

# $H_\infty$ optimal singular and normal filter design for uncertain singular systems

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**Abstract:** The robust  $H_\infty$  optimal filtering problem for singular systems with norm-bounded uncertainties in all system matrices is considered. On the basis of the admissibility assumption of the uncertain singular systems, one singular and two normal filter design methods are proposed under the linear-matrix inequality framework. A numerical example is provided to illustrate the application of all three proposed methods.

## 1 Introduction

In the past decade, the  $H_\infty$  optimal filtering problem for singular systems has been an important research topic. This is due, not only to the theoretical interest, but also to the relevance of the topic in various engineering applications. Many works, such as Xu *et al.* [1] and Yue and Han [2], consider the filters with the same structure as the concerned singular system in the derivative term. Indeed, this is often a natural and convenient choice. However, it is more suitable for the cases in which the coefficient matrix for the derivative term has no uncertainty. In practical applications, more flexibility will be gained if all system matrices are allowed to contain uncertainties.

In this paper, the robust  $H_\infty$  optimal filtering problem is studied for singular systems, which contain norm-bounded uncertainties in all system matrices. First of all, a filter design method that generates singular filters is proposed as an extension of the method by Xu *et al.* [1]. Then, two kinds of normal filters (i.e. those with the system matrix for the derivative term being the identity matrix) [3] are considered. For singular as well as normal filters, the goal is to minimise the  $H_\infty$  performance level of the corresponding filtering error dynamics. The consideration of normal filters is beneficial when sometimes the physical realisations of singular filters are not easy [3, 4]. In order to realise a singular system, often it needs special algorithms [5] to convert a singular system model into a normal state-space form.

To handle systems with uncertainties in the system matrix for the derivative term, it is assumed that the uncertain systems are admissible and the concept of the restricted system equivalence (r.s.e.) [3] is applied. In addition, some preliminary results in Lin *et al.* [6], which considers the stabilisation problem for singular systems using the algebraic Riccati equation method, provide the further assumptions for the theoretical development of this paper. For easier applications, all proposed filter design methods are formulated within the linear-matrix inequality (LMI) framework.

Some notations to be used subsequently are introduced here. The inequality  $X \geq \theta$  means that the matrix  $X$  is symmetric and positive semi-definite, and  $X \geq Y$  means  $X - Y \geq \theta$ . Similar definitions apply to symmetric positive/negative definite matrices. For a matrix  $M$ ,  $\|M\|$  denotes the spectral norm of  $M$ , and for a stable continuous-time transfer function matrix  $G(s)$ ,  $\|G\|_\infty = \sup_{\omega \in [0, \infty)} \|G(j\omega)\|$  is its  $H_\infty$  norm.  $I_r$  is the identity matrix with dimension  $r$ , the superscript T represents the transpose of a matrix and  $\text{diag}(X, Y, \dots, Z)$  is the block diagonal matrix with diagonal elements  $X, Y, \dots, Z$ . Finally,  $*$  is used to simplify the presentation of symmetric matrices.

## 2 Preliminaries and problem formulation

First, consider the following nominal singular system

$$\Sigma_n: \begin{cases} E\dot{x}(t) = Ax(t) + Bu(t) \\ z(t) = Lx(t) \end{cases} \quad (1)$$

where  $x(t) \in \mathcal{R}^n$  and  $\text{rank} E = r < n$ . The unforced singular system pair  $(E, A)$  of (1) with  $u = \theta$  is regular if  $\det(sE - A)$  is not identically zero. If  $\deg(\det(sE - A)) = \text{rank} E$ , then  $(E, A)$  is said to be impulse free. The pair  $(E, A)$  is stable if all the roots of  $\det(sE - A) = \theta$  have negative real parts. Finally,  $(E, A)$  is admissible if it is regular, impulse free and stable. For  $\Sigma_n$ , its transfer function matrix from  $u(t)$  to  $z(t)$  is  $G(s) = L(sE - A)^{-1}B$ .

*Definition 1 [3]:* Suppose  $\Sigma_n$  in (1) is regular. Let  $P$  and  $Q$  be two  $n \times n$  non-singular matrices and  $E_r = PEQ$ ,  $A_r = PAQ$ ,  $B_r = PB$ ,  $L_r = LQ$ . The system

$$\Sigma_r: \begin{cases} E_r \dot{x}_r(t) = A_r x_r(t) + B_r u(t) \\ z(t) = L_r x_r(t) \end{cases} \quad (2)$$

with  $x_r(t) = Q^{-1}x(t)$  is r.s.e. to  $\Sigma_n$ .

For any given regular  $\Sigma_n$ , there exist [3] non-singular matrices  $P$  and  $Q$  such that

$$E_r = \begin{bmatrix} I_r & \theta \\ \theta & \theta \end{bmatrix}, \quad A_r = \begin{bmatrix} A_1 & A_2 \\ A_3 & A_4 \end{bmatrix}, \quad B_r = \begin{bmatrix} B_1 \\ B_2 \end{bmatrix}, \quad (3)$$

$$L_r = [L_1 \quad L_2]$$

Now consider the following uncertain singular system

$$\Sigma_u: \begin{cases} E\dot{x}(t) = (A + \delta A)x(t) + (B + \delta B)u(t) \\ z(t) = (L + \delta L)x(t) + (J + \delta J)u(t) \end{cases} \quad (4)$$

where  $\delta A$ ,  $\delta B$ ,  $\delta L$  and  $\delta J$  are constant uncertainties.

**Definition 2** [6, 7]: The unforced pair  $(E, A + \delta A)$  in  $\Sigma_u$  is said to be quadratically admissible for all uncertainty  $\delta A$  if there exists a matrix  $X$  such that

$$E^T X = X^T E \geq 0 \quad (5)$$

$$(A + \delta A)^T X + X^T (A + \delta A) < 0 \quad (6)$$

The quadratic admissibility is also called the generalised quadratic stability by Xu and Yang [8] and Xu et al. [9], and it implies [7] the admissibility of a system.

