies ranging from 0 to 100 Hz (the expected working frequency range) and peak magnitude of 50 volts, which has a standard deviation of $\sigma_{\dot{u}}=35$, would yield a suitable linearized model for (11.1.2). For this excitation, we obtain $\sigma_{z}=5 \times 10^{-7}, \rho_{\dot{u} z}=5 \times 10^{-3}$

$$
\begin{gather*}
c_{1}=1.9947 \times 10^{-9}, \quad c_{2}=3.9894 \times 10^{-7}  \tag{11.2.10}\\
c_{3}=27.9260, \quad c_{4}=0.1396 \tag{11.2.11}
\end{gather*}
$$

and

$$
\begin{equation*}
k_{1}=3.5382 \times 10^{-7}, \quad k_{2}=-0.9597 \tag{11.2.12}
\end{equation*}
$$

The stochastic linearization model of the nonlinear hysteretic dynamics of (11.1.2) is then given by

$$
\begin{equation*}
\dot{\hat{z}}=k_{1} \dot{u}+k_{2} \hat{z}=3.5382 \times 10^{-7} \dot{u}-0.9597 \hat{z} \tag{11.2.13}
\end{equation*}
$$

For future use, let us define the linearization error as

$$
\begin{equation*}
e_{z}=z-\hat{z} \tag{11.2.14}
\end{equation*}
$$

Figure 11.2 .1 shows the open-loop simulation results of the nonlinear hysteresis and its linearized model, as well as their error for a typical sine wave input signal $u$. The results are quite satisfactory. Here we should note that because of the nature of our approach in controller design later in the next section, the variation of the linearized model within a certain range, which might result in larger linearization error, $e_{z}$, will not much affect the overall performance of the closed-loop system. We will formulate $e_{z}$ as a disturbance input and our controller will automatically reject it from the output response.

### 11.3. Formulation of the Problem as an $H_{\infty}$-ADDPMS

This section is the heart of this chapter. We will first formulate our control system design for the piezoelectric bimorph actuator into a standard $H_{\infty}$ almost disturbance decoupling problem, and then apply the results of Chapter 7 to check the solvability of the proposed problem. Finally, we will utilize the results in Chapter 7 to find an internally stabilizing controller that solves the proposed almost disturbance decoupling problem. Of course, most importantly, the resulting closed-loop system and its responses should meet all the design specifications as listed in Section 11.1. To do this, we will have to convert the


Figure 11.2.1: Responses of hysteresis and its linearized model to a sine input.


Figure 11.3.1: Augmented linearized model with controller.
dynamic model of (11.1.1) with the linearized model of the hysteresis into a state space form. Let us first define a new state variable

$$
\begin{equation*}
v=\hat{z}-k_{1} u \tag{11.3.1}
\end{equation*}
$$

Then from (11.2.13), we have

$$
\begin{equation*}
\dot{v}=\dot{\hat{z}}-k_{1} \dot{u}=k_{2} \hat{z}=k_{2} v+k_{1} k_{2} u \tag{11.3.2}
\end{equation*}
$$

Substituting (11.2.14) and (11.3.1) into (11.1.1), we obtain

$$
\begin{equation*}
\ddot{x}_{1}+\frac{b}{m} \dot{x}_{1}+\frac{k}{m} x_{1}+\frac{k}{m} v=\frac{k\left(d-k_{1}\right)}{m} u-\frac{k}{m} e_{z} . \tag{11.3.3}
\end{equation*}
$$

The overall controller structure of our approach is then depicted in Figure 11.3.1. Note that in Figure 11.3 .1 we have augmented two integrators after $e$, the tracking error between the displacement $x_{1}$ and the reference input signal $r$. We have observed a very interesting property of this problem, i.e., the more
integrators that we augment after the tracking error $e$, the smaller the tracking error we can achieve for the same level of control input $u$. Because our control input $u$ is limited to the range from -112.5 to 112.5 volts, it turns out that two integrators are needed in order to meet all the design specifications. It is clear to see that the augmented system has an order of five. Next, let us define the state of the augmented system as

$$
x=\left(\begin{array}{lllll}
x_{1} & \dot{x}_{1} & v & x_{4} & x_{5} \tag{11.3.4}
\end{array}\right)^{\prime},
$$

and the measurement output

$$
y=\left(\begin{array}{l}
y_{1}  \tag{11.3.5}\\
y_{2} \\
y_{3}
\end{array}\right)=\left(\begin{array}{l}
x_{1} \\
x_{4} \\
x_{5}
\end{array}\right)
$$

i.e., the original measurement of displacement $x_{1}$ plus two augmented states. The auxiliary disturbance input is

$$
\begin{equation*}
w=\binom{e_{z}}{r} \tag{11.3.6}
\end{equation*}
$$

and the output to be controlled, $h$, is simply the double integration of the tracking error. The state space model of the overall augmented system is then given by

$$
\Sigma:\left\{\begin{array}{l}
\dot{x}=A x+B u+E w  \tag{11.3.7}\\
y=C_{1} x+D_{1} w \\
h=C_{2} x+D_{2} u
\end{array}\right.
$$

with

$$
\begin{align*}
A & =\left[\begin{array}{ccccc}
0 & 1 & 0 & 0 & 0 \\
-k / m & -b / m & -k / m & 0 & 0 \\
0 & 0 & k_{2} & 0 & 0 \\
1 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 1 & 0
\end{array}\right] \\
& =\left[\begin{array}{ccccc}
0 & 1 & 0 & 0 & 0 \\
-274921.63 & -73.2915 & -274921.63 & 0 & 0 \\
0 & 0 & -0.9597 & 0 & 0 \\
1 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 1 & 0
\end{array}\right],  \tag{11.3.8}\\
B & =\left[\begin{array}{cc}
0 & 0 \\
k\left(d-k_{1}\right) / m \\
k_{1} k_{2} \\
0 \\
0
\end{array}\right]=\left[\begin{array}{cc}
0.12841 \\
-3.39561 \times 10^{-7} \\
0 \\
0
\end{array}\right] \tag{11.3.9}
\end{align*}
$$

$$
\begin{align*}
E & =\left[\begin{array}{crr}
0 & 0 \\
-k / m & 0 \\
0 & 0 \\
0 & -1 \\
0 & 0
\end{array}\right]=\left[\begin{array}{crr}
0 & 0 \\
-274921.63 & 0 \\
0 & 0 \\
0 & -1 \\
0 & 0
\end{array}\right]  \tag{11.3.10}\\
C_{1} & =\left[\begin{array}{lllll}
1 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 1 & 0 \\
0 & 0 & 0 & 0 & 1
\end{array}\right]  \tag{11.3.11}\\
D_{1} & =\left[\begin{array}{ll}
0 & 0 \\
0 & 0 \\
0 & 0
\end{array}\right],  \tag{11.3.12}\\
C_{2} & =\left[\begin{array}{lllll}
0 & 0 & 0 & 0 & 1
\end{array}\right]  \tag{11.3.13}\\
D_{2} & =0 \tag{11.3.14}
\end{align*}
$$

For the problem that we are considering here, it is simple to verify that the system $\Sigma$ of (11.3.7) has the following properties:

1. The subsystem $\left(A, B, C_{2}, D_{2}\right)$ is invertible and of minimum phase with one invariant zero at -1.6867 . It also has one infinite zero of order 4.
2. The subsystem $\left(A, E, C_{1}, D_{1}\right)$ is left invertible and of minimum phase with one invariant zero at -0.9597 and two infinite zeros of orders 1 and 2 , respectively.

Then it follows from Theorem 7.2.1 or Theorem 7.2.1 that the $H_{\infty}$-ADDPMS for (11.3.7) is solvable. In fact, one can design either a full order observer based controller or a reduced order observer based controller to solve this problem. For the full order observer based controller, the order of the disturbance decoupling controller (see Figure 11.3.1) will be 5 and the order of the final overall controller (again see Figure 11.3.1) will be 7 (the disturbance decoupling controller plus two integrators). On the other hand, if we use a reduced order observer in the disturbance decoupling controller, the total order of the resulting final overall controller will be reduced to 4 . From the practical point of view, the latter is much more desirable than the former. Thus, in what follows we will only focus on the controller design based on a reduced order observer. We can separate our controller design into two steps:

1. In the first step, we assume that all five states of $\Sigma$ in (11.3.7) are available and then design a static and parameterized state feedback control law,

$$
\begin{equation*}
u=F(\varepsilon) x \tag{11.3.15}
\end{equation*}
$$

such that it solves the almost disturbance decoupling problem for the state feedback case, i.e., $y=x$, by adjusting the tuning parameter $\varepsilon$ to an appropriate value.
2. In the second step, we design a reduced order observer based controller. It has a parameterized reduced order observer gain matrix $K_{2}(\varepsilon)$ that can be tuned to recover the performance achieved by the state feedback control law in the first step.

We will use the asymptotic time-scale and eigenstructure assignment (ATEA) design method of Chapter 7 to construct both the state feedback law and the reduced order observer gain. We would like to note that in principle, one can also apply the ARE (algebraic Riccati equation) based $H_{\infty}$ optimization technique (see for example Zhou and Khargonekar [116]) to solve this problem. However, because the numerical conditions of our system $\Sigma$ are very bad, we are unable to obtain any satisfactory solution from the ARE approach. We cannot get any meaningful solution for the associated $H_{\infty}$-CARE in MATLAB. In this sense and at least for this problem, the ATEA method is much more powerful than the ARE one. The software realization of the ATEA algorithm can be found in the Linear Systems and Control Toolbox developed by Chen [12]. The following is a closed form solution of the static state feedback parameterized gain matrix $F(\varepsilon)$ obtained using the ATEA method.

$$
F(\varepsilon)=\left[\begin{array}{lll}
\left(2.1410 \times 10^{6}-62.3004 / \varepsilon^{2}\right) & (570.7619-31.1502 / \varepsilon) \\
2.1410 \times 10^{6} & -62.3004 / \varepsilon^{3} & -31.1502 / \varepsilon^{4} \tag{11.3.16}
\end{array}\right], ~ \$
$$

where $\varepsilon$ is the tuning parameter that can be adjusted to achieve almost disturbance decoupling. It can be verified that the closed-loop system matrix, $A+B F(\varepsilon)$ is asymptotically stable for all $0<\varepsilon<\infty$ and the closed-loop transfer function from the disturbance $w$ to the controlled output $z, T_{z w}(\varepsilon, s)$, satisfying

$$
\begin{equation*}
\left\|T_{z w}(\varepsilon, s)\right\|_{\infty}=\left\|\left[C_{2}+D_{2} F(\varepsilon)\right][s I-A-B F(\varepsilon)]^{-1} E\right\|_{\infty} \rightarrow 0 \tag{11.3.17}
\end{equation*}
$$

as $\varepsilon \rightarrow 0$.
The next step is to design a reduced order observer based controller that will recover the performance of the above state feedback control law. First, let us perform the following nonsingular (permutation) state transformation to the system $\Sigma$ of (11.3.7),

$$
\begin{equation*}
x=T \tilde{x}, \tag{11.3.18}
\end{equation*}
$$

where

$$
T=\left[\begin{array}{lllll}
1 & 0 & 0 & 0 & 0  \tag{11.3.19}\\
0 & 0 & 0 & 1 & 0 \\
0 & 0 & 0 & 0 & 1 \\
0 & 1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0
\end{array}\right]
$$

such that the transformed measurement matrix has the form of

$$
C_{1} T=\left[\begin{array}{lllll}
1 & 0 & 0 & 0 & 0  \tag{11.3.20}\\
0 & 1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0
\end{array}\right]=\left[\begin{array}{ll}
I_{3} & 0
\end{array}\right]
$$

Clearly, the first three states of the transformed system, or $x_{1}, x_{4}$ and $x_{5}$ of the original system $\Sigma$ in (11.3.7), need not be estimated as they are already available from the measurement output. Let us now partition the transformed system as follows:

$$
\begin{align*}
T^{-1} A T & =\left[\begin{array}{l|l}
A_{11} & A_{12} \\
\hline A_{21} & A_{22}
\end{array}\right] \\
& =\left[\begin{array}{ccc|cc}
0 & 0 & 0 & 1 & 0 \\
1 & 0 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 \\
\hline-274921.63 & 0 & 0 & -73.2915 & -274921.63 \\
0 & 0 & 0 & 0 & -0.9597
\end{array}\right]  \tag{11.3.21}\\
T^{-1} B & =\left[\begin{array}{c}
B_{1} \\
\hline B_{2}
\end{array}\right]=\left[\right]  \tag{11.3.22}\\
T^{-1} E & =\left[\begin{array}{c}
E_{1} \\
\hline E_{2}
\end{array}\right]=\left[\begin{array}{cc}
0 & 0 \\
0 & -1 \\
0 & 0 \\
\hline-274921.63 & 0 \\
0 & 0
\end{array}\right] \tag{11.3.23}
\end{align*}
$$

Also, we partition

$$
\begin{align*}
F(\varepsilon) T & =\left[\begin{array}{lll}
F_{1}(\varepsilon) \mid F_{2}(\varepsilon)
\end{array}\right]  \tag{11.3.24}\\
& =\left[\begin{array}{lll}
\left(2.1410 \times 10^{6}-62.3004 / \varepsilon^{2}\right) & -62.3004 / \varepsilon^{3} & -31.1502 / \varepsilon^{4}
\end{array}\right. \\
& (570.7619-31.1502 / \varepsilon) \tag{11.3.25}
\end{align*}
$$

