

Robust H_∞ Filtering for LPV Discrete-Time State-Delayed Systems

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Abstract: This paper examines the problems of robust H_∞ filtering design for linear parameter-varying discrete-time systems with time-varying state delay. We present parameter-dependent robust H_∞ filters, which are derived using appropriately selected Lyapunov-Krasovskii functional. The resulting filters can be obtained from the solution of convex optimization problems in terms of parameterized linear matrix inequalities, which can be solved via efficient interior-point algorithms. The admissible filters guarantee a prescribed H_∞ noise attenuation level, relating exogenous signals to the estimation error for all possible parameters that vary in compact sets. A numerical example illustrates the feasibility of the proposed methodologies. [Nature and Science, 2004,2(2):36-44]

Key words: Robust H_∞ filtering; linear parameter-varying systems; parameterized linear matrix inequality; state-delay

1 Introduction

Stability analysis and control synthesis problems of linear parameter-varying (LPV) continuous-time systems where the state-space matrices depend on time-varying parameters, whose values are not known *a priori*, but can be measured in real time, have received considerable attention recently [1-6]. In contrast to continuous-time cases, discrete-time LPV systems [7-9] received relatively less attention despite their importance in digital control and signal processing applications.

On another front of systems control research, time delay often appears in many control systems either in the state, the control input, or the measurements. Time-delay is, in many cases, a source of instability. The stability issue and the performance of LPV systems with delay are, therefore, of theoretical and practical importance. Recently, much attention has been devoted to the analysis and synthesis problems of LPV time-delay systems. To mention a few, [5] investigated L_2 - L_2 control problems for LPV systems with parameter-varying delay and [6] proposed delay-independent and delay-dependent stability conditions for LPV systems with constant delay. However, it is worth noting that the filter obtained for LPV time-delay systems are still very limited, especially for LPV discrete time-delay systems.

H_∞ estimation has been widely studied during the past decades. One of its main advantages is the fact that it is insensitive to the exact knowledge of the statistics of the noise signals. This estimation procedure ensures that the L_2 -induced gain from the noise signals to the estimation error will be less than a prescribed level, where the noise signals are arbitrary energy-bounded signals. However, less study [8-10] has been done for the design of H_∞ filters for LPV systems.

This paper is interested in the H_∞ filtering problem for LPV discrete-time systems that include time-varying state delay. Using parameter-dependent Lyapunov-Krasovskii functional, we obtained a new H_∞ performance criterion that depended on parameter and the magnitude of delay-varying. Then we further modified the obtained criterion by adopting the idea [4, 11] of decoupling between the positive matrices and the system matrices, which is enabled by the introduction of addition slack variable to obtain another parameterized linear matrix inequalities (PLMIs) representation. The corresponding filter designs are finally cast into convex optimization problems, which can be solved via the efficient interior-point algorithms [12]. The obtained filters design procedure is shown, via a numerical example, to be effective.

The notation used throughout the paper is fairly standard. The superscript “ T ” stands for matrix transposition, R^n denotes the n dimensional Euclidean space, $R^{m \times n}$ is the set of all $m \times n$ real matrices, and the no-

tation $P > 0$ for $P \in R^{n \times n}$ means that P is symmetric and positive definite. In addition, in symmetric block matrices or long matrix expressions, we use $*$ as an ellipsis for the terms that are introduced by symmetry and $diag\{\cdot, \cdot\}$ stands for a block-diagonal matrix.