**Definition 3** [7]: Given a  $\mu > 0$ , the system  $\Sigma_u$  is said to be quadratically admissible with disturbance attenuation  $\mu$  for all uncertainties  $\delta A$ ,  $\delta B$ ,  $\delta L$  and  $\delta J$  if there exists a matrix  $X$  such that

$$E^T X = X^T E \geq 0 \quad (7)$$

$$\begin{bmatrix} (A + \delta A)^T X & * & * \\ + X^T (A + \delta A) & & \\ (B + \delta B)^T X & -\mu^2 I & * \\ L + \delta L & J + \delta J & -I \end{bmatrix} < 0 \quad (8)$$

It is known [7] that quadratically admissible with disturbance attenuation  $\mu$  means that the  $H_\infty$  norm of the transfer function matrix from  $u(t)$  to  $z(t)$  is less than  $\mu$ .

## 2.1 Problem formulation

The uncertain singular system to be discussed is

$$\Sigma: \begin{cases} (E_0 + \delta E)\dot{x}(t) = (A_0 + \delta A)x(t) + (B_0 + \delta B)u(t) \\ y(t) = (C_0 + \delta C)x(t) + (D_0 + \delta D)u(t) \\ z(t) = (L_0 + \delta L)x(t) + (J_0 + \delta J)u(t) \end{cases} \quad (9)$$

where  $x(t) \in \mathcal{R}^n$  is the state vector,  $y(t) \in \mathcal{R}^p$  the measured output vector,  $z(t) \in \mathcal{R}^q$  the vector to be estimated and  $u(t) \in \mathcal{R}^m$  the disturbance input vector. The matrices  $E_0$ ,  $A_0$ ,  $B_0$ ,  $C_0$ ,  $D_0$ ,  $L_0$  and  $J_0$  are known real constant matrices with appropriate dimensions. The constant uncertainty matrices satisfy

$$\begin{bmatrix} \delta A & \delta B \\ \delta C & \delta D \\ \delta L & \delta J \end{bmatrix} = \begin{bmatrix} M_x \\ M_y \\ M_z \end{bmatrix} \Delta [N_x \ N_u] \quad (10)$$

$\delta E = M\Delta N$  and  $\Delta^T \Delta \leq I$ . Suppose that the pair  $(E_0 + \delta E, A_0 + \delta A)$  is admissible and  $\text{rank}(E_0 + \delta E) = \text{rank} E_0 = r < n$  for all  $\Delta$  under discussion. Let the non-singular  $P$  and  $Q$  be such that

$$\begin{aligned} PE_0 Q &= \begin{bmatrix} I_r & 0 \\ 0 & 0 \end{bmatrix}, & PA_0 Q &= \begin{bmatrix} A_{r1} & A_{r2} \\ A_{r3} & A_{r4} \end{bmatrix} \\ PM &= \begin{bmatrix} M_1 \\ M_2 \end{bmatrix}, & NQ &= [N_1 \ N_2] \end{aligned} \quad (11)$$

By Lin *et al.* [6], it can be assumed without loss of generality that either  $M_2 = 0$  or  $N_2 = 0$ . Here, it is further assumed that  $\|N_1 M_1\| < 1$ .

To estimate  $z(t)$ , the following filter is adopted

$$\Sigma_f: \begin{cases} E_f \dot{x}_f(t) = A_f x_f(t) + B_f y(t) \\ z_f(t) = C_f x_f(t) + D_f y(t) \end{cases} \quad (12)$$

where  $x_f(t) \in \mathcal{R}^{n_f}$  and  $z_f(t) \in \mathcal{R}^q$ . The matrices  $E_f$ ,  $A_f$ ,  $B_f$ ,  $C_f$  and  $D_f$  are to be determined. From  $\Sigma$  in (9) and  $\Sigma_f$  in (12), the filtering error dynamics may be written as

$$\Sigma_e: \begin{cases} E_e \dot{x}_e(t) = A_e x_e(t) + B_e u(t) \\ e(t) = C_e x_e(t) + D_e u(t) \end{cases} \quad (13)$$

where  $e(t) = z(t) - z_f(t)$ ,  $x_e^T(t) = [x^T(t) \ x_f^T(t)]$  and

$$\begin{aligned} E_e &= \begin{bmatrix} E_0 + \delta E & 0 \\ 0 & E_f \end{bmatrix}, & A_e &= \begin{bmatrix} A_0 + \delta A & 0 \\ B_f(C_0 + \delta C) & A_f \end{bmatrix} \\ B_e &= \begin{bmatrix} B_0 + \delta B \\ B_f(D_0 + \delta D) \end{bmatrix} \\ C_e &= [(L_0 + \delta L) - D_f(C_0 + \delta C) \quad -C_f] \\ D_e &= (J_0 + \delta J) - D_f(D_0 + \delta D) \end{aligned} \quad (14)$$

The purpose here is to design the filter  $\Sigma_f$  such that the pair  $(E_f, A_f)$  is admissible and

$$\sup_{\Delta} \|C_e(sE_e - A_e)^{-1}B_e + D_e\|_\infty < \mu_e \quad (15)$$

for a prescribed  $H_\infty$ -norm bound  $\mu_e > 0$ .

## 2.2 Restricted system equivalence

Under the assumption that the pair  $(E_0 + \delta E, A_0 + \delta A)$  is admissible, there exist [6] non-singular matrices  $P$ ,  $Q$ ,  $P_\Delta$  and  $Q_\Delta$  such that the system  $\Sigma$  in (9) is r.s.e. to the system

$$\tilde{\Sigma}_r: \begin{cases} \tilde{E}_r \dot{\tilde{x}}(t) = \tilde{A}_r \tilde{x}(t) + \tilde{B}_r u(t) \\ y(t) = \tilde{C}_r \tilde{x}(t) + \tilde{D}_r u(t) \\ z(t) = \tilde{L}_r \tilde{x}(t) + \tilde{J}_r u(t) \end{cases} \quad (16)$$