Then the reduced order observer based controller (see Chapter 7) is given in the form of

$$
\Sigma_{c}:\left\{\begin{array}{l}
\dot{v}=A_{c}(\varepsilon) v+B_{c}(\varepsilon) y  \tag{11.3.26}\\
u=C_{c}(\varepsilon) v+D_{c}(\varepsilon) y
\end{array}\right.
$$

with

$$
\begin{align*}
A_{c}(\varepsilon)= & A_{22}+K_{2}(\varepsilon) A_{12}+B_{2} F_{2}(\varepsilon)+K_{2}(\varepsilon) B_{1} F_{2}(\varepsilon)  \tag{11.3.27}\\
B_{c}(\varepsilon)= & A_{21}+K_{2}(\varepsilon) A_{11}-\left[A_{22}+K_{2}(\varepsilon) A_{12}\right] K_{2}(\varepsilon) \\
& \quad+\left[B_{2}+K_{2}(\varepsilon) B_{1}\right]\left[F_{1}(\varepsilon)-F_{2}(\varepsilon) K_{2}(\varepsilon)\right]  \tag{11.3.28}\\
C_{c}(\varepsilon)= & F_{2}(\varepsilon)  \tag{11.3.29}\\
D_{c}(\varepsilon)= & F_{1}(\varepsilon)-F_{2}(\varepsilon) K_{2}(\varepsilon), \tag{11.3.30}
\end{align*}
$$

where $K_{2}(\varepsilon)$ is the parameterized reduced order observer gain matrix and is to be designed such that $A_{22}+K_{2}(\varepsilon) A_{12}$ is asymptotically stable for sufficiently small $\varepsilon$ and also

$$
\begin{equation*}
\left\|\left[s I-A_{22}-K_{2}(\varepsilon) A_{12}\right]^{-1}\left[E_{2}+K_{2}(\varepsilon) E_{1}\right]\right\|_{\infty} \rightarrow 0 \tag{11.3.31}
\end{equation*}
$$

as $\varepsilon \rightarrow 0$. Again, using the software package of Chen [12], we obtained the following parameterized reduced order observer gain matrix

$$
K_{2}(\varepsilon)=\left[\begin{array}{ccc}
73.2915-1 / \varepsilon & 0 & 0  \tag{11.3.32}\\
0 & 0 & 0
\end{array}\right]
$$

Then the explicitly parameterized matrices of the state space model of the reduced order observer based controller are given by

$$
\begin{aligned}
& A_{c}(\varepsilon)=\left[\begin{array}{cc}
73.2915-4 / \varepsilon-1 / \varepsilon & 0 \\
-1.9381 \times 10^{-4}+1.0577 \times 10^{-5} / \varepsilon & -1.6867
\end{array}\right] \\
& C_{c}(\varepsilon)=\left[\begin{array}{lll}
570.7619-31.1502 / \varepsilon & 2140967
\end{array}\right] \\
& D_{c}(\varepsilon)=\left[\begin{array}{ccc}
2099135.4+2853.8095 / \varepsilon-93.4506 / \varepsilon^{2} & -62.3004 / \varepsilon^{3} & -31.1502 / \varepsilon^{4}
\end{array}\right] \\
& B_{c}(\varepsilon)=\left[\begin{array}{lll}
\psi_{1} & -8 / \varepsilon^{3} & -4 / \varepsilon^{4} \\
\psi_{2} & 2.1155 \times 10^{-5} / \varepsilon^{3} & 1.0577 \times 10^{-5} / \varepsilon^{4}
\end{array}\right]
\end{aligned}
$$

where

$$
\begin{equation*}
\psi_{1}=-5731.6533-13 / \varepsilon^{2}+439.7492 / \varepsilon \tag{11.3.33}
\end{equation*}
$$

and

$$
\begin{equation*}
\psi_{2}=-0.7128+3.1732 \times 10^{-5} / \varepsilon^{2}-9.6904 \times 10^{-4} / \varepsilon \tag{11.3.34}
\end{equation*}
$$

The overall closed loop system comprising the system $\Sigma$ of (11.3.7) and the above controller would be asymptotically stable as long as $\varepsilon \in(0, \infty)$. In fact, the closed loop poles are exactly located at -1.6867 , two pairs at $-1 / \varepsilon \pm j 1 / \varepsilon$, -0.9597 and $-1 / \varepsilon$. The plots of the maximum singular values of the closed


Figure 11.3.2: Max. singular values of closed loop transfer function $T_{z w}(\varepsilon, s)$.
loop transfer function matrix from the disturbance $w$ to the controlled output $z$, namely $T_{z w}(\varepsilon, s)$, for several values of $\varepsilon$, i.e., $\varepsilon=1 / 100, \varepsilon=1 / 400$ and $\varepsilon=1 / 3000$, in Figure 11.3 .2 show that as $\varepsilon$ becomes smaller and smaller, the $H_{\infty}$ norms of $T_{z w}(\varepsilon, s)$ are also smaller and smaller. Hence, almost disturbance decoupling is indeed achieved. These are the properties of our control system in the frequency domain. In the next section, we will address its time domain properties, which are of course much more important as all the design specifications are in the time domain.

### 11.4. Final Controller and Simulation Results

In this section, we will put our design of the previous section into a final controller as depicted in Figure 11.1.2. It is simple to derive the state space model of the final overall controller by observing its interconnection with the disturbance decoupling controller $\Sigma_{c}(\varepsilon)$ of (11.3.26) (see Figure 11.3.1). We will also present simulation results of the responses of the overall design to several different types of reference input signals. They clearly show that all the design specifications are successfully achieved. Furthermore, because our controller is explicitly parameterized by a tuning parameter, it is very easy to adjust to
meet other design specifications without going through it all over again from the beginning. This will also be discussed next.

As mentioned earlier, the final overall controller of our design will be of the order of 4 , of which two are from the disturbance decoupling controller and two from the augmented integrators. It has two inputs: one is the displacement $x_{1}$ and the other is the reference signal $r$. It is straightforward to verify that the state space model of the final overall controller is given by

$$
\Sigma_{o c}(\varepsilon):\left\{\begin{align*}
\dot{v}_{o c} & =A_{o c}(\varepsilon) v_{o c}+B_{o c}(\varepsilon) x_{1}+G_{o c} r  \tag{11.4.1}\\
u & =C_{o c}(\varepsilon) v_{o c}+D_{o c}(\varepsilon) x_{1}
\end{align*}\right.
$$

where $A_{o c}(\varepsilon)$ is given by

$$
\begin{aligned}
& {\left[\begin{array}{cccc}
73.2915-5 / \varepsilon & 0 & -8 / \varepsilon^{3} & -4 / \varepsilon^{4} \\
-0.0002+1.0577 \times 10^{-5} / \varepsilon & -1.6867 & 2.1155 \times 10^{-5} / \varepsilon^{3} & 1.0577 \times 10^{-5} / \varepsilon^{4} \\
0 & 0 & 0 & 0 \\
0 & 0 & 1 & 0
\end{array}\right],} \\
& G_{o c}=\left[\begin{array}{r}
0 \\
0 \\
-1 \\
0
\end{array}\right], \quad B_{o c}(\varepsilon)=\left[\begin{array}{c}
\psi_{1} \\
\psi_{2} \\
1 \\
0
\end{array}\right],
\end{aligned}
$$

with $\psi_{1}$ and $\psi_{2}$ given by (11.3.33) and (11.3.34), respectively,

$$
C_{o c}(\varepsilon)=\left[\begin{array}{lllll}
570.7619-31.1502 / \varepsilon & 2140967 & -62.3004 / \varepsilon^{3} & -31.1502 / \varepsilon^{4}
\end{array}\right]
$$

and

$$
D_{o c}(\varepsilon)=2099135.4-93.4506 / \varepsilon^{2}+2853.8095 / \varepsilon
$$

There are some very interesting and very useful properties of the above parameterized controller. After repeatedly simulating the overall design, we found that the maximum peak values of the control signal $u$ are independent of the frequencies of the reference signals. They are only dependent on the initial error between displacement, $x_{1}$, and the reference, $r$. The larger the initial error is, the bigger the peak that occurs in $u$. Because the working range of our actuator is within $\pm 1 \mu \mathrm{~m}$, we will assume that the largest magnitude of the initial error in any situation should not be larger that $1 \mu \mathrm{~m}$. This assumption is reasonable as we can always reset our displacement, $x_{1}$, to 0 before the system is to track any reference and hence the magnitude of initial tracking error can never be larger than $1 \mu \mathrm{~m}$. Let us consider the worst case, i.e., the magnitude of the initial error is $1 \mu \mathrm{~m}$. Then interestingly, we are able to obtain a clear relationship between the tuning parameter $1 / \varepsilon$ and the maximum peak of $u$. The result is plotted in Figure 11.4.1. We also found that the tracking error is independent


Figure 11.4.1: Parameter $1 / \varepsilon$ vs max. peaks of $u$ in worst initial errors.
of initial errors. It only depends on the frequencies of the references, i.e., the larger the frequency that the reference signal $r$ has, the larger the tracking error that occurs. Again, we can obtain a simple and linear relationship between the tuning parameter $\varepsilon$ and the maximum frequency that a reference signal can have such that the corresponding tracking error is no larger than $1 \%$, which is one of our main design specifications. The result is plotted in Figure 11.4.2.

Clearly, from Figure 11.4.1, we know that due to the constraints on the control input, i.e., it must be kept within $\pm 112.5$ volts, we have to select our controller with $\varepsilon>1 / 3370$. From Figure 11.4.2, we know that in order to meet the first design specification, i.e., the steady state tracking errors should be less than $1 \%$ for reference inputs that have frequencies up to 30 Hz , we have to choose our controller with $\varepsilon<1 / 2680$. Hence, the final controller as given in (11.4.1) to (11.4) will meet all the design goals for our piezoelectric actuator system. i.e., (11.1.1) and (11.1.2), for all $\varepsilon \in(1 / 3370,1 / 2680)$. Let us choose $\varepsilon=1 / 3000$. We obtain the overall controller as in the form of (11.4.1) with

$$
A_{o c}=\left[\begin{array}{cccc}
-14926.7085 & 0 & -2.16 \times 10^{11} & 3.24 \times 10^{14}  \tag{11.4.2}\\
0.0315 & -1.6867 & 5.7118 \times 10^{5} & 8.5677 \times 10^{8} \\
0 & 0 & 0 & 0 \\
0 & 0 & 1 & 0
\end{array}\right]
$$



Figure 11.4.2: Parameter $1 / \varepsilon$ vs max. frequency of $r$ that has $1 \%$ tracking error.

$$
\begin{gather*}
B_{o c}=\left[\begin{array}{c}
-1.1569 \times 10^{8} \\
281.9699 \\
1 \\
0
\end{array}\right], \quad G_{o c}=\left[\begin{array}{r}
0 \\
0 \\
-1 \\
0
\end{array}\right]  \tag{11.4.3}\\
C_{o c}=\left[\begin{array}{lll}
-92879.9041 & 2140967-1.6821 \times 10^{12} & -2.5232 \times 10^{15}
\end{array}\right], \tag{11.4.4}
\end{gather*}
$$

and

$$
\begin{equation*}
D_{o c}=-8.3040 \times 10^{8} \tag{11.4.5}
\end{equation*}
$$

The simulation results presented in the following are done using the MATLAB SIMULINK package, which is widely available everywhere these days. The SIMULINK simulation block diagram for the overall piezoelectric bimorph actuator system is given in Figure 11.4.3. Two different reference inputs are simulated using the Runge-Kutta 5 method in SIMULINK with a minimum step size of 10 micro-seconds and a maximum step size of 100 micro-seconds as well as a tolerance of $10^{-5}$. These references are: 1) a cosine signal with a frequency of 30 Hz and peak magnitude of $1 \mu \mathrm{~m}$, and 2) a sine signal with a frequency of 34 Hz and peak magnitude of $1 \mu \mathrm{~m}$. The results for the cosine signal are given in Figures 11.4.4 to 11.4.6. In Figure 11.4.4, the solid-line curve is $x_{1}$ and the dash-dotted curve is the reference. The tracking error and the control signal corresponding to this reference are given in Figures 11.4.5 and
11.4.6, respectively. Similarly, Figures 11.4 .7 to 11.4 .9 are the results corresponding to the sine signal. All these results show that our design goals are fully achieved. To be more specific, the tracking error for a 30 Hz cosine wave reference is about $0.8 \%$, which is better than the specification, and the worst peak magnitude of the control signal is less than 90 volts, which is of course less than the saturated level, i.e., 112.5 volts. Furthermore, the $1 \%$ tracking error settling times for both cases are less than 0.003 seconds.

Because the piezoelectric actuator is designed to be operated in a small neighborhood of its equilibrium point, the stability properties of the overall closed loop system of the nonlinear piezoelectric bimorph actuator should be similar to those of its linearized model. This fact can also be verified from simulations. In fact, the performance of the actual closed loop system is even better than that of its linear counterpart.

Finally, we would like to note that currently, we are still working on the actual implementation of our design. The outcome and result of the implementation will be reported in [21].


Figure 11.4.3: Simulation block diagram for the overall actuator control system.


Figure 11.4.4: Responses of the displacement and the 30 Hz cosine reference.


Figure 11.4.5: Tracking error for the 30 Hz cosine reference.

(a) Control signal from 0 to 0.004 seconds

(b) Control signal from 0.004 to 0.1 seconds

Figure 11.4.6: Control signal for the 30 Hz cosine reference.


Figure 11.4.7: Responses of the displacement and the 34 Hz sine reference.


Figure 11.4.8: Tracking error for the 34 Hz sine reference.


Figure 11.4.9: Control signal for the 34 Hz sine reference.