$$\begin{aligned} x(k+1) &= A(\rho(k))x(k) + A_d(\rho(k))x(k-d(k)) + B(\rho(k))\omega(k) \\ y(k) &= C(\rho(k))x(k) + C_d(\rho(k))x(k-d(k)) + D(\rho(k))\omega(k) \\ z(k) &= H(\rho(k))x(k) + H_d(\rho(k))x(k-d(k)) + L(\rho(k))\omega(k) \end{aligned} \quad (1)$$

where $x(k) \in R^n$ is the state; $y(k) \in R^m$ is the measured output; $z(k) \in R^p$ is the signal to be estimated; $\omega(k) \in R^l$ is the noise input; $\rho(k) = (\rho_1(k), \dots, \rho_s(k))$ is a vector of time-varying parameters which belongs to a compact set $\mathfrak{S} \in R^s$; $d(k) > 0$ is time-varying delay. It is assumed that there exist two positive constants d_m and d_M such that the following inequality holds

$$d_m \leq d(k) \leq d_M, \quad \forall k \geq 0 \quad (2)$$

The system matrices $A(\cdot)$, $A_d(\cdot)$, $B(\cdot)$, $C(\cdot)$, $C_d(\cdot)$, $D(\cdot)$, $H(\cdot)$, $H_d(\cdot)$, $L(\cdot)$ are known functions of $\rho(\cdot)$. For simplicity, ρ_k denotes the time-varying parameter vector $\rho(k)$ throughout the paper. Here we are interested in designing an estimator or full-order filter described by:

$$\begin{aligned} x_F(k+1) &= A_F(\rho_k)x_F(k) + B_F(\rho_k)y(k), \quad x_F(0) = 0 \\ z_F(k) &= C_F(\rho_k)x_F(k) + D_F(\rho_k)y(k) \end{aligned} \quad (3)$$

Augmenting the model of (1) to include the states of the filter, we obtain the filtering error system as follows:

$$\xi(k+1) = \bar{A}(\rho_k)\xi(k) + \bar{A}_d(\rho_k)K\xi(k-d(k)) + \bar{B}(\rho_k)\omega(k) \quad (4)$$

$$e(k) = \bar{C}(\rho_k)\xi(k) + \bar{C}_d(\rho_k)K\xi(k-d(k)) + \bar{D}(\rho_k)\omega(k)$$

where

$$\xi(k) = \{x^T(k), x_F^T(k)\}^T, \quad e(k) = z(k) - z_F(k) \quad (5)$$

$$\bar{A}(\rho_k) = \begin{bmatrix} A(\rho_k) & 0 \\ B_F(\rho_k)C(\rho_k) & A_F(\rho_k) \end{bmatrix},$$

2 Problem Formulation

Consider the following LPV discrete-time state-delayed system presented in state-space form by:

$$\begin{aligned} \bar{A}_d(\rho_k) &= \begin{bmatrix} A_d(\rho_k) \\ B_F(\rho_k)C_d(\rho_k) \end{bmatrix} \\ \bar{B}(\rho_k) &= \begin{bmatrix} B(\rho_k) \\ B_F(\rho_k)D(\rho_k) \end{bmatrix}, \\ \bar{C}(\rho_k) &= [H(\rho_k) - D_F(\rho_k)C(\rho_k) \quad -C_F(\rho_k)], \\ \bar{C}_d(\rho_k) &= H_d(\rho_k) - D_F(\rho_k)C_d(\rho_k), \\ \bar{D}(\rho_k) &= L(\rho_k) - D_F(\rho_k)D(\rho_k), \quad K = [I \quad 0] \end{aligned}$$

and I denotes an identity matrix with an appropriate dimension.

Our objective is to develop a robust H_∞ filter of the form (3) such that for all admissible parameter trajectories:

- (a) The filtering error system (4) is asymptotically stable.
 - (b) The filtering error system (4) guarantees, under zero-initial condition,
- $$\|e\|_2 \leq \gamma \|\omega\|_2 \quad (6)$$

for all nonzero $\omega \in l_2[0, \infty)$ and a given positive constant γ .

3 Robust H_∞ Filtering Analysis

In this section, we will derive new H_∞ performance criteria for filtering analyses and syntheses of system (1).