where

$$\begin{aligned} \tilde{E}_r &= E_{0r} = PE_0 Q, & \tilde{A}_r &= P_\Delta(A_{0r} + M_{xr}\Delta N_{xr})Q_\Delta \\ \tilde{B}_r &= P_\Delta(B_{0r} + M_{xr}\Delta N_{ur}), & \tilde{C}_r &= (C_{0r} + M_{yr}\Delta N_{xr})Q_\Delta \\ \tilde{D}_r &= (D_{0r} + M_{yr}\Delta N_{ur}), & \tilde{L}_r &= (L_{0r} + M_{zr}\Delta N_{xr})Q_\Delta \\ \tilde{J}_r &= (J_{0r} + M_{zr}\Delta N_{ur}) \end{aligned} \quad (17)$$

and

$$\begin{aligned} A_{0r} &= PA_0 Q, & B_{0r} &= PB_0, & C_{0r} &= C_0 Q \\ D_{0r} &= D_0, & L_{0r} &= L_0 Q, & J_{0r} &= J_0 \\ M_{xr} &= PM_x, & M_{yr} &= M_y, & M_{zr} &= M_z \\ N_{xr} &= N_x Q, & N_{ur} &= N_u \end{aligned} \quad (18)$$

More specifically, for  $N_2 = 0$  in (11),  $P_\Delta = I_n - M_{0r}\tilde{\Delta}N_{0r}$  and  $Q_\Delta = I_n$ , and for  $M_2 = 0$  in (11),  $P_\Delta = I_n$  and  $Q_\Delta = I_n - M_{0r}\tilde{\Delta}N_{0r}$ , where  $M_{0r} = PM$ ,  $N_{0r} = NQ$  and  $\tilde{\Delta} = \Delta(I + N_1 M_1 \Delta)^{-1}$ .

Thus, the error dynamics  $\tilde{\Sigma}_e$  from  $\tilde{\Sigma}_r$  in (16) and  $\Sigma_f$  in (12) is

$$\tilde{\Sigma}_e: \begin{cases} \tilde{E}_e \dot{\tilde{x}}_e(t) = \tilde{A}_e \tilde{x}_e(t) + \tilde{B}_e u(t) \\ e(t) = \tilde{C}_e \tilde{x}_e(t) + \tilde{D}_e u(t) \end{cases} \quad (19)$$

where  $\tilde{\mathbf{x}}_e^T(t) = [\tilde{\mathbf{x}}^T(t) \quad \mathbf{x}_f^T(t)]$  and

$$\begin{aligned} \tilde{\mathbf{E}}_e &= \begin{bmatrix} \mathbf{E}_{0r} & \mathbf{0} \\ \mathbf{0} & \mathbf{E}_f \end{bmatrix}, \quad \tilde{\mathbf{A}}_e = \begin{bmatrix} \tilde{\mathbf{A}}_r & \mathbf{0} \\ \mathbf{B}_f \tilde{\mathbf{C}}_r & \mathbf{A}_f \end{bmatrix}, \quad \tilde{\mathbf{B}}_e = \begin{bmatrix} \tilde{\mathbf{B}}_r \\ \mathbf{B}_f \tilde{\mathbf{D}}_r \end{bmatrix}, \\ \tilde{\mathbf{C}}_e &= [\tilde{\mathbf{L}}_r - \mathbf{D}_f \tilde{\mathbf{C}}_r \quad -\mathbf{C}_f], \quad \tilde{\mathbf{D}}_e = \tilde{\mathbf{J}}_r - \mathbf{D}_f \tilde{\mathbf{D}}_r \end{aligned} \quad (20)$$

*Lemma 1:* Suppose there exist non-singular matrices  $\mathbf{P}_s$  and  $\mathbf{Q}_s$  such that the regular systems  $\Sigma_{s1}$  and  $\Sigma_{s2}$  are r.s.e., and there exist non-singular matrices  $\mathbf{P}_f$  and  $\mathbf{Q}_f$  such that the filters  $\Sigma_{f1}$  and  $\Sigma_{f2}$  are r.s.e., where for  $i = 1, 2$

$$\Sigma_{si}: \begin{cases} \mathbf{E}_{si} \dot{\mathbf{x}}_{si}(t) = \mathbf{A}_{si} \mathbf{x}_{si}(t) + \mathbf{B}_{si} \mathbf{u}_s(t) \\ \mathbf{y}_s(t) = \mathbf{C}_{si} \mathbf{x}_{si}(t) + \mathbf{D}_s \mathbf{u}_s(t) \\ \mathbf{z}_s(t) = \mathbf{L}_{si} \mathbf{x}_{si}(t) + \mathbf{J}_s \mathbf{u}_s(t) \end{cases} \quad (21)$$

$$\Sigma_{fi}: \begin{cases} \mathbf{E}_{fi} \dot{\hat{\mathbf{x}}}_{fi}(t) = \mathbf{A}_{fi} \hat{\mathbf{x}}_{fi}(t) + \mathbf{B}_{fi} \mathbf{y}_s(t) \\ \hat{\mathbf{z}}_f(t) = \mathbf{C}_{fi} \hat{\mathbf{x}}_{fi}(t) + \mathbf{D}_f \mathbf{y}_s(t) \end{cases} \quad (22)$$

Then, the corresponding error dynamics  $\Sigma_{e1}$  and  $\Sigma_{e2}$  are also r.s.e., where for  $i = 1, 2$

$$\Sigma_{ei}: \begin{cases} \mathbf{E}_{ei} \dot{\hat{\mathbf{x}}}_{ei}(t) = \mathbf{A}_{ei} \hat{\mathbf{x}}_{ei}(t) + \mathbf{B}_{ei} \mathbf{u}_s(t) \\ \hat{\mathbf{z}}(t) = \mathbf{C}_{ei} \hat{\mathbf{x}}_{ei}(t) + \mathbf{D}_e \mathbf{u}_s(t) \end{cases} \quad (23)$$