## Chapter 12

## A Gyro-stabilized Mirror Targeting System Design

### 12.1. Introduction

Electro-optical (E-O) SEnsors that are mounted on vehicles such as aircraft, helicopter and tanks are subjected to vibrations introduced by these platforms. These vibrations cause the line-of-sight (LOS) of the E-O sensors to shift, resulting in serious degradation of the image quality (see for example [5]). This problem is even more pronounced in systems with high magnification property. One way of overcoming it is to use free gyro-stabilization. A gyroscope or gyro is basically an axially symmetrical mass rotating at a high constant speed. With the magnitude of the angular inertia and the speed of rotation both kept constant, the momentum generated is also fixed. Bearing in mind that the momentum is a vector quantity, this implies that the directional orientation is maintained. Therefore, under the absence of large external forces, a gyro is capable of maintaining the orientation of its spin axis in the inertia space. By choosing an appropriate high value for the speed of rotation, the vibrational torque produced by the platforms can be made insignificant as compared to the momentum generated. The LOS can thus be stabilized by simply designing a system such that the LOS and the gyro's spin axis are parallel in space. However, a spinning gyro has another property known as precession. This means that if a torque is applied to one axis, it will contrary to the intuitions of mechanics, and rotate in the direction of another axis [78]. Thus, to enable for changes in the space orientation of the LOS, a gyro with at least two degrees of freedom is needed. This property also poses a problem in controlling the LOS as movement about one axis will cause a coupled movement in the other.

Therefore a controller has to be designed to provide the correct slewing (i.e., the application of a calculated torque to cause a desired precession).

In this chapter, we consider a multivariable servomechanism gyro-stabilized mirror system. More specifically, it is a two-input-and-two-output system. The control of this multiple-input-multiple-output system is not a simple problem using conventional PID controllers as there exist cross-coupling interactions between the dynamics of the two axes. In addition, it has to maintain stable operation even when there are changes in the system dynamics. Over the years, many researchers have worked on this system and the control methodologies studied include adaptive with feedforward paradigm (see e.g., [57]), neural network control (see e.g., [44]) and fuzzy logic (see e.g., [106] and [56]). Unfortunately, the controllers obtained using these techniques, except the one of [44], are in general too complicated to be implemented in the real system. Here we are tackling this problem using an $H_{\infty}$ control approach to design a simple and low order controller such that the overall closed-loop system would have fast tracking and good robustness performance. The work of this chapter was originally reported in a recent work of Siew, Chen and Lee [97].

This chapter is organized as follows: In Section 12.2, the mechanical setup of the free gyro-stabilized mirror system as well as its dynamical equations are given. This is followed by Section 12.3 where we formulate our controller design into an $H_{\infty}$ control problem by properly defining the disturbance input and the controlled output. A technique so-called asymptotic time-scale and eigenstructure assignment (ATEA) of Chapter 7 is then used to solve the proposed problem. Section 12.4 presents the simulation and implementation studies of our overall design. The results of both studies clearly show that all the design specifications are met and the overall performance is very satisfactory.

### 12.2. The Free Gyro-stabilized Mirror System

This section aims to give a brief overview of the hardware used in the whole free gyro-stabilized mirror system. The whole system consists of four main parts: a) a gyro mirror; b) a system interface assembly; c) a data acquisition board; and d) a personal computer. The overall hardware setup was pictured in Figure 12.2.1. In what follows, we give some brief descriptions of these four hardware parts.

## The Gyro Mirror

The most crucial part of the free gyro-stabilized mirror system is naturally the gyro-mirror itself. Figure 12.2 .2 is a schematic diagram of the gyro mirror. It


Figure 12.2.1: A gyro-stabilized mirror system.
consists of the following essential components: i) a flywheel and its spin motor; ii) gimbals that provide two degrees of freedom to the flywheel and two torque motors for slewing purposes; and iii) a mirror that is geared to the gimbals through a 2:1 reduction drive mechanism.

As no rigid body is able to spin forever in this practical world, a piece of pancake spin motor (flywheel) is used as the gyroscope (gyro). By adjusting the input torque, it can be made to spin at a high constant velocity about its spin axis (Axis 3 in Figure 12.2.2). The flywheel is mounted on an inner gimbal so that it can rotate freely up and down. This axis of rotation is named pitch axis and corresponds to Axis 2 in Figure 12.2.2. The inner gimbal is in turn mounted on an outer gimbal, which provides another axis of freedom (yaw axis or Axis 1) in moving left and right. Note that with these three axes being orthogonal to each other, the system's line-of-sight (LOS) can be made parallel to Axis 3 by aligning the mirror axis to the pitch axis.

A torque motor is attached to each of the inner and outer gimbals. These torque motors move the gyro either in the yaw or in the pitch direction, and are thus named the yaw and the pitch motors, respectively. By providing appropriate torque through these motors, the system can be precessed relative to the inertia space to achieve some desired line-of-sight (LOS). Once these input torques are removed, the LOS will be stabilized in its new position. The angular positions about the yaw and the pitch axes are defined as $\theta_{1}$ and $\theta_{2}$,


Figure 12.2.2: Schematic diagram of the gyroscope mirror.
respectively. $\theta_{1}$ and $\theta_{2}$ can be measured through potentiometers mounted on the inner and outer gimbals. There are however, no velocity sensor to sense $\dot{\theta}_{1}$ and $\dot{\theta}_{2}$. Due to physical constrains, the workspace for the gyro-stabilized mirror is limited to $-50^{\circ} \leq \theta_{1} \leq 50^{\circ}$ and $-30^{\circ} \leq \theta_{2} \leq 30^{\circ}$. Also, the maximum torques for both yaw and pitch motors are physically limited to a range from -0.5 Nm to 0.5 Nm .

In this particular system, a mirror is used in place of the actual electrooptical ( $\mathrm{E}-\mathrm{O}$ ) sensors. The advantage of doing this is that the E-O sensors will not form an integral part of the system. Therefore any E-O sensor can be used without affecting the system's dynamics. The mirror is connected to the flywheel-gimbal structure via a 2:1 reduction drive. This 2:1 reduction drive is required because when the mirror is tilted by an angle $\alpha$, the reflected LOS is rotated by $2 \alpha$.

The dynamical equations of the gyro mirror were developed by applying the well-known Lagrange's motion equation [72]:

$$
\begin{align*}
& M_{1}(\theta) \ddot{\theta}_{1}+H_{1}(\theta, \dot{\theta})+G_{1}\left(\theta, \dot{\theta}, \dot{\theta}_{3}\right)=u_{1},  \tag{12.2.1}\\
& M_{2}(\theta) \ddot{\theta}_{2}+H_{2}(\theta, \dot{\theta})+G_{2}\left(\theta, \dot{\theta}, \dot{\theta}_{3}\right)=u_{2}, \tag{12.2.2}
\end{align*}
$$

where $\theta=\left(\theta_{1}, \theta_{2}\right)^{\prime} ; u_{1}$ and $u_{2}$ are the actuator torques for the yaw and the pitch axes; $\dot{\theta}_{3}$ is the spin velocity of the flywheel. The parameters in equations (12.2.1)-(12.2.2) are defined as follows:

$$
\begin{align*}
& M_{1}=\bar{a}+\bar{d}+(\bar{b}-\bar{d}+\bar{\ell}) \cos ^{2} \theta_{2}+\frac{1}{2}(\bar{e}+\bar{g})+\frac{1}{2}(\bar{e}-\bar{g}) \sin \theta_{2},  \tag{12.2.3}\\
& H_{1}=-(\bar{b}-\bar{d}+\bar{\ell}) \dot{\theta}_{1} \dot{\theta}_{2} \sin 2 \theta_{2}+\frac{1}{2}(\bar{e}-\bar{g}) \dot{\theta}_{1} \dot{\theta}_{2} \cos \theta_{2}+\bar{k} \theta_{1} \dot{\theta}_{2} \sin \theta_{2} \cos \theta_{2},  \tag{12.2.4}\\
& G_{1}=\bar{k} \dot{\theta}_{2} \dot{\theta}_{3} \cos \theta_{2},  \tag{12.2.5}\\
& M_{2}=\bar{c}+\frac{\bar{f}}{4}+\bar{\ell},  \tag{12.2.6}\\
& H_{2}=\frac{1}{2}(\bar{b}-\bar{d}+\bar{\ell}) \dot{\theta}_{1}{ }^{2} \sin 2 \theta_{2}-\frac{1}{4}(\bar{e}-\bar{g}) \dot{\theta}_{1}{ }^{2} \cos \theta_{2}-\bar{k} \dot{\theta}_{1}{ }^{2} \sin \theta_{2} \cos \theta_{2},  \tag{12.2.7}\\
& G_{2}=-\bar{k} \dot{\theta}_{1} \dot{\theta}_{3} \cos \theta_{2}, \tag{12.2.8}
\end{align*}
$$

where $\bar{a}, \bar{b}, \bar{c}, \bar{d}, \bar{e}, \bar{f}, \bar{g}, \bar{\ell}$ and $\bar{k}$ are all physical constants representing the various moment of the inertia of the system. These constants were identified earlier by [72] and [56], and took on the following values:

$$
\begin{gathered}
\bar{a}=0.004, \quad \bar{b}=0.00128, \quad \bar{c}=0.00098, \quad \bar{d}=0.02 \\
\bar{e}=0.0049, \quad \bar{f}=0.0025, \quad \bar{g}=0.00125, \quad \bar{\ell}=0.0032, \quad \bar{k}=0.0025
\end{gathered}
$$



Figure 12.2.3: System interface assembly layout.

The above parameters all carry a unit of $\mathrm{kg} \cdot \mathrm{m}^{2}$. As can be seen from the above equations, the system is highly nonlinear and there exist cross-coupling terms between the yaw and the pitch axes.

## The System Interface Assembly

The torque motors and position sensors on the gyro-mirror have to be connected to a data acquisition board on the personal computer. This is accomplished via the system interface assembly. Figure 12.2 .3 shows the layout of the components assembled in this platform.

1. Power Supplies. The power supply units $A$ and $B$ are of single 28 V DC regulated type. They are connected in series to give a $-24 \mathrm{~V}-0 \mathrm{~V}-+24 \mathrm{~V}$ DC supply. This combined power unit supplies all the currents required by the torque motors, the position sensors and the analogue filters. Power supply unit C is rated 24 V DC. It is used solely to drive the flywheel controller.
2. Flywheel Controller. This is a dedicated driver unit (model MCH20-20-002CL) purchased commercially from BEI Motion Systems Company. It provides adjustable speed control to the spin motor via a potentiometer. The spin velocity ranges from 0 rpm up to around 5000 rpm .


Figure 12.2.4: Block diagram of experimental setup.
3. Current Amplifiers. There are two current amplifiers, one for the yaw motor and the other for the pitch motor. The inputs of the amplifiers are connected directly to the D/A outputs of the ADDA card, and their outputs are connected to the torque motors. They are built using a Power Operational Amplifier (PA51) from Apex Microtechnology Corporation and the outputs range from -25 V to +25 V . These outputs will produce corresponding torques ranging from -0.5 Nm to 0.5 Nm .
4. Analogue Filters. The position signals from the potentiometers are first passed through these filters before being connected to the A/D inputs of the ADDA card. They are low pass filters with cutoff frequency at 19 Hz so as to reject high frequency noises.

## The Data Acquisition Board

The analog-digital and digital-analog (ADDA) card used is DT2821 from Data Translation. Two analog input channels and two analog output channels are used. The analog inputs are the filtered position signals of the yaw and the pitch axes while the analog outputs are the torques to control the motors. The signals in all channels range from -10 V to +10 V DC, with a 12 bit accuracy.

## The Personal Computer

The controller is implemented on a personal computer via an ADDA card mounted within. The block diagram of the experimental setup is given in Figure 12.2.4. The personal computer configuration is an IBM PC compatible with: 1) an Intel Pentium 75 Processor; 2) a Numerical Co-processor, Intel 80387; 3) a 8 M-byte Main Board Memory; and 4) an MS-DOS 6.0 Operating System.


Figure 12.3.1: Structure of control system for gyro-stabilized mirror system.

### 12.3. Controller Design Using an $H_{\infty}$ Approach

In this section, we formulate our controller design for the free gyro-stabilized mirror system as an $H_{\infty}$ optimal control problem and then use a so-called asymptotic time-scale and eigenstructure assignment (ATEA) method of Chapter 7 to carry out the design of the controller. Our goal is to design a simple and low order controller as structured in Figure 12.3.1 such that the overall system will: i) have fast tracking in both the yaw and the pitch axes for step input commands with small or no overshoot; ii) minimize the cross-coupling interactions between the yaw and the pitch axes; and iii) ensure that the overall system is robust to external disturbances and changes in system parameters. As will be seen shortly, our controller is very simple and of low order. Thus, it can easily be implemented using low speed personal computers and A/D and D/A cards.

