# A Delay-Dependent Approach to Robust $H_{\infty}$ Filtering for Uncertain Distributed Delay Systems

Shengyuan Xu, James Lam, Senior Member, IEEE, Tongwen Chen, Senior Member, IEEE, and Yun Zou

Abstract—This paper is concerned with the problem of robust  $H_{\infty}$  filtering for linear systems with both discrete and distributed delays, which are subject to norm-bounded time-varying parameter uncertainties. Both the state and measurement equations are assumed to have discrete and distributed delays. A delay-dependent condition for the existence of  $H_{\infty}$  filters is proposed, which is less conservative than existing ones in the literature. Via solutions to certain linear matrix inequalities, general full-order filters are designed that ensure asymptotic stability and a prescribed  $H_{\infty}$  performance level, irrespective of the parameter uncertainties. An illustrative example is provided to demonstrate the effectiveness and the reduced conservatism of the proposed method.

Index Terms— $H_{\infty}$  filtering, distributed delay, linear matrix inequality, robust filtering, time-delay systems, uncertain systems.

#### I. INTRODUCTION

VER the past few years, a great deal of interest has been devoted to the study of  $H_{\infty}$  filtering problem, which is concerned with the design of estimators such that the  $\mathcal{L}_2$ -induced norm (for continuous systems) or  $l_2$ -induced norm (for discrete systems) from the noise signal to the estimation error is less than a prescribed level [1], [11], [15]. In the  $H_{\infty}$  setting, the noises are assumed to be arbitrary deterministic signals with bounded energy (or average power). Compared with traditional Kalman filtering, the  $H_{\infty}$  filtering approach does not require knowledge of the statistical properties of the external noises. In addition,  $H_{\infty}$  filtering is insensitive to uncertainty in the exogenous signal statistics as well as to uncertainty in dynamic models [19]. These features make the  $H_{\infty}$  filtering technique useful in many applications [3], [18].

Time delays are frequently encountered in many practical engineering systems, such as communication, electronics, hydraulic, and chemical systems. It is now well known that time delay is one of the main causes of instability and poor performance of a control system [5], [13], [14]. Therefore, there has been an increasing interest in the control and estimation for time-delay systems, and a great number of results on these topics

Manuscript received January 21, 2004; revised November 9, 2004. This work was supported by RGC HKU 7028/04P, the Foundation for the Author of National Excellent Doctoral Dissertation of China under Grant 200240, the National Natural Science Foundation of China under Grants 60304001 and 60474078, and the Fok Ying Tung Education Foundation under Grant 91061. The associate editor coordinating the review of this manuscript and approving it for publication was Prof. Trac D. Tran.

- S. Xu and Y. Zou are with the Department of Automation, Nanjing University of Science and Technology, Nanjing 210094, China.
- J. Lam is with the Department of Mechanical Engineering, University of Hong Kong, Hong Kong.
- T. Chen is with the Department of Electrical and Computer Engineering, University of Alberta, Edmonton, AB, Canada T6G 2V4.

Digital Object Identifier 10.1109/TSP.2005.855109

have been reported in the literature; see, e.g., [10], [12], [13], and the references therein. Recently, the  $H_{\infty}$  filtering problem for time-delay systems was studied in [17], where a Riccati equation approach was developed to solve the problem. When time delays appear in both the state and measurements, a sufficient condition for the existence of  $H_{\infty}$  filters was proposed in [6], in which a design method based on Riccati equations was presented. In the case when time delays and parameter uncertainties appear simultaneously, the robust  $H_{\infty}$  filtering problem was solved in [21] and [25] via a Riccati equation approach and a linear matrix inequality (LMI) approach, respectively. The corresponding results for the discrete case can be found in [16], [20], and [23]. It is noted that all the above  $H_{\infty}$  filtering results are derived for systems with discrete delays. When the number of summands in a system equation is increased and the differences between neighboring argument values are decreased, systems with distributed delays will arise. One application of distributed delay systems can be found in the modeling of feeding systems and combustion chambers in a liquid monopropellant rocket motor with pressure feeding [4], [7]. Very recently, the robust  $H_{\infty}$  filtering problem for such systems has been dealt with in [24], where sufficient conditions for the existence of  $H_{\infty}$  filters have been obtained in terms of LMIs. However, it should be pointed out that the aforementioned results for both the discrete delay case and distributed delay case are delay-independent, that is, they do not include any information on the size of delays. It is known that delay-dependent conditions are generally less conservative than delay-independent ones, especially when the size of the delay is small. Although delay-dependent results on the robust  $H_{\infty}$  filtering problem for systems with discrete delays were presented in [8] and [9], respectively, no delay-dependent  $H_{\infty}$  filtering results on distributed delay systems are available in the literature, which motivates the present study.

This paper deals with the problem of robust  $H_\infty$  filtering for linear uncertain systems with both discrete and distributed delays. The time delays are assumed to appear in both the state and measurement equations, and the parameter uncertainties are assumed to be time-varying but norm-bounded which appear in all the matrices in both the state and measurement equations. A delay-dependent condition for the existence of  $H_\infty$  filters is proposed and an LMI approach is developed, which is less conservative than existing ones in the literature. A general full order filter is sought to guarantee that the resulting error system is asymptotically stable and satisfies a prescribed  $H_\infty$  performance level for all admissible uncertainties. Desired  $H_\infty$  filters can be obtained by the solution to certain LMIs, which can be solved numerically and efficiently by resorting to standard numerical algorithms [2]. Finally, an illustrative example is pro-

vided to demonstrate the less conservatism and effectiveness of the proposed method.

**Notation.** Throughout this paper, the notation  $X \geq Y$  (respectively, X > Y) for real symmetric matrices X and Y means that the matrix X - Y is positive semi-definite (respectively, positive definite). The superscript "T" represents the transpose. I is an identity matrix with appropriate dimension.  $\mathcal{L}_2[0, \infty)$ is the space of square-integrable vector functions over  $[0, \infty)$ . The notation | refers to the Euclidean vector norm, whereas  $\|\cdot\|_2$  stands for the usual  $\mathcal{L}_2[0,\infty)$  norm. Matrices, if not explicitly stated, are assumed to have compatible dimensions.

#### II. PROBLEM FORMULATION

Consider the following uncertain distributed delay system:

$$\begin{split} (\Sigma): \qquad \dot{x}(t) &= A(t)x(t) + A_{d1}(t)x\left(t - \tau_{1}\right) \\ &+ A_{d2}(t) \int_{t - \tau_{2}}^{t} x(s)ds + B_{1}\omega(t) \quad (1) \\ y(t) &= C(t)x(t) + C_{d1}(t)x\left(t - \tau_{1}\right) \\ &+ C_{d2}(t) \int_{t - \tau_{2}}^{t} x(s)ds + B_{2}\omega(t) \quad (2) \\ z(t) &= Lx(t) + L_{d1}x\left(t - \tau_{1}\right) \quad (3) \end{split}$$

where  $x(t) \in \mathbb{R}^n$  is the state vector;  $y(t) \in \mathbb{R}^r$  is the measurement;  $z(t) \in \mathbb{R}^q$  is the signal to be estimated, and  $\omega(t) \in$  $\mathbb{R}^p$  is the noise input which belongs to  $\mathcal{L}_2[0,\infty)$ . The scalars  $\tau_1 > 0, \tau_2 > 0$  represent the time delays of the system;  $\tau =$  $\max(\tau_1, \tau_2), \varphi(t)$  is a real-valued continuous initial function on  $[-\tau,0]$ .  $B_1$ ,  $B_2$ , L, and  $L_{d1}$  are known real constant matrices. The matrices A(t),  $A_{d1}(t)$ ,  $A_{d2}(t)$ , C(t),  $C_{d1}(t)$ , and  $C_{d2}(t)$ are of the form