$\hat{\mathbf{z}}(t) = \mathbf{z}_s(t) - \hat{\mathbf{z}}_f(t)$ ,  $\mathbf{x}_{ei}^T(t) = [\mathbf{x}_{si}^T(t) \quad \mathbf{x}_{fi}^T(t)]$  and

$$\begin{aligned} \mathbf{E}_{ei} &= \begin{bmatrix} \mathbf{E}_{si} & \mathbf{0} \\ \mathbf{0} & \mathbf{E}_{fi} \end{bmatrix}, \quad \mathbf{A}_{ei} = \begin{bmatrix} \mathbf{A}_{si} & \mathbf{0} \\ \mathbf{B}_{fi} \mathbf{C}_{si} & \mathbf{A}_{fi} \end{bmatrix} \\ \mathbf{B}_{ei} &= \begin{bmatrix} \mathbf{B}_{si} \\ \mathbf{B}_{fi} \mathbf{D}_s \end{bmatrix}, \quad \mathbf{C}_{ei} = [\mathbf{L}_{si} - \mathbf{D}_{fi} \mathbf{C}_{si} \quad -\mathbf{C}_{fi}] \\ \mathbf{D}_e &= \mathbf{J}_s - \mathbf{D}_f \mathbf{D}_s \end{aligned} \quad (24)$$

Moreover, the transfer function matrices  $\mathbf{G}_{e1}(s)$  and  $\mathbf{G}_{e2}(s)$  of  $\Sigma_{e1}$  and  $\Sigma_{e2}$ , respectively, are the same.

*Proof:* By the assumptions, there exist non-singular matrices  $\mathbf{P}_s$ ,  $\mathbf{Q}_s$ ,  $\mathbf{P}_f$  and  $\mathbf{Q}_f$  such that

$$\begin{aligned} \mathbf{E}_{\xi 2} &= \mathbf{P}_\xi \mathbf{E}_{\xi 1} \mathbf{Q}_\xi, \quad \mathbf{A}_{\xi 2} = \mathbf{P}_\xi \mathbf{A}_{\xi 1} \mathbf{Q}_\xi, \quad \mathbf{B}_{\xi 2} = \mathbf{P}_\xi \mathbf{B}_{\xi 1}, \\ \mathbf{C}_{\xi 2} &= \mathbf{C}_{\xi 1} \mathbf{Q}_\xi, \quad \mathbf{L}_{\xi 2} = \mathbf{L}_{\xi 1} \mathbf{Q}_\xi \end{aligned} \quad (25)$$

for  $\xi = s$  and  $f$ . It is easy to see that the matrices in (24) satisfy

$$\begin{aligned} \mathbf{E}_{e2} &= \hat{\mathbf{P}}_e \mathbf{E}_{e1} \hat{\mathbf{Q}}_e, \quad \mathbf{A}_{e2} = \hat{\mathbf{P}}_e \mathbf{A}_{e1} \hat{\mathbf{Q}}_e \\ \mathbf{B}_{e2} &= \hat{\mathbf{P}}_e \mathbf{B}_{e1}, \quad \mathbf{C}_{e2} = \mathbf{C}_{e1} \hat{\mathbf{Q}}_e \end{aligned} \quad (26)$$

where  $\hat{\mathbf{P}}_e = \text{diag}(\mathbf{P}_s, \mathbf{P}_f)$  and  $\hat{\mathbf{Q}}_e = \text{diag}(\mathbf{Q}_s, \mathbf{Q}_f)$ . Therefore  $\Sigma_{e1}$  and  $\Sigma_{e2}$  are also r.s.e. By (26)

$$\begin{aligned} \mathbf{G}_{e2}(s) &= \mathbf{C}_{e2}(s\mathbf{E}_{e2} - \mathbf{A}_{e2})^{-1} \mathbf{B}_{e2} + \mathbf{D}_e \\ &= \mathbf{C}_{e1} \hat{\mathbf{Q}}_e (s\hat{\mathbf{P}}_e \mathbf{E}_{e1} \hat{\mathbf{Q}}_e - \hat{\mathbf{P}}_e \mathbf{A}_{e1} \hat{\mathbf{Q}}_e)^{-1} \hat{\mathbf{P}}_e \mathbf{B}_{e1} + \mathbf{D}_e \\ &= \mathbf{C}_{e1}(s\mathbf{E}_{e1} - \mathbf{A}_{e1})^{-1} \mathbf{B}_{e1} + \mathbf{D}_e = \mathbf{G}_{e1}(s) \end{aligned} \quad (27)$$

□

On Lemma 1, the filtering problem formulated in the last section can be reformulated as follows: design the filter  $\Sigma_f$  in (12) such that the pair  $(\mathbf{E}_f, \mathbf{A}_f)$  is admissible and

$$\sup_{\Delta} \|\tilde{\mathbf{C}}_e (s\tilde{\mathbf{E}}_e - \tilde{\mathbf{A}}_e)^{-1} \tilde{\mathbf{B}}_e + \tilde{\mathbf{D}}_e\|_{\infty} < \mu_e \quad (28)$$

### 2.3 Two useful lemmas

*Lemma 2 [10]:* Let  $\mathbf{I} - \tilde{\mathbf{J}}^T \tilde{\mathbf{J}} > \mathbf{0}$  and define the set

$$\mathbf{Y} = \{\Delta(\mathbf{I} - \tilde{\mathbf{J}}\Delta)^{-1}, \quad \Delta^T \Delta \leq \mathbf{I}\}$$

Then,  $\mathbf{Y} = \{\tilde{\mathbf{J}}^T(\mathbf{I} - \tilde{\mathbf{J}}\tilde{\mathbf{J}}^T)^{-1} + \mathbf{\Pi}^T(\mathbf{I} - \tilde{\mathbf{J}}\tilde{\mathbf{J}}^T)^{-1/2}, \mathbf{\Pi}^T \mathbf{\Pi} \leq (\mathbf{I} - \tilde{\mathbf{J}}^T \tilde{\mathbf{J}})^{-1}\}$ .