First of all, we need to linearize the dynamical model given in equations (12.2.1)-(12.2.2) and cast it into the standard state space form. The linearized state space model is given as follows:

$$
\begin{equation*}
\dot{x}_{g}=A_{g} x_{g}+B_{g} u+E_{g} w_{g} \tag{12.3.1}
\end{equation*}
$$

where $x_{g}=\left(\theta_{1}, \dot{\theta_{1}}, \theta_{2}, \dot{\theta_{2}}\right)^{\prime}, u=\left(u_{1}, u_{2}\right)^{\prime}$, and $w_{g} \in \mathcal{L}_{2}$ is the viscous damping coefficients for the system, which can be regarded as disturbances. The matrices $A_{g}, B_{g}$ and $E_{g}$ are given by

$$
A_{g}=\left[\begin{array}{cccc}
0 & 1 & 0 & 0  \tag{12.3.2}\\
0 & 0 & 0 & -\bar{k} \dot{\theta_{3}} / N_{1} \\
0 & 0 & 0 & 1 \\
0 & \bar{k} \dot{\theta_{3}} / N_{2} & 0 & 0
\end{array}\right]
$$

and

$$
B_{g}=\left[\begin{array}{cc}
0 & 0  \tag{12.3.3}\\
1 / N_{1} & 0 \\
0 & 0 \\
0 & 1 / N_{2}
\end{array}\right], \quad E_{g}=\left[\begin{array}{rr}
0 & 0 \\
-1 & 0 \\
0 & 0 \\
0 & -1
\end{array}\right]
$$

where

$$
\begin{equation*}
N_{1}=\bar{a}+\bar{b}+\frac{\bar{e}+\bar{g}}{2}+\bar{\ell}, \quad N_{2}=\bar{c}+\frac{\bar{f}}{4}+\bar{\ell} \tag{12.3.4}
\end{equation*}
$$

The measurement output of the free gyro-stabilized mirror system is

$$
\begin{equation*}
\theta=\binom{\theta_{1}}{\theta_{2}} \tag{12.3.5}
\end{equation*}
$$

Since we are interested in the changes in the orientation of the LOS, we focus only on the case where the command input $r(t)$ is a step function. To be more specific, we consider

$$
r(t)=\left[\begin{array}{l}
r_{1}(t)  \tag{12.3.6}\\
r_{2}(t)
\end{array}\right]=\left[\begin{array}{l}
\psi_{1} \\
\psi_{2}
\end{array}\right] 1(t)=\Psi \cdot 1(t)
$$

where $1(t)$ is the unit step function, and $\psi_{1}, \psi_{2}$ are some constants. Then, we have

$$
\dot{r}(t)=\left[\begin{array}{l}
\dot{r}_{1}(t)  \tag{12.3.7}\\
\dot{r}_{2}(t)
\end{array}\right]=\left[\begin{array}{l}
\psi_{1} \\
\psi_{2}
\end{array}\right] \delta(t)=\Psi \cdot \delta(t)
$$

where $\delta(t)$ is the unit impulse function. Let us define a controlled output $h$ as the difference between the actual output $\theta$ and the command input $r$, i.e.,

$$
\begin{equation*}
h=\theta-r=\binom{\theta_{1}-r_{1}}{\theta_{2}-r_{2}} \tag{12.3.8}
\end{equation*}
$$

Obviously, $h$ is simply the tracking error. Finally, we obtain the following system in the standard state space form:

$$
\Sigma:\left\{\begin{array}{l}
\dot{x}=A x+B u+E w  \tag{12.3.9}\\
y=C_{1} x+D_{1} w \\
h=C_{2} x+D_{2} u
\end{array}\right.
$$

with

$$
x=\binom{\tilde{x}_{g}}{r}=\left(\begin{array}{c}
\theta_{1}  \tag{12.3.10}\\
\dot{\theta}_{1} \\
\theta_{2} \\
\dot{\theta}_{2} \\
r_{1} \\
r_{2}
\end{array}\right), \quad u=\binom{u_{1}}{u_{2}}, \quad w=\binom{w_{g}}{\delta(t)}, \quad y=\left(\begin{array}{c}
\theta_{1} \\
\theta_{2} \\
r_{1} \\
r_{2}
\end{array}\right)
$$

$$
A=\left[\begin{array}{cc}
A_{g} & 0  \tag{12.3.11}\\
0 & 0
\end{array}\right], \quad B=\left[\begin{array}{c}
B_{g} \\
0
\end{array}\right], \quad E=\left[\begin{array}{cc}
E_{g} & 0 \\
0 & \Psi
\end{array}\right], \quad D_{1}=0, \quad D_{2}=0
$$

and

$$
C_{1}=\left[\begin{array}{llllll}
1 & 0 & 0 & 0 & 0 & 0  \tag{12.3.12}\\
0 & 0 & 1 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & 0 & 0 & 0 & 1
\end{array}\right], \quad C_{2}=\left[\begin{array}{rrrrrr}
1 & 0 & 0 & 0 & -1 & 0 \\
0 & 0 & 1 & 0 & 0 & -1
\end{array}\right] .
$$

At first glance, the matrix pair $(A, B)$ may look scary as two uncontrollable modes at $s=0$ are added. We would like to note that the augmented state $r(t)$ is actually the command input and hence does not need to be controlled. These uncontrollable modes will disappear when the final controller is implemented to the original free gyro-stabilized mirror system and the overall closed-loop system will be asymptotically stable. As will be seen shortly, a perfect tracking can be achieved with the above formulation. Our next step is to use the ATEA method of Chapter 7 to design a controller of the form:

$$
\Sigma_{c}:\left\{\begin{array}{l}
\dot{v}=A_{c} v+B_{c} y  \tag{12.3.13}\\
u=C_{c} v+D_{c} y
\end{array}\right.
$$

such that the effects of the 'disturbance' $w$ to the tracking error or controlled output $h$ is minimized. Here we note that we have no problem at all to handle the uncontrollable modes using the ATEA method. We just treat them as stable modes and then carry out our design. As mentioned earlier, these modes will disappear in the closed-loop system comprising the original system and the controller (12.3.13). If one wishes to solve the problem using an approach involved solving Riccati equations, then matrix $A$ should be replaced by

$$
\tilde{A}=\left[\begin{array}{cc}
A_{g} & 0  \tag{12.3.14}\\
0 & -\varepsilon I_{2}
\end{array}\right]
$$

where $\varepsilon$ is a small positive scalar. Using the toolbox of [12] or [60], we can show that

1. The subsystem $\left(A, B, C_{2}, D_{2}\right)$ is invertible with two invariant zeros at 0 , which comes from the command input. It also has two infinite zeros of order 2.
2. The subsystem $\left(A, E, C_{1}, D_{1}\right)$ is left invertible and of minimum phase with no invariant zero. It has one infinite zero of order 1 and two infinite zeros of order 2 .

In fact, it can be shown that for such a system we can achieve a robust and perfect tracking for the proposed problem, i.e., we can design a controller of (12.3.13) whose gain matrices are parameterized by a tuning variable, say $\varepsilon$, such that

$$
\begin{equation*}
\int_{0}^{\infty} h(t, \varepsilon)^{\prime} h(t, \varepsilon) d t=\int_{0}^{\infty}[\theta(t, \varepsilon)-r(t)]^{\prime}[\theta(t, \varepsilon)-r(t)] d t \rightarrow 0 \tag{12.3.15}
\end{equation*}
$$

as $\varepsilon \rightarrow 0$ for all $w \in \mathcal{L}_{2}$. Thus, in principle, $\theta(t)$ is capable to track the command $r(t)$ perfectly with no overshoot and with no time. Of course, the price one needs to pay for this kind of excellent performances is that the control input must be unlimited, i.e., using infinite gains. This is not possible in the real world. As mentioned earlier, the control inputs $u_{1}$ and $u_{2}$ of our problem are actually bounded from -0.5 Nm to 0.5 Nm . Therefore, a trade-off is needed.

Using the result of Chapter 7, one can either design a full order observer based controller or a reduced order observer based controller to solve the above problem. For the full order observer based controller, the order of the controller will be 6. On the other hand, a reduced order observer based controller will have an order of 2 since we only need to reconstruct the velocity states. Therefore from the practical point of view, a reduced order observer based controller is more desirable. We separate our controller design into the following two steps:

1. In the first step, we assume that all six states of $\Sigma$ in (12.3.9) are available and then design a static state feedback control law,

$$
\begin{equation*}
u=F x \tag{12.3.16}
\end{equation*}
$$

such that the closed-loop system has desired properties.
2. In the second step, we design a reduced order observer based controller. It has a reduced order observer gain matrix $K_{\mathrm{R}}$ that can recover the performance achieved by the state feedback control law in the first step.

Using the m-function atea. $m$ of the toolbox [12] and after a few iterations, we obtained the following state feedback gain:

$$
F=-\left[\begin{array}{rrrrrr}
2.3732 & 1.0271 & 1.4264 & 0.0000 & -2.3732 & -1.4264  \tag{12.3.17}\\
-1.4264 & 0.0000 & 2.3732 & 1.0113 & 1.4264 & -2.3732
\end{array}\right]
$$

Simulation result showed that the performance of the closed-loop system with the above state feedback law is quite satisfactory. Next, we proceed to design
the reduced order observer based controller. Let us first perform the following nonsingular state transformation to the system $\Sigma$,

$$
x=T \tilde{x}=\left[\begin{array}{llllll}
1 & 0 & 0 & 0 & 0 & 0  \tag{12.3.18}\\
0 & 0 & 0 & 0 & 1 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 1 \\
0 & 0 & 1 & 0 & 0 & 0 \\
0 & 0 & 0 & 1 & 0 & 0
\end{array}\right] \tilde{x}
$$

such that

$$
C_{1} T=\left[\begin{array}{llllll}
1 & 0 & 0 & 0 & 0 & 0  \tag{12.3.19}\\
0 & 1 & 0 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0 & 0 \\
0 & 0 & 0 & 1 & 0 & 0
\end{array}\right]=\left[\begin{array}{ll}
I_{4} & 0
\end{array}\right] .
$$

Clearly, the first four states of $\tilde{x}$, which are corresponding to $\theta_{1}, \theta_{2}, r_{1}$ and $r_{2}$, need not to be estimated. We further partition accordingly the transformed system as follows:

$$
T^{-1} A T=\left[\begin{array}{ll}
A_{11} & A_{12}  \tag{12.3.20}\\
A_{21} & A_{22}
\end{array}\right], T^{-1} B=\left[\begin{array}{l}
B_{1} \\
B_{2}
\end{array}\right], T^{-1} E=\left[\begin{array}{l}
E_{1} \\
E_{2}
\end{array}\right], F T=\left[\begin{array}{ll}
F_{1} & F_{2}
\end{array}\right],
$$

and define a reduced order system,

$$
\begin{equation*}
\left(A_{\mathrm{R}}, B_{\mathrm{R}}, C_{\mathrm{R}}, D_{\mathrm{R}}\right)=\left(A_{22}, E_{2}, A_{12}, E_{1}\right) . \tag{12.3.21}
\end{equation*}
$$

The reduced order observer based controller is then given as in the form of (12.3.13) with

$$
\begin{align*}
& A_{c}=A_{22}+K_{\mathrm{R}} A_{12}+B_{2} F_{2}+K_{\mathrm{R}} B_{1} F_{2}  \tag{12.3.22}\\
& B_{c}=A_{21}+K_{\mathrm{R}} A_{11}-\left(A_{22}+K_{\mathrm{R}} A_{12}\right) K_{\mathrm{R}}+\left(B_{2}+K_{\mathrm{R}} B_{1}\right)\left(F_{1}-F_{2} K_{\mathrm{R}}\right)  \tag{12.3.23}\\
& C_{c}=F_{2}  \tag{12.3.24}\\
& D_{c}=F_{1}-F_{2} K_{\mathrm{R}} \tag{12.3.25}
\end{align*}
$$

where $K_{\mathrm{R}}$ is the reduced order observer gain matrix for the reduced order system $\left(A_{\mathrm{R}}, B_{\mathrm{R}}, C_{\mathrm{R}}, D_{\mathrm{R}}\right)$, and is chosen such that $A_{\mathrm{R}}+K_{\mathrm{R}} C_{\mathrm{R}}$ is asymptotically stable and the properties associated the state feedback law is recovered. Once again, using the $m$-function atea. $m$ in the toolbox of [12] and after a few iterations and simulations, we found that the following reduced order observer gain matrix $K_{\mathrm{R}}$,

$$
K_{\mathrm{R}}=-\left[\begin{array}{rrrr}
85.4439 & 21.2201 & 0 & 0  \tag{12.3.26}\\
21.2201 & 122.3176 & 0 & 0
\end{array}\right]
$$

will yield a good performance. Finally, substituting this $K_{\mathrm{R}}$ into equations (12.3.22)-(12.3.25), we have

$$
\begin{align*}
A_{c} & =\left[\begin{array}{rr}
-174.3280 & -74.2370 \\
106.2743 & -332.7939
\end{array}\right],  \tag{12.3.27}\\
B_{c} & =\left[\begin{array}{rrrr}
-83.3798 & -64.5160 & 1.0269 & 0.6172 \\
11.5772 & -194.7265 & -1.4843 & 2.4695
\end{array}\right]  \tag{12.3.28}\\
C_{c} & =\left[\begin{array}{rrr}
-205.4112 & 0 \\
0 & -202.2678
\end{array}\right]  \tag{12.3.29}\\
D_{c} & =\left[\begin{array}{rrrr}
-90.1288 & -23.2207 & 2.3732 & 1.4264 \\
-20.0343 & -126.0777 & -1.4264 & 2.3732
\end{array}\right] \tag{12.3.30}
\end{align*}
$$

### 12.4. Simulation and Implementation Results

In order to implement our controller designed in the previous section using our hardware setup, we need to discretize it. The performance of this discretized controller is then evaluated using MATLAB SIMULINK. Finally, it is applied to the actual free gyro-stabilized mirror system. Using the well-known bilinear transformation (see also Chapter 4) with a sampling time of 4 ms , we obtained the following discretized controller,

$$
\Sigma_{d}:\left\{\begin{align*}
v(k+1) & =A_{d} v(k)+B_{d} y(k),  \tag{12.4.1}\\
u(k) & =C_{d} v(k)+D_{d} y(k),
\end{align*}\right.
$$

where

$$
\begin{align*}
A_{d} & =\left[\begin{array}{rr}
0.4624 & -0.1304 \\
0.1866 & 0.1841
\end{array}\right]  \tag{12.4.2}\\
B_{d} & =\left[\begin{array}{rrrr}
-61.7225 & -34.4820 & 0.8476 & 0.2904 \\
-0.9257 & -121.3119 & -0.7830 & 1.5197
\end{array}\right]  \tag{12.4.3}\\
C_{d} & =\left[\begin{array}{rr}
-0.6008 & 0.0536 \\
-0.0755 & -0.4790
\end{array}\right],  \tag{12.4.4}\\
D_{d} & =\left[\begin{array}{rrrr}
-64.7719 & -9.0547 & 2.0249 & 1.3072 \\
-19.6598 & -77.0027 & -1.1097 & 1.7584
\end{array}\right] \tag{12.4.5}
\end{align*}
$$

The SIMULINK simulation block diagram for the the free gyro-stabilized mirror system is given in Figure 12.4.1. In order to achieve more accurate results, the nonlinear model given in equations (12.2.1)-(12.2.2) is used in the gyro block. Simulations are carried out using the Runge-Kutta 5 method with both


Figure 12.4.1: Simulation block patched up in SIMULINK.
minimum and maximum step sizes set to be the same as the sampling period, i.e., 4 ms . To account for the limitations in the torque motors, a saturation block is added to each of them. The limits are set to be $\pm 0.5 \mathrm{Nm}$. Throughout the simulations, the gyro's spin velocity is set to be 2500 rpm .