Theorem 1: Consider the system of (1). For a prescribed $\gamma > 0$, if there exist matrices

$$\begin{aligned} 0 &< Q^T = Q \in R^{n \times n}, \\ 0 &< P^T(\rho) = P(\rho) \in R^{2n \times 2n} \end{aligned}$$

that satisfy the following PLMI

$$\begin{bmatrix} -P(\rho_k) + d_\Delta K^T Q K & * & * & * & * \\ 0 & -Q & * & * & * \\ 0 & 0 & -\gamma^2 I & * & * \\ P(\rho_{k+1}) \bar{A}(\rho_k) & P(\rho_{k+1}) \bar{A}_d(\rho_k) & P(\rho_{k+1}) B(\rho_k) & -P(\rho_{k+1}) & * \\ \bar{C}(\rho_k) & \bar{C}_d(\rho_k) & \bar{D}(\rho_k) & 0 & -I \end{bmatrix} < 0 \quad (7)$$

for all nonzero $\omega \in l_2[0, \infty)$ and parameter trajectory, the filtering error system (4) is asymptotically stable with an H_∞ noise attenuation level γ . Where

$$d_\Delta = d_M - d_m + 1.$$

Proof: Construct a Lyapunov-Krasovskii functional as:

$$\begin{aligned} V(\xi(k)) &:= V_1 + V_2 + V_3 \\ V_1 &:= \xi^T(k) P(\rho_k) \xi(k) \end{aligned} \quad (8)$$

$$V_2 := \sum_{i=k-d(k)}^{k-1} \xi^T(i) K^T Q K \xi(i)$$

$$V_3 := \sum_{j=-d_M+2}^{-d_m+1} \sum_{i=k+j-1}^{k-1} \xi^T(i) K^T Q K \xi(i)$$

where $P(\rho_k) > 0, Q > 0$

Define $\Delta V := V(\xi(k+1)) - V(\xi(k))$, and along the trajectory of system (4) under the zero disturbance input, we have

$$\begin{aligned} \Delta V_1 &= \xi^T(k) [\bar{A}^T(\rho_k) P(\rho_{k+1}) \bar{A}(\rho_k) - P(\rho_k)] \xi(k) + 2 \xi^T(k) \bar{A}^T(\rho_k) P(\rho_{k+1}) \bar{A}_d(\rho_k) K \xi(k-d(k)) \\ &\quad + \xi^T(k-d(k)) K^T \bar{A}_d^T(\rho_k) P(\rho_{k+1}) \bar{A}_d(\rho_k) K \xi(k-d(k)) \end{aligned} \quad (9)$$

$$\begin{aligned} \Delta V_2 &= \sum_{i=k-d(k+1)+1}^k \xi^T(i) K^T Q K \xi(i) - \sum_{i=k-d(k)}^{k-1} \xi^T(i) K^T Q K \xi(i) \\ &= \xi^T(k) K^T Q K \xi(k) - \xi^T(k-d(k)) K^T Q K \xi(k-d(k)) \\ &\quad + \sum_{i=k-d(k+1)+1}^{k-1} \xi^T(i) K^T Q K \xi(i) - \sum_{i=k-d(k)+1}^{k-1} \xi^T(i) K^T Q K \xi(i) \end{aligned} \quad (10)$$

where

$$\sum_{i=k-d(k+1)+1}^{k-1} \xi^T(i) K^T Q K \xi(i) = \sum_{i=k-d_m+1}^{k-1} \xi^T(i) K^T Q K \xi(i) + \sum_{i=k-d(k+1)+1}^{k-d_m} \xi^T(i) K^T Q K \xi(i) \quad (11)$$

$$\sum_{i=k-d(k)+1}^{k-1} \xi^T(i) K^T Q K \xi(i) \geq \sum_{i=k-d_M+1}^{k-1} \xi^T(i) K^T Q K \xi(i) \quad (12)$$