 $x(t) = \varphi(t), \quad \forall t \in [-\tau, 0]$ 

$$A(t) = A + \Delta A(t), \quad A_{d1}(t) = A_{d1} + \Delta A_{d1}(t)$$

$$A_{d2}(t) = A_{d2} + \Delta A_{d2}(t), \quad C(t) = C + \Delta C(t)$$

$$C_{d1}(t) = C_{d1} + \Delta C_{d1}(t), \quad C_{d2}(t) = C_{d2} + \Delta C_{d2}(t)$$

where A,  $A_{d1}$ ,  $A_{d2}$ , C,  $C_{d1}$ , and  $C_{d2}$  are known real constant matrices, and  $\Delta A(t)$ ,  $\Delta A_{d1}(t)$ ,  $\Delta A_{d2}(t)$ ,  $\Delta C(t)$ ,  $\Delta C_{d1}(t)$ , and  $\Delta C_{d2}(t)$  are unknown matrices representing time-varying parameter uncertainties, which are assumed to be of the form

$$\begin{bmatrix} \Delta A(t) & \Delta A_{d1}(t) & \Delta A_{d2}(t) \\ \Delta C(t) & \Delta C_{d1}(t) & \Delta C_{d2}(t) \end{bmatrix} = \begin{bmatrix} M_1 \\ M_2 \end{bmatrix} F(t) \begin{bmatrix} N_1 & N_2 & N_3 \end{bmatrix} \quad (5)$$

where  $M_1$ ,  $M_2$ ,  $N_1$ ,  $N_2$ , and  $N_3$  are known real constant matrices, and  $F(\cdot): \mathbb{R} \to \mathbb{R}^{k \times l}$  is an unknown time-varying matrix function satisfying

$$F(t)^T F(t) \le I, \quad \forall t.$$
 (6)

The uncertain matrices  $\Delta A(t)$ ,  $\Delta A_{d1}(t)$ ,  $\Delta A_{d2}(t)$ ,  $\Delta C(t)$ ,  $\Delta C_{d1}(t)$ , and  $\Delta C_{d2}(t)$  are said to be admissible if both (5) and (6) hold.

Remark 1: The distributed delay model in (1)–(4) can be used to describe some real systems, such as feeding systems and combustion chambers in a liquid monopropellant rocket motor with pressure feeding; see, e.g., [4], [7], and the references therein.

For system  $(\Sigma)$ , we now consider the following general fullorder filter for the estimate of z(t):

$$(\Sigma_f): \quad \dot{\hat{x}}(t) = A_f \hat{x}(t) + B_f y(t) \tag{7}$$

$$\hat{z}(t) = C_f \hat{x}(t) \tag{8}$$

where  $\hat{x}(t) \in \mathbb{R}^n$  and  $\hat{z}(t) \in \mathbb{R}^q$ .  $A_f, B_f$ , and  $C_f$  are matrices to be determined. Denote

$$e(t) = [x(t)^T \ \hat{x}(t)^T]^T, \ \hat{z}(t) = z(t) - \hat{z}(t).$$
 (9)

Then, the filtering error dynamics from the systems  $(\Sigma)$  and  $(\Sigma_f)$  can be obtained as

$$(\tilde{\Sigma}): \quad \dot{e}(t) = A_c(t)e(t) + A_{cd1}(t)He(t - \tau_1) + A_{cd2}(t)H \int_{t - \tau_2}^{t} e(s)ds + B_c\omega(t)$$
(10)  

$$\tilde{z}(t) = L_ce(t) + L_{cd1}He(t - \tau_1)$$
(11)

where

$$A_c(t) = A_c + \Delta A_c(t), \quad A_{cd1}(t) = A_{cd1} + \Delta A_{cd1}(t)$$
(12)

$$A_{cd2}(t) = A_{cd2} + \Delta A_{cd2}(t), \quad A_c = \begin{bmatrix} A & 0 \\ B_f C & A_f \end{bmatrix}$$
(13)

$$A_{cd1} = \begin{bmatrix} A_{d1} \\ B_f C_{d1} \end{bmatrix}, \quad A_{cd2} = \begin{bmatrix} A_{d2} \\ B_f C_{d2} \end{bmatrix}$$
 (14)

$$\Delta A_c(t) = \begin{bmatrix} \Delta A(t) & 0 \\ B_f \Delta C(t) & 0 \end{bmatrix}$$
 (15)

$$\Delta A_{cd1}(t) = \begin{bmatrix} \Delta A_{d1}(t) \\ B_f \Delta C_{d1}(t) \end{bmatrix}$$
 (16)

$$\Delta A_{cd1}(t) = \begin{bmatrix} \Delta A_{d1}(t) \\ B_f \Delta C_{d1}(t) \end{bmatrix}$$

$$\Delta A_{cd2}(t) = \begin{bmatrix} \Delta A_{d2}(t) \\ B_f \Delta C_{d2}(t) \end{bmatrix}, \quad B_c = \begin{bmatrix} B_1 \\ B_f B_2 \end{bmatrix}$$

$$H = \begin{bmatrix} I & 0 \end{bmatrix}, \quad L_c = \begin{bmatrix} L & -C_f \end{bmatrix}, \quad L_{cd1} = L_{d1}.$$

$$(18)$$

The purpose of this paper is to develop delay-dependent conditions for the existence of robust  $H_{\infty}$  filters for the uncertain distributed delay system  $(\Sigma)$ . Specifically, for given scalars  $\overline{\tau}_1 > 0$  and  $\overline{\tau}_2 > 0$ , we are concerned with finding an asymptotically stable filter  $(\Sigma_f)$  in the form of (10) and (11) such that for any constant time delays  $\tau_1$  and  $\tau_2$  satisfying  $0 < \tau_1 \le \overline{\tau}_1$ ,  $0 < \tau_2 \le \overline{\tau}_2$  the filtering error system  $(\Sigma)$  is asymptotically stable, and

$$\|\tilde{z}\|_2 < \gamma \|\omega\|_2 \tag{19}$$

under zero-initial conditions for any nonzero  $\omega(t) \in \mathcal{L}_2[0, \infty)$ and all admissible uncertainties, where  $\gamma > 0$  is a given scalar.

Before concluding this section, we present the following lemma, which will be used in the proof of our main results in Section III.

*Lemma:* [25] Let  $\mathcal{D}$ ,  $\mathcal{S}$ , and F be real matrices of appropriate dimensions with F satisfying  $F^TF < I$ . Then, for any scalar  $\epsilon > 0$ 

$$\mathcal{D}F\mathcal{S} + (\mathcal{D}F\mathcal{S})^T < \epsilon^{-1}\mathcal{D}\mathcal{D}^T + \epsilon\mathcal{S}^T\mathcal{S}.$$

### III. MAIN RESULTS

The following theorem is essential for solving the robust  $H_{\infty}$  filtering problem formulated in the previous section.