The gyro is first commanded to move simultaneously to (yaw, pitch) = $\left(5^{\circ},-5^{\circ}\right)$. On the fifth seconds, it is moved from this new position to $\left(20^{\circ},-20^{\circ}\right)$. A horizontal span is then carried out, i.e., the gyro is moved horizontally from $20^{\circ}$ to $-5^{\circ}$ while keeping the pitch position at $-20^{\circ}$. This is followed by a vertical span; this time the yaw position is fixed at $-5^{\circ}$ while the pitch position is changed from $-20^{\circ}$ to $5^{\circ}$. Finally, it is pushed to its extreme position $\left(-50^{\circ}, 30^{\circ}\right)$ before returning back to its zero position. The gyro's response as well as the torque input to each axis are plotted in Figures 12.4.2-12.4.3.

The various set-points in the above tests are chosen such that from one position to another, the displacement ranges from as small as $5^{\circ}$ up to $45^{\circ}$. This is to verify that our controller works well within the whole workspace although it is designed based on a linearized model. The simultaneous movement is to test whether our controller is capable of achieving perfect tracking in both axes while


Figure 12.4.2: Simulation result: Responses of $\theta_{1}$ and $\theta_{2}$.


Figure 12.4.3: Simulation result: Control inputs $u_{1}$ and $u_{2}$.
the spans are conducted to investigate how well does our controller 'decouple' the gyro-stabilized mirror system. As can be seen from the responses in Figure 12.4.2, the gyro is able to reach all commanded positions without steady state errors. Furthermore, none of the responses exhibits any overshoots. The settling time from its extreme position back to the zero position is about 3.5 seconds. The maximum coupled movement in $\theta_{1}$ caused by moving $\theta_{2}$ is around $0.15^{\circ}$. The maximum coupled movement in $\theta_{2}$ caused by moving $\theta_{1}$ is about $0.5^{\circ}$. A check with Figure 12.4.3 shows that all these are accomplished with the torques kept within the constrain of $\pm 0.5 \mathrm{Nm}$. Thus we conclude that our controller designed in the previous section is very satisfactory.

Next, we implement this controller on the actual free gyro-stabilized mirror system via a computer (see Figure 12.2.1) and perform the whole test once again. The results obtained are shown in Figures 12.4.4-12.4.5.

Comparing Figures 12.4.2-12.4.3 with Figures 12.4.4-12.4.5, we note that the general waveforms are the same. However, there exist steady state errors in both axes. Furthermore, the real system takes a slightly longer time before settling at its set-point. For example, it now takes about 5 seconds instead of 3.5 seconds to move from its extreme position back to zero. The coupled interaction caused by movement in the other axis is also larger than our simulation results ( $1.6^{\circ}$ in the yaw axis and $0.55^{\circ}$ in the pitch axis). The performance of the controller during the implementation is clearly not as good as in the simulation. The reason is due to the imperfection of the hardware system.

The biggest defect that the system has may be the dead zones of the torque motors. Studying Figure 12.4.5, we observe that although the torques are still non-zero, the positions have already reached their steady states. This can only happen if the torque motors are working within their dead zones. In fact, after running a few tests, we find that the dead zone in the pitch motor is more pronounce and it does not remain constant throughout operation. According to one past documentation (see e.g., [56]), the dead zone is related to the mechanical vibration on the gyro-mirror. In situations when the gyro-mirror vibrates, the vibrations cause the system to 'loosen up' and result in a small dead zone; at other times when the gyro-mirror is stabilized and spinning smoothly, a large dead zone exists. This behaviour makes the dead zone compensation extremely difficult. Nevertheless through trial and error, we observe that the magnitude of the dead zone compensation seems to be related to the set-points in the following way:

$$
\begin{equation*}
u_{o s 1}=\alpha_{1} r_{1}+\beta_{1} r_{2} \tag{12.4.6}
\end{equation*}
$$


(a) Response for $\theta_{1}$.

(b) Response for $\theta_{2}$.

Figure 12.4.4: Implementation result: Responses of $\theta_{1}$ and $\theta_{2}$.


Figure 12.4.5: Implementation result: Control inputs $u_{1}$ and $u_{2}$.
and

$$
\begin{equation*}
u_{o s 2}=\alpha_{2} r_{1}+\beta_{2} r_{2} \tag{12.4.7}
\end{equation*}
$$

where $u_{o s 1}$ and $u_{o s 2}$ are the values to be added to $u_{1}$ and $u_{2}$, respectively. Various sets of $\left(r_{1}, r_{2}\right)$ are used to tune $\alpha_{1}, \alpha_{2}, \beta_{1}$ and $\beta_{2}$ so as to obtain suitable offsets to be added to the control inputs such that the dead zone effects can be minimized. Figures 12.4.6-12.4.7 are the results we obtain from our controller with a dead zone compensation whose parameters are chosen as follows:

$$
\begin{equation*}
\alpha_{1}=-0.001125, \quad \beta_{1}=-0.000125, \quad \alpha_{2}=-0.0049875, \quad \beta_{2}=-0.00059375 \tag{12.4.8}
\end{equation*}
$$

With these results, we once again show that our controller is able to perform fast tracking without overshoots in the both axes and minimize the coupled effect ( $0.8^{\circ}$ in the yaw axis and $0.5^{\circ}$ in the pitch axis).

In order to test the robustness of this controller, we send a command to move the gyro simultaneously in the yaw $\left(+20^{\circ}\right)$ and pitch $\left(-20^{\circ}\right)$ direction. Then we purposely introduce some disturbance (through knocking on the gimbals) to the system. As shown in Figure 12.4.8, our controller is robust to this external disturbance.

During implementation, the gyro's spin velocity is controlled via a potentiometer. Hence it is very difficult to set an exact speed of rotation. To make things worse, the gyro will vary it's spinning velocity by itself. Since the free gyro-stabilized mirror system dynamics are dependent on its spin velocity (see equations (12.2.1)-(12.2.2)), the system dynamics is changed too. Furthermore, the physical constants $\bar{a}, \bar{b}, \bar{c}, \bar{d}, \bar{e}, \bar{f}, \bar{g}, \bar{\ell}$ and $\bar{k}$ were obtained from experiments conducted on the free gyro-stabilized mirror system a few years back. Over these years, the free gyro-stabilized mirror system has broken down and has been serviced for many times. Thus, these values may not be accurate anymore. Yet in view of these model uncertainties, the performance of our controller remains very satisfactory.


Figure 12.4.6: Implementation with dead zone compensation: $\theta_{1}$ and $\theta_{2}$.


Figure 12.4.7: Implementation with dead zone compensation: $u_{1}$ and $u_{2}$.


Figure 12.4.8: Robustness test for the final system.

## Chapter 13

## An Open Problem

Of COURSE, THERE are still quite a number of problems associated with $H_{\infty}$ control remaining unsolved in the literature. We conclude this work by posting an open problem related to the exact computation of the infimum, $\gamma^{*}$. For simplicity, we will only focus on the exact computation of $\gamma^{*}$ for the continuoustime full information feedback problem, i.e., we consider

$$
\left\{\begin{array}{l}
\left.\dot{x}=A+B u+\begin{array}{c}
E \\
y=\binom{I}{0} x \\
h= \\
I
\end{array}\right) w,  \tag{13.0.1}\\
I
\end{array}\right.
$$

As usual, we let $\Sigma_{\mathrm{P}}$ denote the quadruple $\left(A, B, C_{2}, D_{2}\right)$. The algorithm that yields the exact value of $\gamma^{*}$ for this type of problem in Chapter 5 was built based on the following crucial assumption,

$$
\begin{equation*}
\operatorname{Im}(E) \subset \mathcal{V}^{-}\left(\Sigma_{\mathrm{P}}\right)+\mathcal{S}^{-}\left(\Sigma_{\mathrm{P}}\right) \tag{13.0.2}
\end{equation*}
$$

and some minor ones. As will be seen shortly in an example, the assumption of (13.0.2) is not a necessary condition for obtaining the exact value of $\gamma^{*}$. Here is the open problem.

Open Problem. How to compute the exact value of the infimum, i.e., $\gamma^{*}$, associated with the full information feedback system of (13.0.1) without posing the condition as given in (13.0.2)?

We believe that the above problem is solvable or at least partially solvable. The following is an example for which we are able to obtain the exact value of $\gamma^{*}$ without posing the condition of (13.0.2).

Example 13.0.1. Consider a full information feedback system of (13.0.1) with

$$
A=\left[\begin{array}{ll}
1 & 0  \tag{13.0.3}\\
0 & 1
\end{array}\right], \quad B=\left[\begin{array}{ll}
1 & 0 \\
0 & 1
\end{array}\right], \quad E=\left[\begin{array}{ll}
1 & 0 \\
0 & 1
\end{array}\right],
$$

and

$$
C_{2}=\left[\begin{array}{ll}
0 & 0  \tag{13.0.4}\\
0 & 0 \\
1 & 0 \\
0 & 2
\end{array}\right], \quad D_{2}=\left[\begin{array}{ll}
1 & 0 \\
0 & 1 \\
0 & 0 \\
0 & 0
\end{array}\right], \quad D_{22}=0
$$

It is simple to check using the linear system tools of Chapter 2 that

$$
\begin{equation*}
\mathcal{V}^{-}\left(\Sigma_{\mathrm{P}}\right)+\mathcal{S}^{-}\left(\Sigma_{\mathrm{P}}\right)=\{0\} \tag{13.0.5}
\end{equation*}
$$

and hence the condition (13.0.2) is not valid. It is also straightforward to verify that the existence of a $\gamma$-suboptimal control law with $\gamma>\gamma^{*} \geq 0$ for (13.0.1) is equivalent to the existence of a positive definite solution $P$ for the following algebraic Riccati equation,

$$
\begin{equation*}
P A+A^{\prime} P+P E E^{\prime} P / \gamma^{2}-P B B^{\prime} P+C_{2}^{\prime} C_{2}=0 \tag{13.0.6}
\end{equation*}
$$

Let

$$
P:=\left[\begin{array}{ll}
P_{1} & P_{0}  \tag{13.0.7}\\
P_{0} & P_{2}
\end{array}\right] \quad \text { and } \quad \frac{1}{\alpha}:=\frac{1}{\gamma^{2}}-1 .
$$

Then (13.0.6) is equivalent to

$$
\left[\begin{array}{cc}
P_{0}^{2}+P_{1}^{2}+2 \alpha P_{1}+\alpha & P_{0}\left(P_{1}+P_{2}+2 \alpha\right)  \tag{13.0.8}\\
P_{0}\left(P_{1}+P_{2}+2 \alpha\right) & P_{0}^{2}+P_{2}^{2}+2 \alpha P_{2}+4 \alpha
\end{array}\right]=0
$$

or

$$
\begin{align*}
P_{0}\left(P_{1}+P_{2}+2 \alpha\right) & =0,  \tag{13.0.9}\\
P_{0}^{2}+P_{1}^{2}+2 \alpha P_{1}+\alpha & =0,  \tag{13.0.10}\\
P_{0}^{2}+P_{2}^{2}+2 \alpha P_{2}+4 \alpha & =0 . \tag{13.0.11}
\end{align*}
$$

Equation (13.0.9) implies that either

$$
\begin{equation*}
P_{0}=0 \quad \text { or } \quad P_{1}+P_{2}+2 \alpha=0 \tag{13.0.12}
\end{equation*}
$$

If we choose $P_{1}+P_{2}+2 \alpha=0$, then we have

$$
\begin{equation*}
P_{1}=-P_{2}-2 \alpha \tag{13.0.13}
\end{equation*}
$$

which together with (13.0.10) imply that

$$
\begin{equation*}
P_{0}^{2}+P_{2}^{2}+2 \alpha P_{2}+\alpha=0 \tag{13.0.14}
\end{equation*}
$$

Clearly, (13.0.11) and (13.0.14) imply that $\alpha=0$ or equivalently $\gamma=0$. Note that $\gamma>\gamma^{*} \geq 0$. Hence, it is a contradiction. Thus, we will have to choose $P_{0}=0$. Then (13.0.10) and (13.0.11) are reduced to

$$
\begin{align*}
P_{1}^{2}+2 \alpha P_{1}+\alpha & =0  \tag{13.0.15}\\
P_{2}^{2}+2 \alpha P_{2}+4 \alpha & =0 \tag{13.0.16}
\end{align*}
$$

It can be readily verified that the above equations have positive solutions $P_{1}$ and $P_{2}$ if and only if $\alpha<0$, or equivalently $\gamma>1$. Therefore, the exact value of the infimum is given by $\gamma^{*}=1$. Moreover, the positive definite solution $P$ of (13.0.6) is given by

$$
P=\left[\begin{array}{cc}
\frac{\gamma}{\gamma^{2}-1}\left(\sqrt{2 \gamma^{2}-1}+\gamma\right) & 0  \tag{13.0.17}\\
0 & \frac{\gamma}{\gamma^{2}-1}\left(\sqrt{5 \gamma^{2}-4}+\gamma\right)
\end{array}\right]
$$

for any given $\gamma>\gamma^{*}=1$.

In general, we feel that there is a large class of systems that do not necessarily satisfy the geometric condition (13.0.2) but their infima are exactly computable. It is an interesting and of course very challenging problem.