Then

$$\begin{aligned} \Delta V_2 &= \xi^T(k) K^T Q K \xi(k) - \xi^T(k-d(k)) K^T Q K \xi(k-d(k)) + \sum_{i=k-d(k+1)+1}^{k-d_m} \xi^T(i) K^T Q K \xi(i) \\ &\leq \xi^T(k) K^T Q K \xi(k) - \xi^T(k-d(k)) K^T Q K \xi(k-d(k)) + \sum_{i=k-d_M+1}^{k-d_m} \xi^T(i) K^T Q K \xi(i) \end{aligned} \quad (13)$$

$$\begin{aligned} \Delta V_3 &= \sum_{j=-d_M+2}^{-d_m+1} \sum_{i=k+j}^k \xi^T(k) K^T Q K \xi(k) - \sum_{j=-d_M+2}^{-d_m+1} \sum_{i=k+j-1}^{k-1} \xi^T(k) K^T Q K \xi(k) \\ &= \sum_{j=-d_M+2}^{-d_m+1} \left[\xi^T(k) K^T Q K \xi(k) - \xi^T(k+j-1) K^T Q K \xi(k+j-1) \right] \\ &= (d_M - d_m) \xi^T(k) K^T Q K \xi(k) - \sum_{i=k-d_M+1}^{k-d_m} \xi^T(i) K^T Q K \xi(i) \end{aligned} \tag{14}$$

Therefore, from (9)-(14) we can obtain that $\Delta V = \Delta V_1 + \Delta V_2 + \Delta V_3 \leq \bar{\xi}^T(k) M \bar{\xi}(k)$, where

$$\begin{aligned} \bar{\xi}(k) &:= \begin{bmatrix} \xi^T(k) & \xi^T(k-d(k)) K^T \end{bmatrix}^T = \begin{bmatrix} \xi^T(k) & x^T(k-d(k)) \end{bmatrix}^T \\ M &:= \begin{bmatrix} \bar{A}^T(\rho_k) P(\rho_{k+1}) \bar{A}(\rho_k) - P(\rho_k) + d_\Delta K^T Q K & * \\ \bar{A}_d^T(\rho_k) P(\rho_{k+1}) \bar{A}(\rho_k) & \bar{A}_d^T(\rho_k) P(\rho_{k+1}) \bar{A}_d(\rho_k) - Q \end{bmatrix} \end{aligned}$$

Using the Schur complement [14], LMI (7) implies $M < 0$. Then from the Lyapunov-Krasovskii stability theorem, we can conclude that the filtering error system (4) is asymptotically stable.

Now, to establish the H_∞ performance for the filtering error system, assume zero-initial condition and consider the following index

$$J := \sum_{k=0}^{\infty} \left[e^T(k) e(k) - \gamma^2 \omega^T(k) \omega(k) \right] \tag{15}$$

Under zero initial condition, $V(\xi(k))|_{k=0} = 0$ and we have

$$\begin{aligned} J &\leq \sum_{k=0}^{\infty} \left[e^T(k) e(k) - \gamma^2 \omega^T(k) \omega(k) \right] + V(\xi(k))|_{k=\infty} - V(\xi(k))|_{k=0} \\ &= \sum_{k=0}^{\infty} \left[e^T(k) e(k) - \gamma^2 \omega^T(k) \omega(k) + \Delta V(\xi(k)) \right] = \sum_{k=0}^{\infty} \lambda^T(k) \Xi \lambda(k) \end{aligned} \tag{16}$$