*Lemma 3 [11]:* Let  $\mathbf{\Omega}$ ,  $\mathbf{H}$ ,  $\mathbf{F}$  and  $\mathbf{R} > \mathbf{0}$  be real matrices with appropriate dimensions, and the matrix  $\mathbf{\Pi}$  satisfies  $\mathbf{\Pi}^T \mathbf{\Pi} \leq \mathbf{R}$ . Then, for all  $\mathbf{\Pi}^T \mathbf{\Pi} \leq \mathbf{R}$ , the matrix inequality

$$\mathbf{\Omega} + \mathbf{H}\mathbf{\Pi}\mathbf{F} + \mathbf{F}^T \mathbf{\Pi}^T \mathbf{H}^T < \mathbf{0}$$

holds if and only if there exists a scalar  $\varepsilon > 0$  such that

$$\begin{bmatrix} \mathbf{\Omega} & \mathbf{H} \\ \mathbf{H}^T & \mathbf{0} \end{bmatrix} + \varepsilon \begin{bmatrix} \mathbf{F}^T \mathbf{R} \mathbf{F} & \mathbf{0} \\ \mathbf{0} & -\mathbf{I} \end{bmatrix} < \mathbf{0}$$

### 3 Robust filter design

Two kinds of robust filters are discussed here: the singular and the normal filters. For the sake of brevity, only the results for the case with  $N_2 = \mathbf{0}$  [6] will be presented.

#### 3.1 Singular filter design

The filter  $\Sigma_f$  is called a singular filter when  $\text{rank } \mathbf{E}_f < n_f$  in (12). In the literature, for example, Xu *et al.* [1] and Yue and Han [2], the choice  $\mathbf{E}_f = \mathbf{E}$  is the most discussed, as it is often dictated by the derivation process. This case is studied here for comparison with the normal filters. Theorem 1 is developed as an extension of Theorem 1 in Xu *et al.* [1], which concerns nominal singular systems.

*Theorem 1:* If and only if there exist feasible solutions  $\rho$ ,  $\varepsilon$ ,  $\mathbf{Y} \in \mathcal{R}^{n \times n}$ ,  $\mathbf{Z} \in \mathcal{R}^{n \times p}$ ,  $\tilde{\mathbf{M}} \in \mathcal{R}^{n \times n}$ ,  $\tilde{\mathbf{N}} \in \mathcal{R}^{q \times n}$ ,  $\mathbf{P}_c \in \mathcal{R}^{n \times n}$  and  $\mathbf{D}_f \in \mathcal{R}^{q \times p}$  to the LMIs

$$\begin{aligned} \mathbf{E}_{0r}^T \mathbf{P}_c &= \mathbf{P}_c^T \mathbf{E}_{0r} \geq \mathbf{0}, \quad \mathbf{E}_{0r}^T \mathbf{Y} = \mathbf{Y}^T \mathbf{E}_{0r} \geq \mathbf{0} \\ \mathbf{E}_{0r}^T (\mathbf{P}_c - \mathbf{Y}) &\geq \mathbf{0} \end{aligned} \quad (29)$$

$$\begin{bmatrix} \mathbf{A}_{0r}^T \mathbf{\Theta}^T \mathbf{Y} + \mathbf{Y}^T \mathbf{\Theta} \mathbf{A}_{0r} & * & * \\ \left( \mathbf{A}_{0r}^T \mathbf{\Theta}^T \mathbf{Y} + \mathbf{P}_c^T \mathbf{\Theta} \mathbf{A}_{0r} \right) & \left( \mathbf{P}_c^T \mathbf{\Theta} \mathbf{A}_{0r} + \mathbf{A}_{0r}^T \mathbf{\Theta}^T \mathbf{P}_c \right) & * \\ + \mathbf{Z} \mathbf{C}_{0r} + \tilde{\mathbf{M}} & + \mathbf{Z} \mathbf{C}_{0r} + \mathbf{C}_{0r}^T \mathbf{Z}^T & * \\ \mathbf{B}_{0r}^T \mathbf{\Theta}^T \mathbf{Y} & \mathbf{B}_{0r}^T \mathbf{\Theta}^T \mathbf{P}_c + \mathbf{D}_{0r}^T \mathbf{Z}^T & * \\ \mathbf{L}_{0r} - \tilde{\mathbf{N}} & \mathbf{L}_{0r} & * \\ \frac{1}{\varepsilon} \tilde{\Xi}_2 \mathbf{N}_{0r} \mathbf{A}_{0r} & \frac{1}{\varepsilon} \tilde{\Xi}_2 \mathbf{N}_{0r} \mathbf{A}_{0r} & * \\ \mathbf{M}_{0r}^T \mathbf{Y} & \mathbf{M}_{0r}^T \mathbf{P}_c & * \\ \mathbf{M}_{xr}^T \mathbf{\Theta}^T \mathbf{Y} & \mathbf{M}_{xr}^T \mathbf{\Theta}^T \mathbf{P}_c + \mathbf{M}_{yr}^T \mathbf{Z}^T & * \\ \rho \mathbf{N}_{xr} & \rho \mathbf{N}_{xr} & * \\ * & * & * \\ * & * & * \\ -\mu_c^2 \mathbf{I} & * & * \\ \mathbf{J}_{0r} - \mathbf{D}_f \mathbf{D}_{0r} & -\mathbf{I} & * \\ \frac{1}{\varepsilon} \tilde{\Xi}_2 \mathbf{N}_{0r} \mathbf{B}_{0r} & \mathbf{0} & \frac{-1}{\varepsilon} \mathbf{I} \\ \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{M}_{zr}^T - \mathbf{M}_{yr}^T \mathbf{D}_f^T & \frac{1}{\varepsilon} \mathbf{M}_{xr}^T \mathbf{N}_{0r}^T \tilde{\Xi}_2^T \\ \rho \mathbf{N}_{ur} & \mathbf{0} & \mathbf{0} \end{bmatrix}$$