Theorem 1: For any delays  $\tau_1$  and  $\tau_2$  satisfying  $0 < \tau_1 \le \overline{\tau}_1$  and  $0 < \tau_2 \le \overline{\tau}_2$ , the filtering error system  $(\tilde{\Sigma})$  in (10) and (11) is robustly asymptotically stable and (19) is satisfied under zero-initial conditions for any nonzero  $\omega(t) \in L_2[0,\infty)$  and all admissible uncertainties if there exist matrices P>0,  $Q_1>0$ ,  $Q_2>0$ , Z>0,  $W_1$ ,  $W_2$  and a scalar  $\epsilon>0$  such that the LMI in (20), shown at the bottom of the page, holds, where

$$\Upsilon_1 = PA_c + A_c^T P + H^T (Q_1 - W_1 - W_1^T) H$$
 (21)

$$\Upsilon_2 = PA_{cd1} + H^T (W_1 - W_2^T)$$
 (22)

$$\Upsilon_3 = W_2 + W_2^T - Q_1 \tag{23}$$

$$\tilde{N}_1 = [N_1 \quad 0], \quad \tilde{N}_2 = N_2, \quad \tilde{N}_3 = N_3$$
 (24)

$$\tilde{M}_1 = \begin{bmatrix} M_1 \\ B_f M_2 \end{bmatrix}. \tag{25}$$

*Proof:* From (20), it can be seen that there exists a scalar  $\delta > 0$  such that we have (26), shown at the bottom of the page. By Schur complement equivalence, it follows from (26) that we have (27), shown at the bottom of the page. Note

$$\begin{bmatrix} \Delta A_c(t) & \Delta A_{cd1}(t) & \Delta A_{cd2}(t) \end{bmatrix} = \tilde{M}_1 F(t) \begin{bmatrix} \tilde{N}_1 & \tilde{N}_2 & \tilde{N}_3 \end{bmatrix}. \quad (28)$$

Then, using Lemma 1, we have the first equation at the bottom of the next page. This, together with (27), gives that, for any

$$\begin{bmatrix} \Upsilon_{1} + \overline{\tau}_{2}^{2} H^{T} Q_{2} H + \epsilon \tilde{N}_{1}^{T} \tilde{N}_{1} & \Upsilon_{2} + \epsilon \tilde{N}_{1}^{T} \tilde{N}_{2} & P A_{cd2} + \epsilon \tilde{N}_{1}^{T} \tilde{N}_{3} & P B_{c} & \overline{\tau}_{1} H^{T} W_{1} & L_{c}^{T} & \overline{\tau}_{1} A_{c}^{T} H^{T} Z & P \tilde{M}_{1} \\ \Upsilon_{2}^{T} + \epsilon \tilde{N}_{2}^{T} \tilde{N}_{1} & \Upsilon_{3} + \epsilon \tilde{N}_{2}^{T} \tilde{N}_{2} & \epsilon \tilde{N}_{2}^{T} \tilde{N}_{3} & 0 & \overline{\tau}_{1} W_{2} & L_{cd1}^{T} & \overline{\tau}_{1} A_{cd1}^{T} H^{T} Z & 0 \\ A_{cd2}^{T} P + \epsilon \tilde{N}_{3}^{T} \tilde{N}_{1} & \epsilon \tilde{N}_{3}^{T} \tilde{N}_{2} & \epsilon \tilde{N}_{3}^{T} \tilde{N}_{3} - Q_{2} & 0 & 0 & 0 & \overline{\tau}_{1} A_{cd2}^{T} H^{T} Z & 0 \\ B_{c}^{T} P & 0 & 0 & -\gamma^{2} I & 0 & 0 & \overline{\tau}_{1} B_{c}^{T} H^{T} Z & 0 \\ \overline{\tau}_{1} W_{1}^{T} H & \overline{\tau}_{1} W_{2}^{T} & 0 & 0 & -\overline{\tau}_{1} Z & 0 & 0 & 0 \\ L_{c} & L_{cd1} & 0 & 0 & 0 & -\overline{\tau}_{1} Z & 0 & 0 & 0 \\ \overline{\tau}_{1} Z H A_{c} & \overline{\tau}_{1} Z H A_{cd1} & \overline{\tau}_{1} Z H A_{cd2} & \overline{\tau}_{1} Z H B_{c} & 0 & 0 & -\overline{\tau}_{1} Z & \overline{\tau}_{1} Z H \tilde{M}_{1} \\ \tilde{M}_{1}^{T} P & 0 & 0 & 0 & 0 & \overline{\tau}_{1} \tilde{M}_{1}^{T} H^{T} Z & -\epsilon I \end{bmatrix}$$

$$\begin{bmatrix} \Upsilon_{1} + \overline{\tau}_{2}^{2}H^{T}Q_{2}H + \epsilon \tilde{N}_{1}^{T}\tilde{N}_{1} + \delta I & \Upsilon_{2} + \epsilon \tilde{N}_{1}^{T}\tilde{N}_{2} & PA_{cd2} + \epsilon \tilde{N}_{1}^{T}\tilde{N}_{3} & PB_{c} & \overline{\tau}_{1}H^{T}W_{1} & L_{c}^{T} & \overline{\tau}_{1}A_{c}^{T}H^{T}Z & P\tilde{M}_{1} \\ \Upsilon_{2}^{T} + \epsilon \tilde{N}_{2}^{T}\tilde{N}_{1} & \Upsilon_{3} + \epsilon \tilde{N}_{2}^{T}\tilde{N}_{2} & \epsilon \tilde{N}_{2}^{T}\tilde{N}_{3} & 0 & \overline{\tau}_{1}W_{2} & L_{cd1}^{T} & \overline{\tau}_{1}A_{cd1}^{T}H^{T}Z & 0 \\ A_{cd2}^{T}P + \epsilon \tilde{N}_{3}^{T}\tilde{N}_{1} & \epsilon \tilde{N}_{3}^{T}\tilde{N}_{2} & \epsilon \tilde{N}_{3}^{T}\tilde{N}_{3} - Q_{2} & 0 & 0 & 0 & \overline{\tau}_{1}A_{cd2}^{T}H^{T}Z & 0 \\ B_{c}^{T}P & 0 & 0 & -\gamma^{2}I & 0 & 0 & \overline{\tau}_{1}B_{c}^{T}H^{T}Z & 0 \\ \overline{\tau}_{1}W_{1}^{T}H & \overline{\tau}_{1}W_{2}^{T} & 0 & 0 & -\overline{\tau}_{1}Z & 0 & 0 & 0 \\ L_{c} & L_{cd1} & 0 & 0 & 0 & -\overline{\tau}_{1}Z & 0 & 0 & 0 \\ \overline{\tau}_{1}ZHA_{c} & \overline{\tau}_{1}ZHA_{cd1} & \overline{\tau}_{1}ZHA_{cd2} & \overline{\tau}_{1}ZHB_{c} & 0 & 0 & -\overline{\tau}_{1}Z & \overline{\tau}_{1}ZH\tilde{M}_{1} \\ \tilde{M}_{1}^{T}P & 0 & 0 & 0 & 0 & \overline{\tau}_{1}\tilde{M}_{1}^{T}H^{T}Z & -\epsilon I \end{bmatrix}$$

$$\begin{bmatrix} \Upsilon_{1} + \bar{\tau}_{2}^{2}H^{T}Q_{2}H + \delta I & \Upsilon_{2} & PA_{cd2} & PB_{c} & \bar{\tau}_{1}H^{T}W_{1} & L_{c}^{T} & \bar{\tau}_{1}A_{c}^{T}H^{T}Z \\ \Upsilon_{2}^{T} & \Upsilon_{3} & 0 & 0 & \bar{\tau}_{1}W_{2} & L_{cd1}^{T} & \bar{\tau}_{1}A_{cd1}^{T}H^{T}Z \\ A_{cd2}^{T}P & 0 & -Q_{2} & 0 & 0 & 0 & \bar{\tau}_{1}A_{cd2}^{T}H^{T}Z \\ B_{c}^{T}P & 0 & 0 & -\gamma^{2}I & 0 & 0 & \bar{\tau}_{1}B_{c}^{T}H^{T}Z \\ \bar{\tau}_{1}W_{1}^{T}H & \bar{\tau}_{1}W_{2}^{T} & 0 & 0 & -\bar{\tau}_{1}Z & 0 & 0 \\ L_{c} & L_{cd1} & 0 & 0 & 0 & -I & 0 \\ \bar{\tau}_{1}ZHA_{c} & \bar{\tau}_{1}ZHA_{cd1} & \bar{\tau}_{1}ZHA_{cd2} & \bar{\tau}_{1}ZHB_{c} & 0 & 0 & -\bar{\tau}_{1}Z \end{bmatrix}$$