where

$$\begin{aligned} \lambda(k) &:= \begin{bmatrix} \xi^T(k) & \xi^T(k-d(k)) K^T & \omega^T(k) \end{bmatrix}^T = \begin{bmatrix} \xi^T(k) & x^T(k-d(k)) & \omega^T(k) \end{bmatrix}^T \\ \Xi &:= \begin{bmatrix} \left(\bar{A}^T(\rho_k) P(\rho_{k+1}) \bar{A}(\rho_k) - P(\rho_k) \right) & * & * \\ \left(+d_\Delta K^T Q K + \bar{C}^T(\rho_k) \bar{C}(\rho_k) \right) & & \\ \left(\bar{A}_d^T(\rho_k) P(\rho_{k+1}) \bar{A}(\rho_k) \right) & \left(\bar{A}_d^T(\rho_k) P(\rho_{k+1}) \bar{A}_d(\rho_k) \right) & * \\ \left(+\bar{C}_d^T(\rho_k) \bar{C}(\rho_k) \right) & \left(-Q + \bar{C}_d^T(\rho_k) \bar{C}_d(\rho_k) \right) & \\ \left(\bar{B}^T(\rho_k) P(\rho_{k+1}) \bar{A}(\rho_k) \right) & \left(\bar{B}^T(\rho_k) P(\rho_{k+1}) \bar{A}_d(\rho_k) \right) & \left(\bar{B}^T(\rho_k) P(\rho_{k+1}) \bar{B}(\rho_k) \right) \\ \left(+\bar{D}^T(\rho_k) \bar{C}(\rho_k) \right) & \left(+\bar{D}^T(\rho_k) \bar{C}_d(\rho_k) \right) & \left(+\bar{D}^T(\rho_k) \bar{D}(\rho_k) - \gamma^2 I \right) \end{bmatrix} \end{aligned}$$

By Schur complement, PLMI (7) guarantees $\Xi < 0$, therefore $J \leq 0$ and $\|e\|_2 \leq \gamma \|\omega\|_2$. The proof is completed.

Remark 1: It should be noted that the conditions presented in Theorem 1 contain product terms between Lyapunov matrices and system matrices, such that condition (7) is a bilinear matrix inequality when (5) is

considered. In the following, we will present an

improved version of Theorem 1 by introducing a slack variable to decouple these product terms, which is more easily tractable for handling the filtering problems.

Theorem 2: Consider the system of (1).

For a prescribed $\gamma > 0$, if there exist matrices $0 < Q^T = Q \in R^{n \times n}$,

$0 < P^T(\rho_k) = P(\rho_k) \in R^{2n \times 2n}$, $W \in R^{2n \times 2n}$, satisfying

$$\begin{bmatrix} -P(\rho_k) + d_\Delta K^T Q K & * & * & * & * \\ 0 & -Q & * & * & * \\ 0 & 0 & -\gamma^2 I & * & * \\ W^T \bar{A}(\rho_k) & W^T \bar{A}_d(\rho_k) & W^T B(\rho_k) & P(\rho_{k+1}) - (W + W^T) & * \\ \bar{C}(\rho_k) & \bar{C}_d(\rho_k) & \bar{D}(\rho_k) & 0 & -I \end{bmatrix} < 0 \quad (17)$$

for all nonzero $\omega \in l_2[0, \infty)$ and parameter trajectory, the filtering error system (4) is asymptotically stable with an H_∞ noise attenuation level γ .

Proof: we will prove the theorem by showing the equivalence between (7) and (17). If (7) holds, (17) is readily established by choosing $W = W^T = P(\rho_{k+1})$. On the other hand, if (17) holds, we can explore the

facts $W^T + W - P(\rho_{k+1}) > 0$ so that W is a nonsingular matrix. In addition, we have

$$(P(\rho_{k+1}) - W)^T P^{-1}(\rho_{k+1})(P(\rho_{k+1}) - W) \geq 0,$$

which imply that

$$-W^T P^{-1}(\rho_{k+1})W \leq P(\rho_{k+1}) - W^T - W.$$

Therefore we can conclude from (17) that

$$\begin{bmatrix} -P(\rho_k) + d_\Delta K^T Q K & * & * & * & * \\ 0 & -Q & * & * & * \\ 0 & 0 & -\gamma^2 I & * & * \\ W^T \bar{A}(\rho_k) & W^T \bar{A}_d(\rho_k) & W^T B(\rho_k) & -W^T P(\rho_{k+1})W & * \\ \bar{C}(\rho_k) & \bar{C}_d(\rho_k) & \bar{D}(\rho_k) & 0 & -I \end{bmatrix} < 0 \quad (18)$$

Performing congruence transformation to (18) by $\text{diag}\{I, I, I, W^{-1}, I\}$ yields (7), and then the proof is completed.