Finally, we would like to note that most of the algorithms presented in this book have been implemented by the author and/or his co-workers in a Linear Systems and Control Toolbox under the MATLAB environment [12]. The toolbox collects quite a number of m-functions related to linear systems and control theory. Here is a list of some selected $m$-functions from the package:

> * jordan, to compute the Jordan canonical form;
> $\star r_{-}$jordan, to find the real Jordan canonical form;
> * brunovsk, to find the Brunovsky canonical form;
> * bdccf, to find the block diagonal controllability canonical form;
> $\star$ uni_scb, to realize the unified special coordinate basis decomposition;
> $\star \mathrm{v} \_\mathbf{x}$, to find the weakly unobservable geometric subspace $\mathcal{V}^{\mathrm{x}}$;
> $\star \mathrm{s} \_\mathrm{x}$, to find the strongly controllable geometric subspace $\mathcal{S}^{\mathrm{x}}$;
> $\star$ v_lambda, to find the geometric subspace $\mathcal{V}_{\lambda}$;
> * s_lambda, to find the geometric subspace $\mathcal{S}_{\lambda}$;
> $\star$ morseidx, to find the Morse index lists of a given linear system;
> $\star$ intersec, to calculate the intersection of two vector subspaces;

```
* ssorder, to determine the ordering of two vector subspaces;
\star atea, realization of the general ATEA design method;
\star dare, solution to the general discrete-time Riccati equations;
\star h2dare, solution to the H2 discrete-time Riccati equations;
\star h8dare, solution to the }\mp@subsup{H}{\infty}{}\mathrm{ discrete-time Riccati equations;
\star ch2state, to design a continuous-time H2 state feedback law;
\star ch8infmn, to find the infimum }\mp@subsup{\gamma}{}{*}\mathrm{ in continuous-time }\mp@subsup{H}{\infty}{}\mathrm{ optimization;
\star ch8state, to design a continuous-time }\mp@subsup{H}{\infty}{}\mathrm{ state feedback law;
\star ch8finfo, to find a continuous-time }\mp@subsup{H}{\infty}{}\mathrm{ full information feedback law;
\star ch8fout, to find a continuous-time }\mp@subsup{H}{\infty}{}\mathrm{ full order output feedback law;
\star ch8rout, to find a continuous-time }\mp@subsup{H}{\infty}{}\mathrm{ reduced order control law;
\star dh8infmn, to find the infimum }\mp@subsup{\gamma}{}{*}\mathrm{ in discrete-time }\mp@subsup{H}{\infty}{}\mathrm{ optimization;
\star dh8state, to design a discrete-time }\mp@subsup{H}{\infty}{}\mathrm{ state feedback law;
\star dh8finfo, to find a discrete-time }\mp@subsup{H}{\infty}{}\mathrm{ full information feedback law;
\star dh8fout, to find a discrete-time }\mp@subsup{H}{\infty}{}\mathrm{ full order output feedback law;
\star dh8rout, to calculate a discrete-time }\mp@subsup{H}{\infty}{}\mathrm{ reduced order control law.
```

The above list is very incomplete and the author is still implementing some new algorithms. Interested readers can contact the author through email at bmchen@nus.edu.sg, for further details.