$$+\epsilon^{-1}\begin{bmatrix} P\tilde{M}_{1} \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ \bar{\tau}_{1}ZH\tilde{M}_{1} \end{bmatrix} \begin{bmatrix} P\tilde{M}_{1} \\ 0 \\ 0 \\ 0 \\ 0 \\ \bar{\tau}_{1}ZH\tilde{M}_{1} \end{bmatrix}^{T} & \begin{bmatrix} \tilde{N}_{1}^{T} \\ \tilde{N}_{2}^{T} \\ \tilde{N}_{3}^{T} \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} \tilde{N}_{1}^{T} \\ \tilde{N}_{3}^{T} \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} < 0. \tag{27}$$

delays  $\tau_1$  and  $\tau_2$  satisfying  $0 < \tau_1 \le \overline{\tau}_1$  and  $0 < \tau_2 \le \overline{\tau}_2$ , we have (29), shown at the bottom of the page, where

$$\tilde{\Upsilon}_{1}(t) = PA_{c}(t) + A_{c}(t)^{T}P + H^{T}\left(Q_{1} - W_{1} - W_{1}^{T}\right)H \qquad V_{2}(t) = \int_{t-\tau_{1}}^{t} e(\alpha)^{T}H^{T}Q_{1}He(\alpha)d\alpha$$

$$\tilde{\Upsilon}_{2}(t) = PA_{cd_{1}}(t) + H^{T}\left(W_{1} - W_{2}^{T}\right). \qquad (31) \qquad V_{3}(t) = \int_{t-\tau_{1}}^{t} \left[\int_{t-\tau_{1}}^{t} e(\theta)^{T}H^{T}d\theta\right]Q_{2}$$

Now, we show the robust asymptotic stability of the filtering error system  $(\tilde{\Sigma})$ . To this end, we consider (10) with  $\omega(t)=0$ , that is

$$\dot{e}(t) = A_c(t)e(t) + A_{cd1}(t)He(t - \tau_1) + A_{cd2}(t)H \int_{t-\tau_2}^{t} e(s)ds. \quad (32)$$

Define the following Lyapunov function candidate for system (32):

$$V(t) = V_1(t) + V_2(t) + V_3(t) + V_4(t) + V_5(t)$$
(33)

where

$$V_{1}(t) = e(t)^{T} P e(t)$$

$$H \qquad V_{2}(t) = \int_{t-\tau_{1}}^{t} e(\alpha)^{T} H^{T} Q_{1} H e(\alpha) d\alpha$$

$$(30) \qquad V_{3}(t) = \int_{t-\tau_{2}}^{t} \left[ \int_{s}^{t} e(\theta)^{T} H^{T} d\theta \right] Q_{2} \left[ \int_{s}^{t} H e(\theta) d\theta \right] ds$$

$$V_{4}(t) = \int_{0}^{\tau_{2}} \int_{t-\beta}^{t} (\alpha - t + \beta) e(\alpha)^{T} H^{T} Q_{2} H e(\alpha) d\alpha d\beta$$

$$= 0, \qquad V_{5}(t) = \int_{-\tau_{1}}^{0} \int_{t+\beta}^{t} \dot{e}(\alpha)^{T} H^{T} Z H \dot{e}(\alpha) d\alpha d\beta.$$

Then, by Lemma 1, the time derivative of V(t) along the trajectory of the system (32) is given by

$$\dot{V}_{1}(t) = 2e(t)^{T} P \left[ A_{c}(t)e(t) + A_{cd1}(t)He(t - \tau_{1}) + A_{cd2}(t)H \int_{t-\tau_{2}}^{t} e(s)ds \right]$$
(34)

$$\dot{V}_{2}(t) = e(t)^{T} H^{T} Q_{1} H e(t) - e(t - \tau_{1})^{T}$$

$$\times H^{T} Q_{1} H e(t - \tau_{1})$$
(35)

$$\begin{bmatrix} \tilde{\Upsilon}_{1}(t) + \tau_{2}^{2}H^{T}Q_{2}H + \delta I & \tilde{\Upsilon}_{2}(t) & PA_{cd2}(t) & PB_{c} & \tau_{1}H^{T}W_{1} & L_{c}^{T} & \tau_{1}A_{c}(t)^{T}H^{T}Z \\ \tilde{\Upsilon}_{2}(t)^{T} & \tilde{\Upsilon}_{3} & 0 & 0 & \tau_{1}W_{2} & L_{cd1}^{T} & \tau_{1}A_{cd1}(t)^{T}H^{T}Z \\ A_{cd2}(t)^{T}P & 0 & -Q_{2} & 0 & 0 & 0 & \tau_{1}A_{cd2}(t)^{T}H^{T}Z \\ B_{c}^{T}P & 0 & 0 & -\gamma^{2}I & 0 & 0 & \tau_{1}B_{c}^{T}H^{T}Z \\ \tau_{1}W_{1}^{T}H & \tau_{1}W_{2}^{T} & 0 & 0 & -\tau_{1}Z & 0 & 0 \\ L_{c} & L_{cd1} & 0 & 0 & 0 & -I & 0 \\ \tau_{1}ZHA_{c}(t) & \tau_{1}ZHA_{cd1}(t) & \tau_{1}ZHA_{cd2}(t) & \tau_{1}ZHB_{c} & 0 & 0 & -\tau_{1}Z \end{bmatrix} < 0 \quad (29)$$

$$\dot{V}_{3}(t) \leq \frac{\tau_{2}^{2}}{2} e(t)^{T} H^{T} Q_{2} H e(t) - \eta(t)^{T} Q_{2} \eta(t) 
+ \int_{t-\tau_{2}}^{t} (\theta - t + \tau_{2}) e(\theta)^{T} H^{T} Q_{2} H e(\theta) d\theta \quad (36) 
\dot{V}_{4}(t) = \frac{\tau_{2}^{2}}{2} e(t)^{T} H^{T} Q_{2} H e(t) 
- \int_{t-\tau_{2}}^{t} (\alpha - t + \tau_{2}) e(\alpha)^{T} H^{T} Q_{2} H e(\alpha) d\alpha \quad (37) 
\dot{V}_{5}(t) = \tau_{1} \dot{e}(t)^{T} H^{T} Z H \dot{e}(t) 
- \int_{t-\tau_{2}}^{t} \dot{e}(\alpha)^{T} H^{T} Z H \dot{e}(\alpha) d\alpha \quad (38)$$

where

$$\eta(t) = \int_{t-\tau_2}^{t} He(\theta) d\theta.$$

Therefore, noting

$$e(t - \tau_1) = e(t) - \int_{t - \tau_1}^t \dot{e}(\alpha) d\alpha,$$

and using (34)–(38), we have

$$\dot{V}(t) \leq 2e(t)^{T} P \left[ A_{c}(t)e(t) + A_{cd1}(t) H e(t - \tau_{1}) + A_{cd2}(t) \eta(t) \right] 
+ e(t)^{T} H^{T} \left( Q_{1} + \tau_{2}^{2} Q_{2} \right) H e(t) 
- e(t - \tau_{1})^{T} H^{T} Q_{1} H e(t - \tau_{1}) - \eta(t)^{T} Q_{2} \eta(t) 
- \int_{t - \tau_{1}}^{t} \dot{e}(\alpha)^{T} H^{T} Z H \dot{e}(\alpha) d\alpha 
+ \tau_{1} \left[ A_{c}(t)e(t) + A_{cd1}(t) \right] 
\times H e(t - \tau_{1}) + A_{cd2}(t) \eta(t) T^{T} 
\times H^{T} Z H \left[ A_{c}(t)e(t) + A_{cd1}(t) H e(t - \tau_{1}) \right] 
+ A_{cd2}(t) \eta(t) T^{T} 
+ 2e(t)^{T} H^{T} W_{1} H \int_{t - \tau_{1}}^{t} \dot{e}(\alpha) d\alpha - 2e(t)^{T} H^{T}$$

$$\times W_1 H \left[ e(t) - e(t - \tau_1) \right]$$

$$+ 2e(t - \tau_1)^T H^T W_2 H \int_{t - \tau_1}^t \dot{e}(\alpha) d\alpha - 2e(t - \tau_1)^T$$

$$\times H^T W_2 H \left[ e(t) - e(t - \tau_1) \right]$$

$$= \frac{1}{\tau_1} \int_{t - \tau_1}^t \xi(t, \alpha)^T \Omega(t) \xi(t, \alpha) d\alpha$$

$$(39)$$

where we have the two equations at the bottom of the page. Note that (29) implies (40), shown at the bottom of the page. Applying the Schur complement formula to (40), we obtain