Letting

$$\bar{\Omega}_0 = \begin{bmatrix} \bar{A}_{0r}^T \Theta^T Y + Y^T \Theta \bar{A}_{0r} \\ \left( \bar{A}_{0r}^T \Theta^T Y + P_c^T \Theta \bar{A}_{0r} \right) \\ + Z \bar{C}_{0r} + \bar{M} \\ \bar{B}_{0r}^T \Theta^T Y \\ \bar{L}_{0r} - \bar{N} \\ * & * & * \\ \left( P_c^T \Theta \bar{A}_{0r} + \bar{A}_{0r}^T \Theta^T P_c \right) & * & * \\ + Z C_{0r} + \bar{C}_{0r}^T Z^T & * & * \\ \bar{B}_{0r}^T \Theta^T P_c + \bar{D}_{0r}^T Z^T & -\mu_c^2 I & * \\ \bar{L}_{0r} & \bar{J}_{0r} - D_f \bar{D}_{0r} & -I \end{bmatrix}$$

and

$$\bar{H}_\varepsilon^T = [-\bar{\Xi}_2 N_{0r} \bar{A}_{0r} \quad -\bar{\Xi}_2 N_{0r} A_{0r} \quad -\bar{\Xi}_2 N_{0r} \bar{B}_{0r} \quad 0]$$

one sees that (38) with an  $\varepsilon^{-1} > 0$  is equivalent to

$$\bar{\Omega}_0 + \bar{H}_\varepsilon \Pi F_\varepsilon + F_\varepsilon^T \Pi^T \bar{H}_\varepsilon^T < 0 \quad (40)$$

for all  $\Pi^T \Pi \leq \bar{\Xi}_3^{-1}$  by the Schur complement and Lemma 3. With  $P_\Delta = I_n - M_{0r}(-\bar{\Xi}_1 + \Pi^T \bar{\Xi}_2) N_{0r}$ , the inequality (40) may be rewritten as

$$\begin{bmatrix} \bar{A}_{0r}^T P_\Delta^T Y + Y^T P_\Delta \bar{A}_{0r} & * \\ \left( \bar{A}_{0r}^T P_\Delta^T Y + P_c^T P_\Delta \bar{A}_{0r} \right) & \left( P_c^T P_\Delta \bar{A}_{0r} + \bar{A}_{0r}^T P_\Delta^T P_c \right) \\ + Z \bar{C}_{0r} + \bar{M} & + Z C_{0r} + \bar{C}_{0r}^T Z^T \\ \bar{B}_{0r}^T P_\Delta^T Y & \bar{B}_{0r}^T P_\Delta^T P_c + \bar{D}_{0r}^T Z^T \\ \bar{L}_{0r} - \bar{N} & \bar{L}_{0r} \\ * & * \\ * & * \\ -\mu_c^2 I & * \\ \bar{J}_{0r} - D_f \bar{D}_{0r} & -I \end{bmatrix} < 0 \quad (41)$$

Note that  $P_\Delta = I_n - M_{0r} \Delta (I + N_1 M_1 \Delta)^{-1} N_{0r}$  for  $\Delta^T \Delta \leq I$  by Lemma 2 and the assumption  $\|N_1 M_1\| < 1$ . Pre- and post-multiplying (41) by  $\text{diag}(Y^{-T}, I, I, I)$  and  $\text{diag}(Y^{-1}, I, I, I)$ , respectively, show that (41) is equivalent to

$$\begin{bmatrix} P_1^T \bar{A}_e^T P_e P_1 + P_1^T P_e^T \bar{A}_e P_1 & * & * \\ \bar{B}_e^T P_e P_1 & -\mu_c^2 I & * \\ \bar{C}_e P_1 & \bar{D}_e & -I \end{bmatrix} < 0 \quad (42)$$

where  $P_e$  is defined in (34)

$$P_1 = \begin{bmatrix} Y^{-1} & I \\ W & 0 \end{bmatrix}$$

are non-singular, and  $\bar{A}_e$ ,  $\bar{B}_e$ ,  $\bar{C}_e$  and  $\bar{D}_e$  are given in (20) with  $A_f$ ,  $B_f$  and  $C_f$  from (32). Pre- and post-multiplying (42) by  $\text{diag}(P_1^{-T}, I, I)$  and  $\text{diag}(P_1^{-1}, I, I)$ , respectively, one finally obtains

$$\begin{bmatrix} \bar{A}_e^T P_e + P_e^T \bar{A}_e & * & * \\ \bar{B}_e^T P_e & -\mu_c^2 I & * \\ \bar{C}_e & \bar{D}_e & -I \end{bmatrix} < 0 \quad (43)$$

By Definition 3, the conditions in (35) and (43) with the  $P_e$  in (34) imply that  $\tilde{\Sigma}_e$  in (19) is quadratically admissible with disturbance attenuation  $\mu_c$ . The filter  $\tilde{\Sigma}_f$  in (12) is therefore admissible.

(Necessity) As the filtering error dynamics  $\tilde{\Sigma}_e$  in (19) and (20) is quadratically admissible and has a transfer function matrix satisfying (28), there exists a matrix  $X_e$  such that

$$\tilde{E}_e^T X_e = X_e^T \tilde{E}_e \geq 0 \quad (44)$$

$$\begin{bmatrix} \tilde{A}_e^T X_e + X_e^T \tilde{A}_e & * & * \\ \tilde{B}_e^T X_e & -\mu_c^2 I & * \\ \tilde{C}_e & \tilde{D}_e & -I \end{bmatrix} < 0 \quad (45)$$

It is easy to see that (45) implies

$$\tilde{A}_e^T X_e + X_e^T \tilde{A}_e < 0 \quad (46)$$

which implies  $X_e$  is non-singular. Partition  $X_e$  and  $X_e^{-1}$  as

$$X_e = \begin{bmatrix} X_{e1} & X_{e2} \\ U & X_{e3} \end{bmatrix}, \quad X_e^{-1} = \begin{bmatrix} Y_{e1} & Y_{e2} \\ W & Y_{e3} \end{bmatrix} \quad (47)$$

in accordance with the partition of  $\tilde{A}_e$ . The 2–2 block of (46) gives  $A_f^T X_{e3} + X_{e3}^T A_f < 0$ , which implies  $X_{e3}$  is non-singular. By (46)

$$X_e^{-T} \tilde{A}_e^T + \tilde{A}_e X_e^{-1} < 0 \quad (48)$$

The 1–1 block of (48) gives  $\tilde{A}_r Y_{e1} + Y_{e1}^T \tilde{A}_r^T < 0$ , which implies  $Y_{e1}$  is also non-singular. Note that from (44)

$$E_{0r}^T X_{e1} = X_{e1}^T E_{0r} \geq 0, \quad E_f^T X_{e3} = X_{e3}^T E_f \geq 0 \quad (49)$$

and  $X_e^{-T} \tilde{E}_e^T = \tilde{E}_e X_e^{-1} \geq 0$ , whose 1–1 block implies