## Bibliography

[1] K. J. Åström, P. Hagander and J. Sternby, "Zeros of sampled systems," Automatica, Vol. 20, No. 4, pp. 21-38, 1984.
[2] K. J. Åström and B. Wittenmark, Computer Controlled Systems: Theory and Design, Prentice-Hall, Englewood Cliffs, 1984.
[3] S. Barnett, Matrices in Control Theory, Robert E. Krieger Publishing Company, Malabar, Florida, 1984.
[4] T. Başar and P. Bernhard, $H_{\infty}$ Optimal Control and Related Minimax Design Problems: A Dynamic Game Approach, Birkhäuser, Boston, 1991.
[5] W. J. Bigley and V. J. Rizzo, "Wideband linear quadratic control of a gyro-stabilized electro-optical sight system," IEEE Control Systems Magazine, Vol. 7, No. 4, pp. 20-24, 1987.
[6] R. C. Booton, Jr., "Nonlinear control systems with random inputs," IRE Transactions on Circuit Theory, CT-1, pp. 9-18, 1954.
[7] S. W. Chan, G. C. Goodwin and K. S. Sin, "Convergence properties of the Riccati difference equation in optimal filtering of nonstabilizable systems," IEEE Transactions on Automatic Control, Vol. 29, No. 1, pp. 110-118, 1984.
[8] T. P. Chang, Seismic Response Analysis of Nonlinear Structures Using the Stochastic Equivalent Linearization Technique, Ph.D. Dissertation, Columbia University, 1985.
[9] B. M. Chen, Software Manual for the Special Coordinate Basis of Multivariable Linear Systems, Washington State University Technical Report Number: ECE 0094, Pullman, Washington, 1988.
[10] B. M. Chen, Theory of Loop Transfer Recovery for Multivariable Linear Systems, Ph.D. Dissertation, Washington State University, 1991.
[11] B. M. Chen, "A simple algorithm for the stable/unstable decomposition of a linear discrete-time system," International Journal of Control, Vol. 61, pp. 255-260, 1995.
[12] B. M. Chen, Linear Systems and Control Toolbox, Technical Report, Department of Electrical Engineering, National University of Singapore, Singapore, 1997.
[13] B. M. Chen, "Exact computation of infimum in discrete-time $H_{\infty^{-}}$ optimization using measurement feedback," Proceedings of the IFAC Automatic Control 13th Triennial World Congress, San Francisco, Volume G: Education and Robust Control I, pp. 151-156, July 1996.
[14] B. M. Chen, "Exact computation of infimum for a class of continuoustime $H_{\infty}$ optimal control problem with a nonzero direct feedthrough term from the disturbance input to the controlled output," Systems \& Control Letters, Vol. 32, No. 2, pp. 99-109, 1997.
[15] B. M. Chen, "Solvability conditions for the disturbance decoupling problems with static measurement feedback," International Journal of Control, Vol. 68, No. 1, pp. 51-60, 1997.
[16] B. M. Chen and Y.-L. Chen, "Loop transfer recovery design via new observer based and CSS architecture based controllers," International Journal of Robust and Nonlinear Control, Vol. 5, No. 7, pp. 649-669, 1995.
[17] B. M. Chen, Y. Guo and Z. L. Lin, "Non-iterative computation of infimum in discrete-time $H_{\infty}$-optimisation and solvability conditions for the discrete-time disturbance decoupling problem," International Journal of Control, Vol. 65, No. 3, pp. 433-454, 1996.
[18] B. M. Chen, J. He and Y.-L. Chen, "Explicit solvability conditions for general discrete-time $H_{\infty}$ almost disturbance decoupling problem with internal stability," Submitted for publication, 1997.
[19] B. M. Chen and A. Saberi, "Noniterative computation of infimum in $H_{\infty}$-optimisation for plants with invariant zeros on the $j \omega$ axis," IEE Proceedings-Part D: Control Theory and Applications, Vol. 140, No. 5, pp. 298-304, 1993.
[20] B. M. Chen, Z. Lin and C. C. Hang, "Solutions to general $H_{\infty}$ almost disturbance decoupling problem with measurement feedback and internal stability - An eigenstructurre assignment approach," To be presented at the 1998 American Control Conference. Also, submitted for journal publication.
[21] B. M. Chen, T. H. Lee, C. C. Hang, Y. Guo and S. Weerasooriya, "An $H_{\infty}$ almost disturbance decoupling robust controller design for a piezoelectric bimorph actuator with hysteresis," To appear in IEEE Transactions on Control Systems Technology, 1998.
[22] B. M. Chen, A. Saberi, S. Bingulac and P. Sannuti, "Loop transfer recovery for non-strictly proper plants," Control-Theory and Advanced Technology, Vol. 6, No. 4, pp. 573-594, 1990.
[23] B. M. Chen, A. Saberi and U. Ly, "Exact computation of the infimum in $H_{\infty}$-optimization via state feedback," Control-Theory and Advanced Technology, Vol. 8, No. 1, pp. 17-35, 1992.
[24] B. M. Chen, A. Saberi and U. Ly, "Exact computation of the infimum in $H_{\infty}$-optimization via output feedback," IEEE Transactions on $A u$ tomatic Control, Vol. 37, No. 1, pp. 70-78, 1992.
[25] B. M. Chen, A. Saberi and U. Ly, "A non-iterative method for computing the infimum in $H_{\infty}$-optimization," International Journal of Control, Vol. 56, No. 6, pp. 1399-1418, 1992.
[26] B. M. Chen, A. Saberi and U. Ly, "Closed loop transfer recovery with observer based controllers - Part 1: Analysis \& Part 2: Design," Control and Dynamic Systems: Advances in Theory and Applications, Vol. 51, No. 2, pp. 247-348, 1992.
[27] B. M. Chen, A. Saberi and P. Sannuti, "Explicit expressions for cascade factorization of general non-minimum phase systems," IEEE Transactions on Automatic Control, Vol. 37, No. 3, pp. 358-363, 1992.
[28] B. M. Chen, A. Saberi and P. Sannuti, "On blocking zeros and strong stabilizability of linear multivariable systems," Automatica, Vol. 28, No. 5, pp. 1051-1055, 1992.
[29] B. M. Chen, A. Saberi, P. Sannuti and Y. Shamash, "Construction and parameterization of all static and dynamic $H_{2}$-optimal state feedback solutions, optimal fixed modes and fixed decoupling zeros," IEEE Transactions on Automatic Control, Vol. 38, No. 2, pp. 248-261, 1993.
[30] B. M. Chen, A. Saberi and Y. Shamash, "A non-recursive method for solving the general discrete time algebraic Riccati equation related to the $H_{\infty}$ control problem," International Journal of Robust and Nonlinear Control, Vol. 4, No. 4, pp. 503-519, 1994.
[31] B. M. Chen, A. Saberi, Y. Shamash and P. Sannuti, "Construction and parameterization of all static and dynamic $H_{2}$-optimal state feedback solutions for discrete time systems," Automatica, Vol. 30, No. 10, pp. 1617-1624, 1994.
[32] B. M. Chen and S. R. Weller, "Mappings of the finite and infinite zero structures and invertibility structures of general linear multivariable systems under the bilinear transformation," Proceedings of the 2nd Asian Control Conference, Seoul, Korea, Volume II, pp. 139-142, 1997. Also to appear in Automatica.
[33] B. M. Chen and D. Z. Zheng, "Simultaneous finite and infinite zero assignments of linear systems," Automatica, Vol. 31, No. 4, pp. 643648, 1995.
[34] T. K. Caughey, "Derivation and application of the Fokker-Planck equation to discrete nonlinear dynamic systems subjected to white random excitation," Journal of the Acoustical Society of America, Vol. 35, No. 11, pp. 1683-1692, 1963.
[35] C. Commault and J. M. Dion, "Structure at infinity of linear multivariable systems: A geometric approach," IEEE Transactions on Automatic Control, Vol. AC-27, No. 3, pp. 693-696, 1982.
[36] S. T. Crandall, "Perturbation techniques for random vibration of nonlinear systems," Journal of the Acoustical Society of America, Vol. 35, No. 11, pp. 1700-1705, 1963.
[37] J. C. Doyle, Lecture Notes in Advances in Multivariable Control, ONR/Honeywell Workshop, 1984
[38] J. C. Doyle and K. Glover, "State-space formulae for all stabilizing controllers that satisfy an $H_{\infty}$-norm bound and relations to risk sensitivity," Systems \& Control Letters, Vol. 11, pp. 167-172, 1988.
[39] J. Doyle, K. Glover, P. P. Khargonekar and B. A. Francis, "State space solutions to standard $H_{2}$ and $H_{\infty}$ control problems," IEEE Transactions on Automatic Control, Vol. 34, No. 8, pp. 831-847, 1989.
[40] C. Fama and K. Matthews, Linear Algebra IIH, Lecture Notes MP274, Department of Mathematics, The University of Queensland, 1991.
[41] L. S. Fan, H. H. Ottesen, T. C. Reiley and R. W. Wood, "Magnetic recording head positioning at very high track densities using a microactuator based, two stage servo system," IEEE Transaction on Industrial Electronics, pp. 222-233, 1995.
[42] B. A. Francis, A Course in $H_{\infty}$ Control Theory, Lecture Notes in Control and Information Sciences, Vol. 88, Springer-Verlag, Berlin, 1987.
[43] G. F. Franklin, J. D. Powell and M. L. Workman, Digital Control of Dynamic Systems, Addison-Wesley, Reading, Massachusetts, 1990.
[44] S. S. Ge, T. H. Lee and Q. Zhao, "Real-time neural network control of a free gyro-stabilized mirror," Proceedings of the 1997 American Control Conference, Albuquerque, New Mexico, pp. 1076-1080, 1997.
[45] K. Glover, "All optimal Hankel-norm approximations of linear multivariable systems and their $\mathcal{L}_{\infty}$ error bounds," International Journal of Control, Vol. 39, pp. 1115-1193, 1984.
[46] J. W. Grizzle and M. H. Shor, "Sampling, infinite zeros and decoupling of linear systems," Automatica, Vol. 24, No. 3, pp. 387-396, 1988.
[47] P. A. Iglesias and K. Glover, "State space approach to discrete time $H_{\infty}$ control," International Journal of Control, Vol. 54, No. 5, pp. 1031-1073, 1991.
[48] T. Kailath, Linear Systems, Prentice-Hall, Englewood Cliffs, 1980.
[49] P. Khargonekar, I. R. Petersen and M. A. Rotea, " $H_{\infty}$-optimal control with state feedback," IEEE Transactions on Automatic Control, Vol. AC-33, No. 8, pp. 786-788, 1988.
[50] H. Kimura, "Conjugation, interpolation and model-matching in $H_{\infty}$," International Journal Control, Vol. 49, pp. 269-307, 1989.
[51] V. Kučera, "The discrete Riccati equation of optimal control, Kybernetika, Vol. 8, No. 3, pp. 430-447, 1972.
[52] H. Kwakernaak, "A polynomial approach to minmax frequency domain optimization of multivariable feedback systems," International Journal of Control, Vol. 41, pp. 117-156, 1986.
[53] H. Kwakernaak and R. Sivan, Linear Optimal Control Systems, John Wiley, New York, 1972.
[54] P. Lancaster, A. C. M. Ran and L. Rodman, "State space approach to discrete time $H_{\infty}$ control," International Journal of Control, Vol. 44, pp. 777-802, 1986.
[55] A. J. Laub, "A Schur method for solving algebraic Riccati equations," IEEE Transactions on Automatic Control, Vol. 24, pp. 913-921, 1979.
[56] M. W. Lee, An Investigation in Fuzzy Logic, Bachelor of Engineering Thesis, Department of Electrical Engineering, National University of Singapore, 1995.
[57] T. H. Lee, E. K. Koh and M. K. Loh, "Stable adaptive control of multivariable servomechanisms, with application to a passive line-of-sight stabilization system," IEEE Transactions on Industrial Electronics, Vol. 43, No. 1, pp. 98-105, 1996.
[58] D. J. N. Limebeer and B. D. O. Anderson, "An interpolation theory approach to $H_{\infty}$ controller degree bounds," Linear Algebra and its Applications, Vol. 98, pp. 347-386, 1988.
[59] D. J. N. Limebeer, M. Green and D. Walker, "Discrete time $H_{\infty}$ control," Proceedings of IEEE Conference on Decision and Control, Tampa, FL, pp. 392-396, 1989.
[60] Z. Lin, The Implementation of Special Coordinate Basis for Linear Multivariable Systems in Matlab, Washington State University Technical Report Number ECE0100, Pullman, Washington, 1989.
[61] Z. Lin, Global and Semi-global Control Problems for Linear Systems Subject to Input Saturation and Minimum-Phase Input-Output Linearizable Systems, Ph.D. Dissertation, Washington State University, 1994.
[62] Z. Lin, "Almost disturbance decoupling with global asymptotic stability for nonlinear systems with disturbance affected unstable zero dynamics," Submitted to Systems \& Control Letters.
[63] Z. Lin, X. Bao and B. M. Chen, "Further results on almost disturbance decoupling with global asymptotic stability for nonlinear systems," Presented at the 36th IEEE Conference on Decision and Control, San Diego, California, 1997.
[64] Z. Lin, B. M. Chen, A. Saberi and Y. Shamash, "Input-output factorization of discrete-time transfer matrices," IEEE Transactions on Circuits and Systems - I: Fundamental Theory and Applications, Vol. 43, No. 11, pp. 941-945, 1996.
[65] Z. Lin and B. M. Chen, "Solutions to general $H_{\infty}$ almost disturbance decoupling problem with measurement feedback and internal stability for discrete-time systems," Submitted for publication, 1997.
[66] T. S. Low and W. Guo, "Modeling of a three-layer piezoelectric bimorph beam with hysteresis," Journal of Microelectromechanical Systems, Vol. 4, No. 4, pp. 230-237, 1995.
[67] R. H. Lyon, "Response of a nonlinear string to random excitation," Journal of the Acoustical Society of America, Vol. 32, No. 8, pp. 953960, 1960.
[68] A. G. J. MacFarlane and N. Karcanias, "Poles and zeros of linear multivariable systems: A survey of the algebraic, geometric and complex variable theory," International Journal of Control, Vol. 24, pp. 33-74, 1976.
[69] D. K. Miu and Y. C. Tai, "Silicon micromachined SCALED technology," IEEE Transaction on Industrial Electronics, pp. 234-239, 1995.
[70] A. S. Morse, "Structural invariants of linear multivariable systems," SIAM Journal on Control, Vol. 11, pp. 446-465, 1973.
[71] P. Moylan, "Stable inversion of linear systems," IEEE Transactions on Automatic Control, Vol. 22, pp. 74-78, 1977.
[72] W. K. Ng, Design Considerations of a Gyro-stabilized Mirror System, Bachelor of Engineering Thesis, Department of Electrical Engineering, National University of Singapore, 1986.
[73] D. H. Owens, "Invariant zeros of multivariable systems: A geometric analysis," International Journal of Control, Vol. 28, pp. 187-198, 1978.
[74] H. K. Ozcetin, A. Saberi and P. Sannuti, "Design for $H_{\infty}$ almost disturbance decoupling problem with internal stability via state or measurement feedback - singular perturbation approach," International Journal of Control, Vol. 55, No. 4, pp. 901-944, 1993.
[75] H. K. Ozcetin, A. Saberi and Y. Shamash, " $H_{\infty}$-almost disturbance decoupling for non-strictly proper systems-A singular perturbation approach," Control-Theory \& Advanced Technology, Vol. 9, pp. 203-245, 1993.
[76] G. P. Papavassilopoulos and M. G. Safonov, "Robust control design via game theoretic methods," Proceedings of the 28th Conference on Decision and Control, Tampa, Florida, pp. 382-387, 1989.
[77] T. Pappas, A. J. Laub and N. R. Sandell, Jr., "On the numerical solution of the discrete-time algebraic Riccati equation," IEEE Transactions on Automatic Control, Vol. 25, No. 4, pp. 631-641, 1980.
[78] J. Perry, Spinning Tops and Gyroscopic Motion, Dover Publications, New York, 1957.
[79] I. R. Petersen, "Disturbance attenuation and $H_{\infty}$-optimization: A design method based on the algebraic Riccati equation," IEEE Transactions on Automatic Control, Vol. AC-32, No. 5, pp. 427-429, 1987.
[80] I. Postlewaite, C. Tsai and D. W. Gu, "Weighting function selection in $H_{\infty}$ design," Proceedings of the 11th Triennial IFAC World Congress, Tallinn, Estonia, pp. 127-132, 1990.
[81] A. C. Pugh and P. A. Ratcliffe, "On the zeros and poles of a rational matrix," International Journal of Control, Vol. 30, pp. 213-227, 1979.
[82] A. C. M. Ran and R. Vreugdenhill, "Existence and comparison theorems for algebraic Riccati equations for continuous- and discrete-time systems," Linear Algebra and its Applications, Vol. 99, pp. 63-83, 1988.
[83] T. J. Richardson and R. H. Kwong, "On positive definite solutions to the algebraic Riccati equation," Systems \& Control Letters, Vol. 7, pp. 99-104, 1986.
[84] H. H. Rosenbrock, State-space and Multivariable Theory, John-Wiley, New York, 1970.
[85] A. Saberi, B. M. Chen and P. Sannuti, "Theory of LTR for nonminimum phase systems, recoverable target loops, recovery in a subspace - Part 1: Analysis and Part 2: Design," International Journal of Control, Vol. 53, No. 5, pp. 1067-1160, 1991.
[86] A. Saberi, B. M. Chen and Z. L. Lin, "Closed-form solutions to a class of $H_{\infty}$-optimization problem," International Journal of Control, Vol. 60, No. 1, pp. 41-70, 1994.
[87] A. Saberi, B. M. Chen and P. Sannuti, Loop Transfer Recovery: Analysis and Design, Springer-Verlag, London, 1993.
[88] A. Saberi, P. Sannuti and B. M. Chen, $\mathrm{H}_{2}$ Optimal Control, PrenticeHall, London, 1995.
[89] A. Saberi and P. Sannuti, "Squaring down of non-strictly proper systems," International Journal of Control, Vol. 51, No. 3, pp. 621-629, 1990.
[90] A. Saberi and P. Sannuti, "Time-scale structure assignment in linear multivariable systems using high-gain feedback," International Journal of Control, Vol. 49, No. 6, pp. 2191-2213, 1989.
[91] A. Saberi and P. Sannuti, "Observer design for loop transfer recovery and for uncertain dynamical systems," IEEE Transactions on Automatic Control, Vol. 35, pp. 878-897, 1990.
[92] M. Sampei, T. Mita and M. Nakamichi, "An algebraic approach to $H_{\infty^{-}}$ output feedback control problem," Systems \& Control Letters, Vol. 14, pp. 13-24, 1990.
[93] P. Sannuti and A. Saberi, "A special coordinate basis of multivariable linear systems - Finite and infinite zero structure, squaring down and decoupling," International Journal of Control, Vol. 45, No. 5, pp. 16551704, 1987.
[94] C. Scherer, " $H_{\infty}$-control by state feedback and fast algorithm for the computation of optimal $H_{\infty}$ norms," IEEE Transactions on Automatic Control, Vol. 35, No. 10, pp. 1090-1099, 1990.
[95] C. Scherer, " $H_{\infty}$-control by state-feedback for plants with zeros on the imaginary axis," SIAM Journal on Control and Optimization, Vol. 30, No. 1, pp. 123-142, 1992.
[96] C. Scherer, " $H_{\infty}$-optimization without assumptions on finite or infinite zeros," SIAM Journal on Control and Optimization, Vol. 30, No. 1, pp. 143-166, 1992.
[97] B. C. Siew, B. M. Chen and T. H. Lee, "Design and implementation of a robust controller for a Free Gyro-stabilized Mirror System," Submitted for publication, 1997.
[98] L. M. Silverman, "Inversion of multivariable linear systems," IEEE Transactions on Automatic Control, Vol. 14, pp. 270-276, 1969.
[99] L. M. Silverman, "Discrete Riccati equations: Alternative algorithms, asymptotic properties, and system theory interpretations," Control and Dynamical Systems, Vol. 12, pp. 313-386, 1976.
[100] A. A. Stoorvogel, The $H_{\infty}$ Control Problem: A State Space Approach, Prentice Hall, Englewood Cliffs, 1992.
[101] A. A. Stoorvogel and J. W. van der Woude, "The disturbance decoupling problem with measurement feedback and stability for systems with direct feedthrough matrices," Systems \& Control Letters, Vol. 17, pp. 217-226, 1991.
[102] A. A. Stoorvogel, A. Saberi and B. M. Chen, "The discrete-time $H_{\infty}$ control problem with measurement feedback," International Journal of Robust and Nonlinear Control, Vol. 4, No. 4, pp. 457-479, 1994.
[103] A. A. Stoorvogel, A. Saberi and B. M. Chen, "A reduced order observer based controller design for $H_{\infty}$-optimization," IEEE Transactions on Automatic Control, Vol. 39, No. 2, pp. 355-360, 1994.
[104] A. A. Stoorvogel and H. L. Trentelman, "The quadratic matrix inequality in singular $H_{\infty}$-control with state feedback," SIAM Journal on Control and Optimization, Vol. 28, No. 5, pp. 1190-1208, 1990.
[105] G. Tadmor, "Worst-case design in the time domain: The maximum principle and the standard $H_{\infty}$ problem," Mathematics of Control, Signals and Systems, Vol. 3, pp. 301-324, 1990.
[106] W. W. Tan, Fuzzy Control of a Free Gyro-stabilized Mirror, Bachelor of Engineering Thesis, Department of Electrical Engineering, National University of Singapore, 1993.
[107] K. M. Tsuchiura, H. H. Tsukuba, H. O. Toride and T. Takahashi, "Disk system with sub-actuators for fine head displacement," US Patent No: 5189578, 1993.
[108] G. Verghese, Infinite Frequency Behavior in Generalized Dynamical Systems, Ph.D. Dissertation, Stanford University, 1978.
[109] S. Weiland and J. C. Willems, "Almost disturbance decoupling with internal stability," IEEE Transactions on Automatic Control, Vol. 34, pp. 277-286, 1989.
[110] H. Wielandt, "On the eigenvalues of $A+B$ and $A B$," Journal of Research of the National Bureau of Standards-B. Mathematical Science, Vol. 778, Nos. $1 \& 2$, pp. 61-63, 1973.
[111] J. C. Willems, "Almost invariant subspaces: An approach to high gain feedback design- Part I: Almost controlled invariant subspaces," IEEE Transactions on Automatic Control, Vol. AC-26, pp. 235-252, 1981.
[112] J. C. Willems, "Almost invariant subspaces: An approach to high gain feedback design- Part II: Almost conditionally invariant subspaces," IEEE Transactions on Automatic Control, Vol. AC-27, pp. 1071-1085, 1982.
[113] W. M. Wonham, Linear Multivariable Control: A Geometric Approach, Springer-Verlag, New York, 1979.
[114] G. Zames, "Feedback and optimal sensitivity: Model reference transformations, multiplicative seminorms, and approximate inverses," IEEE Transactions on Automatic Control, Vol. AC-26, No. 2, pp. 301-320, 1981.
[115] K. Zhou, J. Doyle and K. Glover, Robust \& Optimal Control, PrenticeHall, New York, 1996.
[116] K. Zhou and P. Khargonekar, "An algebraic Riccati equation approach to $H_{\infty}$-optimization," Systems \& Control Letters, Vol. 11, No. 1, pp. 8591, 1988.