This, together with (39), gives

$$\dot{V}(t) \le -\delta \left| x(t) \right|^2. \tag{42}$$

Therefore, along a similar line as in the proof of [26, Th. 1], it follows from (42) that the filtering error system  $(\tilde{\Sigma})$  is robustly asymptotically stable for any delays  $\tau_1$  and  $\tau_2$  satisfying  $0 < \tau_1 \leq \overline{\tau}_1, 0 < \tau_2 \leq \overline{\tau}_2$ .

Next, we will establish the  $H_{\infty}$  performance of the filtering error system  $(\tilde{\Sigma})$  under zero initial condition. To this end, we introduce

$$J(t) = \int_0^t \left[ \tilde{z}(s)^T \tilde{z}(s) - \gamma^2 \omega(s)^T \omega(s) \right] ds \qquad (43)$$

where t > 0. Noting zero initial condition, it can be shown that for any nonzero  $\omega(t) \in \mathcal{L}_2[0,\infty)$  and t > 0

$$J(t) \le \int_0^t \left[ \tilde{z}(s)^T \tilde{z}(s) - \gamma^2 \omega(s)^T \omega(s) + \dot{V}(s) \right] ds \quad (44)$$

where V(s) is defined in (33). Then, following a similar argument as in the derivation of (39), we can obtain

$$\tilde{z}(s)^T \tilde{z}(s) - \gamma^2 \omega(s)^T \omega(s) + \dot{V}(s)$$

$$\leq \frac{1}{\tau_1} \int_{s-\tau_1}^s \zeta(s,\alpha)^T \hat{\Omega}(s) \zeta(s,\alpha) d\alpha \quad (45)$$

$$\begin{bmatrix} \tilde{\Upsilon}_{1}(t) + \tau_{2}^{2}H^{T}Q_{2}H + \delta I & \tilde{\Upsilon}_{2}(t) & PA_{cd2}(t) & \tau_{1}H^{T}W_{1} & \tau_{1}A_{c}(t)^{T}H^{T}Z \\ \tilde{\Upsilon}_{2}(t)^{T} & \Upsilon_{3} & 0 & \tau_{1}W_{2} & \tau_{1}A_{cd1}(t)^{T}H^{T}Z \\ A_{cd2}(t)^{T}P & 0 & -Q_{2} & 0 & \tau_{1}A_{cd2}(t)^{T}H^{T}Z \\ \tau_{1}W_{1}^{T}H & \tau_{1}W_{2}^{T} & 0 & -\tau_{1}Z & 0 \\ \tau_{1}ZHA_{c}(t) & \tau_{1}ZHA_{cd1}(t) & \tau_{1}ZHA_{cd2}(t) & 0 & -\tau_{1}Z \end{bmatrix} < 0. \tag{40}$$

where we have the first two equations at the bottom of the page. On the other hand, by the Schur complement formula, it follows from (29) that

By this and (45), we have

$$\tilde{z}(s)^T \tilde{z}(s) - \gamma^2 \omega(s)^T \omega(s) + \dot{V}(s) < 0$$

which, together with (44), implies that J(t) < 0 for any nonzero  $\omega(t) \in \mathcal{L}_2[0,\infty)$ . Thus, for any delays  $\tau_1$  and  $\tau_2$  satisfying  $0 < \tau_1 \le \overline{\tau}_1$  and  $0 < \tau_2 \le \overline{\tau}_2$ , the inequality in (19) holds. This completes the proof.

Remark 2: The stability condition for distributed delay system in (1) with  $\omega(t) = 0$  can be easily inferred from Theorem 1; such a stability condition will be less conservative than that in [22] since in [22], some bounding techniques for some cross terms are used, whereas in Theorem 1, such techniques are not used.

Now, we are in a position to present a solution to the robust  $H_{\infty}$  filtering problem.

Theorem 2: Consider the uncertain distributed delay system  $(\Sigma)$ , and let  $\gamma > 0$  be a prescribed constant scalar. Then, for any delays  $\tau_1$  and  $\tau_2$  satisfying  $0 < \tau_1 \le \overline{\tau}_1$ ,  $0 < \tau_2 \le \overline{\tau}_2$ , there exists a filter in the form of (7) and (8) such that the filtering error system  $(\tilde{\Sigma})$  is asymptotically stable, and (19) is satisfied if there exist matrices  $X > 0, Y > 0, Q_1 > 0, Q_2 > 0, Z > 0,$ 

 $W_1, W_2, \Lambda, \Phi, \Psi$ , and a scalar  $\epsilon > 0$  such that LMIs in (46), shown at the bottom of the page, hold, and

$$X - Y > 0 \tag{47}$$

where

$$J_1 = YA + A^TY + Q_1 + \overline{\tau}_2^2 Q_2 - W_1 - W_1^T + \epsilon N_1^T N_1$$
(48)

$$J_2 = A^T X + XA + \Psi C + C^T \Psi^T + Q_1 + \bar{\tau}_2^2 Q_2 - W_1 - W_1^T + \epsilon N_1^T N_1$$
(49)

$$J_3 = W_2 + W_2^T - Q_1 + \epsilon N_2^T N_2 \tag{50}$$

$$J_4 = \epsilon N_3^T N_3 - Q_2 \tag{51}$$

$$G_1 = A^T X + Y A + C^T \Psi^T + \Phi^T + Q_1 + \overline{\tau}_2^2 Q_2 - W_1 - W_1^T + \epsilon N_1^T N_1$$

$$-W_1 - W_1^T + \epsilon N_1^T N_1$$

$$G_2 = Y A_{d1} + W_1 - W_2^T + \epsilon N_1^T N_2$$
(52)

$$G_2 = YA_{d1} + W_1 - W_2^1 + \epsilon N_1^1 N_2 \tag{53}$$

$$G_3 = YA_{d2} + \epsilon N_1^T N_3 \tag{54}$$

$$G_4 = XA_{d1} + \Psi C_{d1} + W_1 - W_2^T + \epsilon N_1^T N_2$$
 (55)

$$G_5 = X A_{d2} + \Psi C_{d2} + \epsilon N_1^T N_3 \tag{56}$$

and the matrix function  $U(K_1, K_2)$  is defined as

$$U(K_1, K_2) = XK_1 + \Psi K_2.$$

In this case, a desired robust  $H_{\infty}$  filter is given in the form of (7) and (8) with parameters as follows:

$$A_f = S_1^{-1} \Phi Y^{-1} S_2^{-T}, \quad B_f = S_1^{-1} \Psi$$

$$C_f = \Lambda Y^{-1} S_2^{-T}$$
(57)