$$E_{0r}^T Y_{e1}^{-1} = Y_{e1}^{-T} E_{0r} \geq 0 \quad (50)$$

As  $U$  and  $W$  can be chosen to have full-column rank [1], it is possible to define two full-column-rank matrices

$$\hat{T} = \begin{bmatrix} Y_{e1} & I \\ W & 0 \end{bmatrix}, \quad \check{T} = \begin{bmatrix} I & X_{e1} \\ 0 & U \end{bmatrix} \quad (51)$$

so that  $X_e \hat{T} = \check{T}$ . Pre- and post-multiplying (45) by  $\text{diag}(\hat{T}^T, I, I)$  and  $\text{diag}(\hat{T}, I, I)$ , respectively, and then pre- and post-multiplying the resulting inequality by  $\text{diag}(Y_{e1}^{-T}, I, I, I)$  and  $\text{diag}(Y_{e1}^{-1}, I, I, I)$ , respectively. Then, (30) is obtained by Lemmas 2 and 3 when

$$\begin{aligned} P_c &= X_{e1}, \quad Y = Y_{e1}^{-1}, \quad A_f = U^{-T} \bar{M} Y_{e1} W^{-1}, \\ B_f &= U^{-T} Z, \quad C_f = \bar{N} Y_{e1} W^{-1} \end{aligned} \quad (52)$$

are substituted. Also, the first inequality of (49) provides the first inequality of (29) and (50) provides the second inequality of (29). As  $Y_{e1} = (X_{e1} - X_{e2} X_{e3}^{-1} U)^{-1}$ , by (44) and the second inequality of (49), one has

$$E_{0r}^T (X_{e1} - Y_{e1}^{-1}) = E_{0r}^T X_{e2} X_{e3}^{-1} U = U^T E_f X_{e3}^{-1} U \geq 0 \quad (53)$$

which provides the last inequality of (29).  $\square$

*Remark 1:* On the basis of Theorem 1, the following convex optimisation problem may be formulated to find the  $H_\infty$  optimal filter of the form (12) such that (28) is satisfied with the minimal  $\mu_c^2$

$$\min_{\mu_c^2, \rho, \varepsilon^{-1}, Y, Z, \bar{M}, \bar{N}, P_c, D_f} \mu_c^2 \quad (54)$$

subject to the LMIs (29) and (30). Note that by Theorem 1, the choice of the non-singular matrices  $U$ ,  $V$ ,  $W$  and  $\bar{W}$  does not affect the result of the optimal  $\mu_c$ .





for all  $\Pi^T \Pi \leq \Xi_3^{-1}$ . With  $P_\Delta = I_n - M_{0r}(-\Xi_1 + \Pi^T \Xi_2)N_{0r}$ , the inequality (63) may be rewritten as

$$\begin{bmatrix} \begin{pmatrix} P_c^T P_\Delta \bar{A}_{0r} + E_{0r}^T \Phi \bar{C}_{0r} \\ + \bar{A}_{0r}^T P_\Delta^T P_c + \bar{C}_{0r}^T \Phi^T E_{0r} \end{pmatrix} & * & * \\ \Gamma P_\Delta \bar{A}_{0r} + \Phi \bar{C}_{0r} + \hat{Q} E_{0r} & \hat{Q} + \hat{Q}^T & * \\ \bar{B}_{0r}^T P_\Delta^T P_c + \bar{D}_{0r}^T \Phi^T E_{0r} & \bar{B}_{0r}^T P_\Delta^T \Gamma + \bar{D}_{0r}^T \Phi^T & * \\ \bar{L}_{0r} - D_f \bar{C}_{0r} & -\Psi & * \\ * & * & * \\ * & * & * \\ -\mu_c^2 I & * & * \\ \bar{J}_{0r} - D_f \bar{D}_{0r} & -I & * \end{bmatrix} < 0 \quad (64)$$

Note that  $P_\Delta = I_n - M_{0r} \Delta (I + N_1 M_1 \Delta)^{-1} N_{0r}$  for  $\Delta^T \Delta \leq I$  by Lemma 2 and the assumption  $\|N_1 M_1\| < 1$ . Pre- and post-multiplying (64) by  $\text{diag}(I, P^T, I, I)$ , one obtains

$$\begin{bmatrix} \tilde{A}_e^T P_e + P_e^T \tilde{A}_e & * & * \\ \tilde{B}_e^T P_e & -\mu_c^2 I & * \\ \tilde{C}_e & \tilde{D}_e & -I \end{bmatrix} < 0 \quad (65)$$

where  $P_e$  is given in (58) and  $\tilde{A}_e, \tilde{B}_e, \tilde{C}_e$  and  $\tilde{D}_e$  are given in (20) with  $A_f, B_f$  and  $C_f$  from (57). Expressions (59) and (65) together imply that  $\tilde{\Sigma}_e$  in (19) is quadratically admissible with disturbance attenuation  $\mu_c$ . The normal filter  $\Sigma_f$  in (12) thus is stable.  $\square$

*Remark 2:* On the basis of Theorem 2, the following convex optimisation problem may be formulated to find the  $H_\infty$  optimal normal filter of order  $n$  such that (28) is satisfied with the minimal  $\mu_c^2$

$$\min_{\mu_c^2, \rho, \varepsilon^{-1}, P_c, \Phi, \Psi, \Gamma, \hat{Q}, D_f} \mu_c^2 \quad (66)$$

subject to the LMIs (55) and (56).

### 3.3 Normal filter design method II

In this section, a method for designing normal filters of lower order than those obtainable by the first method is developed.