## Index

$\gamma$-suboptimal controller, 7
$l_{2}$-norm
continuous signals, 6
discrete signals, 6
$\mathcal{L}_{2}, 167,242$

ATEA design
continuous systems
$H_{\infty}$ control, 133
$H_{\infty}$-ADDPMS, 172
discrete systems
$H_{\infty}$ control, 226
$H_{\infty}$-ADDPMS, 246
gyro mirror system, 318
piezo actuator system, 294

Bilinear transformations, 60, 86, 91, 97, 263
continuous to discrete, 61
controllability, 63, 69
discrete to continuous, 67
geometric subspaces, 65, 71
infinite zero structure, 65, 71
invariant zero structure, 64, 70
invertibility, 64, 70
observability, 63, 69
structural mapping diagram, 73
Block diagonal canonical form, 15, 174

Brunovsky canonical form, 14, 28, 76

Canonical forms
block diagonal, 15
Brunovsky, 14
Jordan, 12
real Jordan, 12
Continuous-time Riccati equations
general CARE, 87
$\mathrm{H}_{2}$-CARE, 97, 104, 205
$H_{\infty}$-CARE, 52, 91
Controllability index, 15, 29

Discrete-time Riccati equations
general DARE, 86
$H_{2}$-DARE, 96
$H_{\infty}$-DARE, 55, 90, 225

Eigenstructure assignment, 133, 172, 226, 246

Finite zero structure (see invariant zeros), 21, 26, 27
Full information case
bilinear transformations, 92
continuous systems
existence conditions, 51,52
infimum $\gamma^{*}, 102,104$
definition, 7
discrete systems
existence conditions, 56
$H_{\infty}$-ADDPMS, 243, 246
infimum $\gamma^{*}, 203,205$
solutions, 224

Full information system, 7
Full order controllers
continuous systems
$H_{\infty}$ control, 140
$H_{\infty}$-ADDPMS, 180
discrete systems
$H_{\infty}$ control, 229
$H_{\infty}$-ADDPMS, 255
Full state feedback system, 7

Geometric subspaces, 29
$\mathcal{S}^{\times}, \mathcal{S}^{*}, \mathcal{S}^{-}, \mathcal{S}^{\odot}, \mathcal{S}^{+}, \mathcal{S}^{\otimes}, 29$
$\mathcal{S}_{\lambda}, 30$
$\mathcal{V}^{\times}, \mathcal{V}^{*}, \mathcal{V}^{-}, \mathcal{V}^{\odot}, \mathcal{V}^{+}, \mathcal{V}^{\otimes}, 29$
$\mathcal{V}_{\lambda}, 30$
bilinear transformation, 65, 71
Gyro mirror system, 309
data acquisition, 315
dead zone compensation, 325
design formulation, 316
design specifications, 316
dynamical model, 313
hardware setup, 310
implementation, 321
interface assembly, 314
introduction, 309
simulation, 321
zero structures, 318
$H_{\infty}$ optimization
$\gamma$-suboptimal controller, 7
existence conditions
continuous systems, 50, 52, 53
discrete systems, 54, 57
full information case, $7,51,52$, 56, 91
infimum, 7, 101, 203, 333
optimal controller, 7
problem statement, 4
regular case, $8,52,57$
singular case, $8,53,54$
solutions
continuous systems, 131
discrete systems, 223
state feedback case, 7, 51, 56
$H_{\infty}$-ADDPMS
continuous systems, 165
applications, 293, 316
solutions, $172,179,182$
solvability conditions, 168,171
discrete systems, 241
solutions, $246,255,258$
solvability conditions, 243,246
$H_{\infty}$-norm
continuous systems, 6
discrete systems, 6
Hamiltonian matrix, 96
Hysteresis, 284
Infimum $\gamma^{*}, 7$
continuous systems, 101
full information, 102, 104
output feedback, 111, 113, 127
discrete systems, 203
full information, 203, 205
output feedback, 215, 217, 221
open problem, 333
Infinite zero structure, 21, 27
bilinear transformation, 65, 71
Invariant zeros, 26
bilinear transformation, 64, 70
multiplicity structure, 27
algebraic multiplicity, 27
geometric multiplicity, 27
Invertibility, 28, 64, 70
degenerate, 28
invertible, 28
left invertible, 28
right invertible, 28
Jordan canonical form, 12, 16, 27
real form, $12,16,189$
Lyapunov equations, 104, 134, 174, 205, 212
Lyapunov functions, 192-194
Morse indices
list $\mathcal{I}_{1}, 27$
list $\mathcal{I}_{2}, 29,64,70$
list $\mathcal{I}_{3}, 29,64,70$
list $\mathcal{I}_{4}, 27$
Normal rank, 26
Piezo actuator system, 283
design formulation, 289
design specifications, 286
dynamical model, 284
hysteretic model, 284
linearization, 287
introduction, 283
simulations, 297
zero structures, 293
Pole-zero cancellations, 138, 142
Quadratic matrix inequalities, 51, $114,147,158$

Reduced order controllers
continuous systems
$H_{\infty}$ control, 146
$H_{\infty}$-ADDPMS, 182
discrete systems
$H_{\infty}$ control, 234
$H_{\infty}$-ADDPMS, 258
gyro mirror design, 321
piezo actuator design, 294

Regular case, 8
continuous systems
existence conditions, 52
solutions, 137, 142
discrete systems
existence conditions, 57
Relative degree, 28
Robust and perfect tracking, 319
Rosenbrock system matrix, 26
Simulink simulation setups
gyro mirror system, 322
piezo actuator system, 302
Singular case, 8
continuous systems
existence conditions, 51
solutions, 131
discrete systems
existence conditions, 54
solutions, 223
Special coordinate basis, $11,20,76$, $103,133,168,204,224$
block diagram, 24
compact form, 23
properties, 25-28, 30
controllability, 25
geometric subspaces, 30
infinite zeros, 27
invariant zeros, 26
invertibility, 28
normal rank, 26
observability, 25
state space decomposition, 29
transformations, 21
Sylvester equations, 121, 124, 126
Tustin's approximation, 61
X-transformer, 11, 36

## Lecture Notes in Control and Information Sciences

## Edited by M. Thoma

## 1993-1998 Published Titles:

Vol. 186: Sreenath, N
Systems Representation of Global Climate Change Models. Foundation for a Systems Science Approach.
288 pp. 1993 [3-540-19824-5]

Vol. 187: Morecki, A.; Bianchi, G.;
Jaworeck, K. (Eds)
RoManSy 9: Proceedings of the Ninth
CISM-IFToMM Symposium on Theory and
Practice of Robots and Manipulators.
476 pp. 1993 [3-540-19834-2]
Vol. 188: Naidu, D. Subbaram
Aeroassisted Orbital Transfer: Guidance and Control Strategies
192 pp. 1993 [3-540-19819-9]

Vol. 189: llchmann, A.
Non-Identifier-Based High-Gain Adaptive Control
220 pp. 1993 [3-540-19845-8]
Vol. 190: Chatila, R.; Hirzinger, G. (Eds)
Experimental Robotics II: The 2nd International Symposium, Toulouse, France, June 25-27 1991
580 pp. 1993 [3-540-19851-2]
Vol. 191: Blondel, V.
Simultaneous Stabilization of Linear Systems
212 pp. 1993 [3-540-19862-8]
Vol. 192: Smith, R.S.; Dahleh, M. (Eds)
The Modeling of Uncertainty in Control Systems
412 pp. 1993 [3-540-19870-9]

Vol. 193: Zinober, A.S.I. (Ed.)
Variable Structure and Lyapunov Control 428 pp. 1993 [3-540-19869-5]

Vol. 194: Cao, Xi-Ren
Realization Probabilities: The Dynamics of Queuing Systems
336 pp. 1993 [3-540-19872-5]
Vol. 195: Liu, D.; Michel, A.N.
Dynamical Systems with Saturation Nonlinearities: Analysis and Design 212 pp. 1994 [3-540-19888-1]

Vol. 196: Battilotti, S.
Noninteracting Control with Stability for Nonlinear Systems
196 pp. 1994 [3-540-19891-1]

Vol. 197: Henry, J.; Yvon, J.P. (Eds)
System Modelling and Optimization
975 pp approx. 1994 [3-540-19893-8]
Vol. 198: Winter, H.; Nüßer, H.-G. (Eds)
Advanced Technologies for Air Traffic Flow Management 225 pp approx. 1994 [3-540-19895-4]

Vol. 199: Cohen, G.; Quadrat, J.-P. (Eds) 11th International Conference on Analysis and Optimization of Systems Discrete Event Systems: Sophia-Antipolis, June 15-16-17, 1994
648 pp. 1994 [3-540-19896-2]
Vol. 200: Yoshikawa, T.; Miyazaki, F. (Eds)
Experimental Robotics III: The 3rd
International Symposium, Kyoto, Japan,
October 28-30, 1993
624 pp. 1994 [3-540-19905-5]

Vol. 201: Kogan, J.
Robust Stability and Convexity
192 pp. 1994 [3-540-19919-5]
Vol. 202: Francis, B.A.; Tannenbaum, A.R. (Eds)
Feedback Control, Nonlinear Systems, and Complexity
288 pp. 1995 [3-540-19943-8]

Vol. 203: Popkov, Y.S.
Macrosystems Theory and its Applications:
Equilibrium Models
344 pp. 1995 [3-540-19955-1]
Vol. 204: Takahashi, S.; Takahara, Y. Logical Approach to Systems Theory 192 pp .1995 [3-540-19956-X]

Vol. 205: Kotta, U. Inversion Method in the Discrete-time Nonlinear Control Systems Synthesis Problems
168 pp .1995 [3-540-19966-7]
Vol. 206: Aganovic, Z.;:Gajic, Z.
Linear Optimal Control of Bilinear Systems
with Applications to Singular Perturbations and Weak Coupling 133 pp .1995 [3-540-19976-4]

Vol. 207: Gabasov, R.; Kirillova, F.M.;
Prischepova, S.V.
Optimal Feedback Control
224 pp .1995 [3-540-19991-8]
Vol. 208: Khalil, H.K.; Chow, J.H.;
loannou, P.A. (Eds)
Proceedings of Workshop on Advances
inControl and its Applications
300 pp .1995 [3-540-19993-4]
Vol. 209: Foias, C.; Özbay, H.;
Tannenbaum, A .
Robust Control of Infinite Dimensional Systems: Frequency Domain Methods 230 pp .1995 [3-540-19994-2]

Vol. 210: De Wilde, P.
Neural Network Models: An Analysis
164 pp. 1996 [3-540-19995-0]
Vol. 211: Gawronski, W. Balanced Control of Flexible Structures 280 pp .1996 [3-540-76017-2]
Vol. 212: Sanchez, A.
Formal Specification and Synthesis of Procedural Controllers for Process Systems 248 pp .1996 [3-540-76021-0]

Vol. 213: Patra, A.; Rao, G.P. General Hybrid Orthogonal Functions and their Applications in Systems and Control 144 pp. 1996 [3-540-76039-3]

Vol. 214: Yin, G.; Zhang, Q. (Eds)
Recent Advances in Control and Optimization of Manufacturing Systems
240 pp. 1996 [3-540-76055-5]
Vol. 215: Bonivento, C.; Marro, G.;
Zanasi, R. (Eds)
Colloquium on Automatic Control
240 pp .1996 [3-540-76060-1]
Vol. 216: Kulhavý, R.
Recursive Nonlinear Estimation: A Geometric
Approach
244 pp. 1996 [3-540-76063-6]
Vol. 217: Garofalo, F.; Glielmo, L. (Eds)
Robust Control via Variable Structure and
Lyapunov Techniques
336 pp. 1996 [3-540-76067-9]
Vol. 218: van der Schaft, A.
$\mathrm{L}_{2}$ Gain and Passivity Techniques in Nonlinear Control
176 pp. 1996 [3-540-76074-1]
Vol. 219: Berger, M.-O.; Deriche, R.;
Herlin, I.; Jaffte, J.; Morel, J.-M. (Eds)
ICAOS '96: 12th Intemational Conference on
Analysis and Optimization of Systems -
Images, Wavelets and PDEs:
Paris, June 26-28 1996
378 pp. 1996 [3-540-76076-8]
Vol. 220: Brogliato, B.
Nonsmooth Impact Mechanics: Models,
Dynamics and Control
420 pp. 1996 [3-540-76079-2]
Vol. 221: Kelkar, A.; Joshi, S.
Control of Nonlinear Multibody Flexible Space
Structures
160 pp. 1996 [3-540-76093-8]
Vol. 222: Morse, A.S.
Control Using Logic-Based Switching
288 pp. 1997 [3-540-76097-0]
Vol. 223: Khatib, O.; Salisbury, J.K.
Experimental Robotics IV: The 4th Intemational Symposium, Stanford, California,
June 30 - July 2, 1995
596 pp. 1997 [3-540-76133-0]

Vol. 224: Magni, J.-F.; Bennani, S.;
Terlouw, J. (Eds)
Robust Flight Control: A Design Challenge 664 pp. 1997 [3-540-76151-9]

Vol. 225: Poznyak, A.S.; Najim, K. Learning Automata and Stochastic Optimization
219 pp. 1997 [3-540-76154-3]
Vol. 226: Cooperman, G.; Michler, G.;
Vinck, H. (Eds)
Workshop on High Performance Computing and Gigabit Local Area Networks
248 pp. 1997 [3-540-76169-1]
Vol. 227: Tarbouriech, S.; Garcia, G. (Eds)
Control of Uncertain Systems with Bounded Inputs
203 pp. 1997 [3-540-76183-7]
Vol. 228: Dugard, L.; Verriest, E.I. (Eds)
Stability and Control of Time-delay Systems 344 pp. 1998 [3-540-76193-4]

Vol. 229: Laumond, J.-P. (Ed.)
Robot Motion Planning and Control
360 pp. 1998 [3-540-76219-1]

Vol. 230: Siciliano, B.; Valavanis, K.P. (Eds) Control Problems in Robotics and Automation 328 pp. 1998 [3-540-76220-5]

Vol. 231: Emel'yanov, S.V.; Burovoi, I.A.; Levada, F.Yu.
Control of Indefinite Nonlinear Dynamic Systems
196 pp. 1998 [3-540-76245-0]
Vol. 232: Casals, A.; de Almeida, A.T.
Experimental Robotics V : The Fifth
International Symposium Barcelona, Catalonia, June 15-18, 1997
190 pp. 1998 [3-540-76218-3]
Vol. 233: Chiacchio, P.; Chiaverini, S.
Complex Robotic Systems
189 pp. 1998 [3-540-76265-5]
Vol. 234: Arena, P, Fortuna, L, Muscato, G and Xibilia, M.G.
Neural Networks in Multidimensional Domains: Fundamentals and New Trends in Modelling and Control
179 pp. 1998 [1-85233-006-6]