where  $S_1$  and  $S_2$  are any nonsingular matrices satisfying

$$S_1 S_2^T + X Y^{-1} = I. (58)$$

$$\begin{split} \zeta(s,\alpha) &= \begin{bmatrix} e(s)^T & (He(s-\tau_1))^T & \eta(s)^T & \omega(s)^T & (H\dot{e}(\alpha))^T \end{bmatrix}^T \\ \hat{\Omega}(s) &= \begin{bmatrix} \hat{\Upsilon}_1(s) + \tau_2^2 H^T Q_2 H + L_c^T L_c & \hat{\Upsilon}_2(s) + L_c^T L_{cd1} & PA_{cd2}(s) & PB_c & \tau_1 H^T W_1 \\ \hat{\Upsilon}_2(s)^T + L_{cd1}^T L_c & \Upsilon_3 + L_{cd1}^T L_{cd1} & 0 & 0 & \tau_1 W_2 \\ A_{cd2}(s)^T P & 0 & -Q_2 & 0 & 0 \\ B_c^T P & 0 & 0 & -\gamma^2 I & 0 \\ \tau_1 W_1^T H & \tau_1 W_2^T & 0 & 0 & -\tau_1 Z \end{bmatrix} \\ &+ \tau_1 \begin{bmatrix} A_c(s)^T H^T \\ A_{cd1}(s)^T H^T \\ A_{cd2}(s)^T H^T \\ B_c^T H^T \\ 0 \end{bmatrix} Z \begin{bmatrix} A_c(s)^T H^T \\ A_{cd1}(s)^T H^T \\ A_{cd2}(s)^T H^T \\ A_{cd2}(s)^T H^T \\ B_c^T H^T \\ 0 \end{bmatrix} . \end{split}$$

$$\begin{bmatrix} J_1 & G_1 & G_2 & G_3 & YB_1 & \overline{\tau}_1W_1 & L^T - \Lambda^T & \overline{\tau}_1A^TZ & YM_1 \\ G_1^T & J_2 & G_4 & G_5 & U(B_1, B_2) & \overline{\tau}_1W_1 & L^T & \overline{\tau}_1A^TZ & U(M_1, M_2) \\ G_2^T & G_4^T & J_3 & \epsilon N_2^T N_3 & 0 & \overline{\tau}_1W_2 & L_{d1}^T & \overline{\tau}_1A_{d1}^TZ & 0 \\ G_3^T & G_5^T & \epsilon N_3^T N_2 & J_4 & 0 & 0 & 0 & \overline{\tau}_1A_{d2}^TZ & 0 \\ B_1^T Y & U(B_1, B_2)^T & 0 & 0 & -\gamma^2 I & 0 & 0 & \overline{\tau}_1B_1^TZ & 0 \\ \overline{\tau}_1W_1^T & \overline{\tau}_1W_1^T & \overline{\tau}_1W_2^T & 0 & 0 & -\overline{\tau}_1Z & 0 & 0 \\ L - \Lambda & L & L_{d1} & 0 & 0 & 0 & -I & 0 & 0 \\ \overline{\tau}_1ZA & \overline{\tau}_1ZA & \overline{\tau}_1ZA_{d1} & \overline{\tau}_1ZA_{d2} & \overline{\tau}_1ZB_1 & 0 & 0 & -\overline{\tau}_1Z & \overline{\tau}_1ZM_1 \\ M_1^T Y & U(M_1, M_2)^T & 0 & 0 & 0 & 0 & \overline{\tau}_1M_1^TZ & -\epsilon I \end{bmatrix} < 0$$

*Proof:* By (47), it is easy to see that  $I - XY^{-1}$  is non-singular. Therefore, there always exist nonsingular matrices  $S_1$  and  $S_2$  such that (58) holds. Similar to [24], we define

$$\Pi_1 = \begin{bmatrix} \bar{Y} & I \\ S_2^T & 0 \end{bmatrix}, \quad \Pi_2 = \begin{bmatrix} I & X \\ 0 & S_1^T \end{bmatrix}$$
(59)

where

$$\bar{Y} = Y^{-1} > 0.$$

Let

$$P = \Pi_2 \Pi_1^{-1}. \tag{60}$$

Then, by (47) and (58), it can be verified that P>0. Now, pre- and post-multiplying the LMI in (46) by diag  $(\bar{Y},I,I,I,I,I,I,I,I)$ , we have (61), shown at the bottom of the page, where  $\Upsilon_1$ ,  $\Upsilon_2$ ,  $\Upsilon_3$ ,  $\tilde{N}_1$ ,  $\tilde{N}_2$ ,  $\tilde{N}_3$ , and  $\tilde{M}_1$  are given in (21)–(25), respectively, and  $A_f$ ,  $B_f$ , and  $C_f$  are given in (57). Finally, pre- and post-multiplying the LMI in (61) by diag  $(\Pi_1^{-T}, I, I, I, I, I, I, I)$  and its transpose, respectively, and then using Theorem 1, we obtain the desired result.

Remark 3: Theorem 2 provides a sufficient condition for the solvability of the robust  $H_{\infty}$  filtering problem for uncertain distributed delay systems. Since the condition in (46) and (47) depends on the size of the delays, it will be less conservative than the existing delay-independent ones in [24]. It is also worth pointing out that a desired  $H_{\infty}$  filter can be obtained by solving the LMIs in (46) and (47), which can be implemented by using standard numerical algorithms [2]. The results in Theorem 2 can be extended to multiple delay case along a similar line as in the derivation of Theorem 2.

## IV. ILLUSTRATIVE EXAMPLE

Consider the uncertain distributed delay system  $(\Sigma)$  with parameters as follows:

$$A = \begin{bmatrix} -1.8 & 0.5 \\ 0.3 & -2.2 \end{bmatrix}, \quad A_{d1} = \begin{bmatrix} -3.6 & 1.5 \\ 1.2 & -2.8 \end{bmatrix}$$

$$A_{d2} = \begin{bmatrix} 0 & 0.5 \\ 0.2 & -0.7 \end{bmatrix}, \quad B_1 = \begin{bmatrix} -0.6 \\ 0 \end{bmatrix}$$

$$C = \begin{bmatrix} -0.8 & 0.2 \end{bmatrix}, \quad C_{d1} = \begin{bmatrix} 0.6 & 0 \end{bmatrix}$$

$$C_{d2} = \begin{bmatrix} -0.2 & 0.8 \end{bmatrix}, \quad B_2 = 0.5$$

$$L = \begin{bmatrix} 0.2 & 0.1 \end{bmatrix}, \quad L_{d1} = \begin{bmatrix} 0 & 0 \end{bmatrix}$$

$$M_1 = \begin{bmatrix} 0.1 \\ -0.2 \end{bmatrix}, \quad M_2 = 0.1$$

$$N_1 = \begin{bmatrix} -0.1 & 0 \end{bmatrix}, \quad N_2 = \begin{bmatrix} 0.1 & -0.2 \end{bmatrix}$$

$$N_3 = \begin{bmatrix} 0.2 & -0.1 \end{bmatrix}.$$

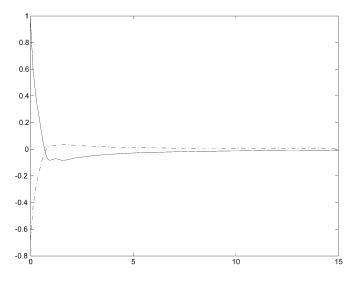


Fig. 1. State response of  $\hat{x}_1(t)$  (—) and  $\hat{x}_2(t)$  (-·).