*Theorem 3:* If there exist feasible solutions  $\varepsilon, \rho, P_c \in \mathcal{R}^{n \times n}$ ,  $P_f \in \mathcal{R}^{r \times r}$ ,  $\tilde{\Phi} \in \mathcal{R}^{r \times p}$ ,  $\tilde{Q} \in \mathcal{R}^{r \times r}$ ,  $C_f \in \mathcal{R}^{q \times r}$  and  $D_f \in \mathcal{R}^{q \times p}$  to the LMIs

$$P_f > 0, \quad E_{0r}^T P_c = P_c^T E_{0r} \geq 0 \quad (67)$$

$$\begin{bmatrix} A_{0r}^T \Theta^T P_c + P_c^T \Theta A_{0r} & * & * \\ \Phi \bar{C}_{0r} & \tilde{Q} + \tilde{Q}^T + 2P_f & * \\ B_{0r}^T \Theta^T P_c & \bar{D}_{0r}^T \tilde{\Phi}^T & -\mu_c^2 I \\ L_{0r} - D_f C_{0r} & -C_f & J_{0r} - D_f D_{0r} \\ \frac{1}{\varepsilon} \Xi_2^T N_{0r} A_{0r} & 0 & \frac{1}{\varepsilon} \Xi_2^T N_{0r} B_{0r} \\ M_{0r}^T P_c & 0 & 0 \\ M_{xr}^T \Theta^T P_c & M_{yr}^T \tilde{\Phi}^T & 0 \\ \rho N_{xr} & 0 & \rho N_{ur} \end{bmatrix}$$

$$\begin{bmatrix} * & * & * & * & * \\ * & * & * & * & * \\ * & * & * & * & * \\ -I & * & * & * & * \\ 0 & \frac{-1}{\varepsilon} I & * & * & * \\ 0 & 0 & \frac{-1}{\varepsilon} \Xi_3 & * & * \\ M_{xr}^T - M_{yr}^T D_f^T & \frac{1}{\varepsilon} M_{xr}^T N_{0r}^T \Xi_2^T & 0 & -\rho I & * \\ 0 & 0 & 0 & 0 & -\rho I \end{bmatrix} < 0 \quad (68)$$

where  $\Theta, \Xi_1, \Xi_2$  and  $\Xi_3$  are given in (31), then any normal filter  $\Sigma_f$  with  $E_f = I_r$  and

$$A_f = P_f^{-1} \tilde{Q} + I_r, \quad B_f = P_f^{-1} \tilde{\Phi}, \quad C_f, \quad D_f \quad (69)$$

is stable and makes (28) hold for  $\tilde{\Sigma}_e$  in (19).

*Proof:* The proof is similar to that for Theorem 2, except  $P_e$  is set to  $\text{diag}(P_c, P_f)$ . The remaining steps are the same and omitted here for the sake of brevity.  $\square$

*Remark 3:* On the basis of Theorem 3, the following convex optimisation problem may be formulated to find the  $H_\infty$  optimal normal filter of order  $r$  such that (28) is satisfied with the minimal  $\mu_c^2$

$$\min_{\mu_c^2, \rho, \varepsilon^{-1}, P_c, P_f, \tilde{\Phi}, \tilde{Q}, C_f, D_f} \mu_c^2 \quad (70)$$

subject to the LMIs (67) and (68).

## 4 Numerical example

In this section, an example is worked out to illustrate the proposed filter design methods. Suppose that the system matrices of the system  $\Sigma$  in (9) are as follows

$$\begin{aligned} E_0 &= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \quad A_0 = \begin{bmatrix} -2 & 1 & 0 \\ 0.5 & -1 & 1 \\ 0 & 0 & -1 \end{bmatrix} \\ B_0 &= \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \quad C_0 = [1 \quad -1 \quad 0] \\ L_0 &= [-2 \quad 1 \quad 0.5], \quad D_0 = 0.5, \quad J_0 = 0 \end{aligned} \quad (71)$$

The uncertainty matrices in (10) are assumed to be

$$\begin{aligned} M &= \begin{bmatrix} 0.1 \\ 0.1 \\ 0.1 \end{bmatrix}, \quad M_x = \begin{bmatrix} 0.5 \\ 1 \\ 0.5 \end{bmatrix}, \quad M_y = -1, \quad M_z = 2 \\ N &= [0.1 \quad -0.5 \quad 0], \quad N_x = [0.1 \quad 0 \quad 0.1], \quad N_u = 1 \end{aligned}$$

and  $|\Delta| \leq 1$ . It is easy to verify that  $(E_0 + M \Delta N, A_0 + M_x \Delta N_x)$  is an admissible pair. Let  $P = Q = I_3$  in (11) and  $P_\Delta = I_3 - M \Delta (I - 0.04 \Delta)^{-1} N$  as well as  $Q_\Delta = I_3$ . Then, the considered system  $\Sigma$  is r.s.e. to the system  $\tilde{\Sigma}_r$  in (16) with  $\tilde{E}_r = E_0$  in (71).

Three  $H_\infty$  optimal filters are designed by solving the convex optimisation problems listed in Remarks 1, 2 and 3, respectively, which are implemented by the MATLAB LMI Control Toolbox [13]. Because in this example  $E_{or} = PE_0Q = \text{diag}(1, 1, 0)$ , the non-strict LMIs in (29), (55) and (67) are manually adjusted to strict ones when applying the MATLAB LMI Control Toolbox. The resulting optimal  $\mu_c$ 's are 2.7591, 2.1415 and 2.4739, respectively, for the singular, third-order normal and second-order normal filters. Obviously, in this case, normal filters have better  $\mu_c$ 's than the singular filter. Also, the third-order normal filter, having larger degrees of freedom, owns a better  $\mu_c$  than that for the second-order normal filter.

As a check, for the  $H_\infty$  optimal singular filter, the pair  $(E_f, A_f)$  corresponding to  $U = V = P_c$  in (33) is

$$\left( \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \begin{bmatrix} -20.3388 & 12.4829 & 20.3279 \\ 16.2319 & -15.9105 & -6.9030 \\ 0.6263 & 0.4655 & -3.4626 \end{bmatrix} \right)$$

which is admissible.

## 5 Conclusion

The  $H_\infty$  optimal filter design problem has been considered for uncertain singular systems, in which uncertainties appear in all system matrices. For designing singular filters, the method proposed in Yu *et al.* [1] was extended, but for designing normal filters, two new methods were proposed, one giving higher-order filters with more degrees of freedom and one giving lower-order filters. By the numerical example used to illustrate the applications of the proposed methods, it is seen that singular filters should not be the only choice when considering uncertain singular systems.

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