In this example, we suppose  $\tau_2=0.8$ , and the  $H_\infty$  performance level  $\gamma=0.6$ . Then, it can be verified that the conditions in [24] are not satisfied for this system, which implies that the conditions in [24] fail to conclude whether or not there exist  $H_\infty$  filters for this system. However, by Theorem 2 in this paper, it can be calculated that for all  $0<\tau_1\le 0.3558$ , there exist robust  $H_\infty$  filters. As an example, we assume  $\tau_1=0.2$ . In this case, we use the Matlab LMI Control Toolbox to solve the LMIs in (46) and (47) and choose

$$S_1 = \begin{bmatrix} -1.2 & 0.6 \\ 0.2 & 1.5 \end{bmatrix}, \quad S_2 = \begin{bmatrix} 0.4458 & -0.1717 \\ -0.0758 & -0.2657 \end{bmatrix}. \quad (62)$$

It can be seen that the matrices  $S_1$  and  $S_2$  chosen in (62) satisfy (58). Therefore, according to Theorem 2, a desired  $H_{\infty}$  filter can be computed as

$$\begin{split} \dot{\hat{x}}(t) &= \begin{bmatrix} -3.5576 & -1.1022 \\ -0.1576 & -3.2383 \end{bmatrix} \hat{x}(t) + \begin{bmatrix} -1.1812 \\ 0.4863 \end{bmatrix} y(t) \\ \hat{z}(t) &= \begin{bmatrix} 0.2310 & -0.1644 \end{bmatrix} \hat{x}(t). \end{split}$$

The simulation result of the state response of the designed filter is given in Fig. 1, where the initial condition is  $\begin{bmatrix} 1 & -0.7 \end{bmatrix}^T$ , and the exogenous disturbance input  $\omega(t)$  is given as

$$\omega(t) = \frac{1}{0.5 + 1.2t}, \quad t \ge 0.$$

Fig. 2 is the simulation result of the error response of  $z(t) - \hat{z}(t)$ . From these simulation results, it can be seen that the designed  $H_{\infty}$  filter satisfies the specified requirements.

$$\begin{bmatrix} \Pi_{1}^{T} \left( \Upsilon_{1} + \epsilon \tilde{N}_{1}^{T} \tilde{N}_{1} \right) & \Pi_{1}^{T} \left( \Upsilon_{2} + \epsilon \tilde{N}_{1}^{T} \tilde{N}_{2} \right) & \Pi_{1}^{T} \left( PA_{cd2} + \epsilon \tilde{N}_{1}^{T} \tilde{N}_{3} \right) & \Pi_{1}^{T} PB_{c} & \overline{\tau}_{1} \Pi_{1}^{T} H^{T} W_{1} & \Pi_{1}^{T} L_{c}^{T} & \overline{\tau}_{1} \Pi_{1}^{T} A_{c}^{T} H^{T} Z & \Pi_{1}^{T} P\tilde{M}_{1} \\ \left( \Upsilon_{2}^{T} + \epsilon \tilde{N}_{2}^{T} \tilde{N}_{1} \right) \Pi_{1} & \Upsilon_{3} + \epsilon \tilde{N}_{2}^{T} \tilde{N}_{2} & \epsilon \tilde{N}_{2}^{T} \tilde{N}_{3} & 0 & \overline{\tau}_{1} W_{2} & L_{cd1}^{T} & \overline{\tau}_{1} A_{cd1}^{T} H^{T} Z & 0 \\ \left( A_{cd2}^{T} P + \epsilon \tilde{N}_{3}^{T} \tilde{N}_{1} \right) \Pi_{1} & \epsilon \tilde{N}_{3}^{T} \tilde{N}_{2} & \epsilon \tilde{N}_{3}^{T} \tilde{N}_{3} - Q_{2} & 0 & 0 & 0 & \overline{\tau}_{1} A_{cd2}^{T} H^{T} Z & 0 \\ B_{c}^{T} P\Pi_{1} & 0 & 0 & -\gamma^{2} I & 0 & 0 & \overline{\tau}_{1} B_{c}^{T} H^{T} Z & 0 \\ \overline{\tau}_{1} W_{1}^{T} H \Pi_{1} & \overline{\tau}_{1} W_{2}^{T} & 0 & 0 & -\overline{\tau}_{1} Z & 0 & 0 \\ L_{c} \Pi_{1} & L_{cd1} & 0 & 0 & 0 & -\overline{\tau}_{1} Z & 0 & 0 \\ \overline{\tau}_{1} Z H A_{c} \Pi_{1} & \overline{\tau}_{1} Z H A_{cd1} & \overline{\tau}_{1} Z H A_{cd2} & \overline{\tau}_{1} Z H B_{c} & 0 & 0 & -\overline{\tau}_{1} Z & \overline{\tau}_{1} Z H \tilde{M}_{1} \\ \tilde{M}_{1}^{T} P\Pi_{1} & 0 & 0 & 0 & 0 & \overline{\tau}_{1} \tilde{M}_{1}^{T} H^{T} Z & -\epsilon I \end{bmatrix}$$

$$(61)$$

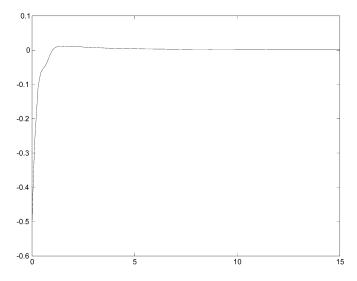


Fig. 2. Error response  $\bar{z}(t)$ .

#### V. CONCLUSION

In this paper, the problem of robust  $H_{\infty}$  filtering for linear uncertain systems with both discrete and distributed delays has been studied. A delay-dependent approach for the design of general full-order filters has been developed. The designed filter guarantees asymptotic stability and a prescribed  $H_{\infty}$  performance level of the error system for all admissible uncertainties. The derived condition is less conservative than existing ones in the literature, which has been demonstrated by an illustrative example.

#### REFERENCES

- [1] S. Bittanti and F. A. Cuzzola, "Continuous-time periodic  $H_{\infty}$  filtering via LMI," *Eur. J. Control*, vol. 7, pp. 2–16, 2001.
- [2] S. Boyd, L. E. Ghaoui, E. Feron, and V. Balakrishnan, *Linear Matrix Inequalities in System and Control Theory*. Philadelphia, PA: SIAM, 1994.
- [3] T. Chen and B. A. Francis, Optimal Sampled-Data Control Systems. London, U.K.: Springer, 1995.
- [4] L. Crocco, "Aspects of combustion stability in liquid propellant rocket motors, part I: Fundamentals-low frequency instability with monopropellants," J. Amer. Rocket Society, vol. 21, pp. 163–178, 1951.
- [5] L. Dugard and E. I. Verriest, Stability and Control of Time-Delay Systems. London, U.K.: Springer-Verlag, 1998.
- [6] A. Fattouh, O. Sename, and J.-M. Dion, " $H_{\infty}$  observer design for time-delay systems," in *Proc. 37th IEEE Conf. Decision Control*, Tampa, FL, Dec. 1998, pp. 4545–4546.
- [7] Y. A. Fiagbedzi and A. E. Pearson, "A multistage reduction technique for feedback stabilizing distributed time-lag systems," *Automatica*, vol. 23, pp. 311–326, 1987.
- [8] E. Fridman, U. Shaked, and L. Xie, "Robust  $H_{\infty}$  filtering of linear systems with time-varying delay," *IEEE Trans. Autom. Control*, vol. 48, no. 1, pp. 159–165, Jan. 2003.
- [9] H. Gao and C. Wang, "Delay-dependent robust  $H_{\infty}$  and  $L_2$ - $L_{\infty}$  filtering for a class of uncertain nonlinear time-delay systems," *IEEE Trans. Autom. Control*, vol. 48, no. 9, pp. 1661–1666, Sep. 2003.
- [10] ——, "A delay-dependent approach to robust  $H_{\infty}$  filtering for uncertain discrete-time state-delayed systems," *IEEE Trans. Signal Process.*, vol. 52, no. 6, pp. 1631–1640, Jun. 2004.

- [11] M. J. Grimble and A. E. Sayed, "Solution of the  $H_{\infty}$  optimal linear filtering problem for discrete-time systems," *IEEE Trans. Acoust., Speech, Signal Process.*, vol. 38, no. 7, pp. 1092–1104, Jul. 1990.
- [12] J. K. Hale, Theory of Functional Differential Equations. New York: Springer-Verlag, 1977.
- [13] V. B. Kolmanovskii and A. D. Myshkis, Applied Theory of Functional Differential Equations. Dordrecht, The Netherlands: Kluwer, 1992.
- [14] X. Li and C. E. de Souza, "Delay-dependent robust stability and stabilization of uncertain linear delay systems: A linear matrix inequality approach," *IEEE Trans. Autom. Control*, vol. 42, no. 8, pp. 1144–1148, Aug. 1997.
- [15] K. M. Nagpal and P. P. Khargonekar, "Filtering and smoothing in an  $H_{\infty}$  setting," *IEEE Trans. Autom. Control*, vol. 36, no. 2, pp. 152–166, Feb. 1991
- [16] R. M. Palhares, C. E. de Souza, and P. L. D. Peres, "Robust  $H_{\infty}$  filtering for uncertain discrete-time state-delayed systems," *IEEE Trans. Signal Process.*, vol. 49, no. 8, pp. 1696–1703, Aug, 2001.
- [17] A. W. Pila, U. Shaked, and C. E. de Souza, " $H_{\infty}$  filtering for continuous-time linear systems with delay," *IEEE Trans. Autom. Control*, vol. 44, no. 7, pp. 1412–1417, Jul. 1999.
- [18] U. Shaked, "H<sub>∞</sub> minimum error state estimation of linear stationary processes," *IEEE Trans. Autom. Control*, vol. 35, no. 5, pp. 554–558, May 1990.
- [19] U. Shaked and Y. Theodor, "H<sub>∞</sub>-optimal estimation: A tutorial," in Proc. 31st IEEE Conf. Decision Control, Tucson, AZ, USA, Dec. 1992, pp. 2278–2286.
- [20] Z. Wang, B. Huang, and H. Unbehauen, "Robust  $H_{\infty}$  observer design of linear state delayed systems with parametric uncertainty: The discrete-time case," *Automatica*, vol. 35, pp. 1161–1167, 1999.
- [21] —, "Robust  $H_{\infty}$  observer design of linear time-delay systems with parametric uncertainty," Syst. Control Lett., vol. 42, pp. 303–312, 2001.
- [22] L. Xie, E. Fridman, and U. Shaked, "Robust  $H_{\infty}$  control of distributed delay systems with application to combustion control," *IEEE Trans. Automat. Control*, vol. 46, no. 12, pp. 1930–1935, Dec. 2001.
- [23] S. Xu, "Robust  $H_{\infty}$  filtering for a class of discrete-time uncertain non-linear systems with state delay," *IEEE Trans. Circuits Syst. I*, vol. 49, no. 12, pp. 1853–1859, Dec. 2002.
- [24] S. Xu and T. Chen, "An LMI approach to the  $H_{\infty}$  filter design for uncertain systems with distributed delays," *IEEE Trans. Circuits Syst. II*, vol. 51, no. 4, pp. 195–201, Apr. 2004.
- [25] S. Xu and P. Van Dooren, "Robust  $H_{\infty}$  filtering for a class of nonlinear systems with state delay and parameter uncertainty," *Int. J. Control*, vol. 75, pp. 766–774, 2002.
- [26] S. Xu, P. Van Dooren, R. Stefan, and J. Lam, "Robust stability and stabilization for singular systems with state delay and parameter uncertainty," *IEEE Trans. Autom. Control*, vol. 47, no. 7, pp. 1122–1128, Jul. 2002.



**Shengyuan Xu** received the B.Sc. degree from Hangzhou Normal University, Hangzhou, China, the M.Sc. degree from Qufu Normal University, Qufu, China, and the Ph.D. degree from Nanjing University of Science and Technology, Nanjing, China, in 1990, 1996, and 1999, respectively.

From 1999 to 2000, he was a Research Associate with the Department of Mechanical Engineering, University of Hong Kong. From December 2000 to November 2001 and December 2001 to September 2002, he was a Postdoctoral Researcher with

CESAME, Universitè Catholique de Louvain, Louvain-la-Neuve, Belgium, and the Department of Electrical and Computer Engineering, University of Alberta, Edmonton, AB, Canada, respectively. From September 2002 to September 2003 and September 2003 to September 2004, he was a William Mong Young Researcher and an Honorary Associate Professor, respectively, both with the Department of Mechanical Engineering, University of Hong Kong. Currently, he is a Professor and Ph.D. Supervisor with the Department of Automation, Nanjing University of Science and Technology, Nanjing, China.

Dr. Xu received the National Excellent Doctoral Dissertation Award in 2002 from the National Education Commission of China. His current research interests include robust filtering and control, singular systems, time-delay systems, multidimensional systems, and nonlinear systems.



James Lam (SM'99) received the first-class B.Sc. degree in mechanical engineering from the University of Manchester, Manchester, U.K., in 1983. He received the M.Phil. and Ph.D. degrees in control engineering in 1985 and 1988, respectively, from the University of Cambridge Cambridge, U.K. His postdoctoral research was carried out in the Australian National University, Canberra, Australia, between 1990 and 1992.

His research interests include model reduction, delay systems, descriptor systems, stochastic sys-

tems, multidimensional systems, robust control and filtering, fault detection, and reliable control. He has held faculty positions at the City University of Hong Kong and the University of Melbourne, Parkville, Australia. He is now an Associate Professor with the Department of Mechanical Engineering, the University of Hong Kong, and holds a Concurrent Professorship at the Northeastern University, Shenyang, China; a Guest Professorship at the Huazhong University of Science and Technology, Wuhan, China; a Consulting Professorship at the South China University of Technology, Guangzhou, China; and Guest Professorship of Shandong University, Jinan, China. He is an Associate Editor of the Asian Journal of Control and the International Journal of Applied Mathematics and Computer Science.

Dr. Lam is a Chartered Mathematician, a Fellow of the Institute of Mathematics and Its Applications (U.K.), and a Member of the Institution of Electrical Engineers (U.K.). He received the Ashbury Scholarship, the A. H. Gibson Prize, and the H. Wright Baker Prize for his academic performance. Dr. Lam is a Scholar (1984) and Fellow (1990) of the Croucher Foundation.



**Tongwen Chen** (SM'97) received the B.Sc. degree from Tsinghua University, Beijing, China, in 1984 and the M.A.Sc. and Ph.D. degrees from the University of Toronto, Toronto, ON, Canada, in 1988 and 1991, respectively, all in electrical engineering.

From 1991 to 1997, he was on the Faculty of the Department of Electrical and Computer Engineering, University of Calgary, Calgary, AB, Canada. Since May 1997, he has been with the Department of Electrical and Computer Engineering, University of Alberta, Edmonton, Canada, and is presently a

Professor of electrical engineering. He held visiting positions at the Hong Kong University of Science and Technology and Tsinghua University. His current research interests include digital signal processing, multirate systems, process and robust control, and their applications to industrial problems. He co-authored, with B. A. Francis, the book *Optimal Sampled-Data Control Systems* (New York: Springer, 1995). He is an Associate Editor for *Systems and Control Letters* and a Member of the Editorial Board of *Dynamics of Continuous, Discrete and Impulsive Systems—Series B: Applications and Algorithms*.

Dr. Chen received a University of Alberta McCalla Professorship from 2000 to 2001. He was an Associate Editor of the IEEE TRANSACTIONS ON AUTOMATIC CONTROL from 1998 to 2000. He is a registered Professional Engineer in Alberta.



Yun Zou was born in Xian, China, in 1962. He received the B.Sc. degree in numerical mathematics from Northwestern University, Xian, in 1983 and the M.Tech. and Ph.D. degrees in automation in 1987 and 1990, respectively, from Nanjing University of Science and Technology (NUST), Nanjing, China.

Since July, 1990, he has been with NUST, where he is now a professor with the Department of Automation. His recent research interests include singular systems, MD systems, and transient stability of power systems.