In the LMI (17), W is additional slack variable, i.e., we don't set any restriction on these two matrices. In such a way, LMI (17) exhibits a kind of decoupling between the system matrices and the positive matrices (there is no product between them).

4 Robust H_∞ Filtering Design

In this section, based on Theorem 2, we will develop linear filter of form (3) assuring robust H_∞ performance for discrete-time state-delayed LPV system (1).

The following theorem provides sufficient conditions for the existence of delay-dependent H_∞ filters.

Theorem 3: Consider the system of (1). For a prescribed $\gamma > 0$, an admissible H_∞ filter of the form (3) exists for all nonzero $\omega \in l_2[0, \infty)$ and parameter trajectory, if there exist matrices $E \in R^{n \times n}$, $F \in R^{n \times n}$,

$$U \in R^{n \times n}, \bar{P}_1^T(\rho) = \bar{P}_1(\rho) \in R^{n \times n},$$

$$\bar{P}_3^T(\rho) = \bar{P}_3(\rho) \in R^{n \times n}, \bar{P}_2(\rho) \in R^{n \times n},$$

$$0 < Q^T = Q \in R^{n \times n}, \bar{A}_F(\rho) \in R^{n \times n},$$

$$\bar{B}_F(\rho) \in R^{n \times m}, \bar{C}_F(\rho) \in R^{p \times n} \text{ and}$$

$$\bar{D}_F(\rho) \in R^{p \times m} \text{ that satisfy the following inequalities}$$

$$\begin{bmatrix}
 -\bar{P}_1(\rho_k) + d_\Delta Q & * & * \\
 -\bar{P}_2^T(\rho_k) + d_\Delta Q & -\bar{P}_3(\rho_k) + d_\Delta Q & * \\
 0 & 0 & -Q \\
 0 & 0 & 0 \\
 E^T A(\rho_k) + \bar{B}_F(\rho_k)C(\rho_k) & E^T A(\rho_k) + \bar{B}_F(\rho_k)C(\rho_k) + \bar{A}_F(\rho_k) & E^T A_d(\rho_k) + \bar{B}_F(\rho_k)C_d(\rho_k) \\
 F^T A(\rho_k) & F^T A(\rho_k) & F^T A_d(\rho_k) \\
 H(\rho_k) - \bar{D}_F(\rho_k)C(\rho_k) & H(\rho_k) - \bar{D}_F(\rho_k)C(\rho_k) - \bar{C}_F(\rho_k) & H_d(\rho_k) - \bar{D}_F(\rho_k)C_d(\rho_k) \\
 * & * & * & * \\
 * & * & * & * \\
 * & * & * & * \\
 -\gamma^2 I & * & * & * \\
 E^T B(\rho_k) + \bar{B}_F(\rho_k)D(\rho_k) & \bar{P}_1(\rho_{k+1}) - E - E^T & * & * \\
 F^T A(\rho_k) & \bar{P}_2^T(\rho_{k+1}) - E - F^T - U & \bar{P}_3(\rho_{k+1}) - F - F^T & * \\
 L(\rho_k) - \bar{D}_F(\rho_k)D(\rho_k) & 0 & 0 & -I
 \end{bmatrix} < 0 \tag{19}$$

$$\begin{bmatrix}
 \bar{P}_1(\rho_k) & * \\
 \bar{P}_2^T(\rho_k) & \bar{P}_3(\rho_k)
 \end{bmatrix} > 0 \tag{20}$$

Proof: First let some matrix variables in Theorem 2 be partitioned as

$$W := \begin{bmatrix} W_{11} & W_{12} \\ W_{21} & W_{22} \end{bmatrix} \quad V := W^{-1} = \begin{bmatrix} V_{11} & V_{12} \\ V_{21} & V_{22} \end{bmatrix} \tag{21}$$

Introduce matrices

$$J_V := \begin{bmatrix} I & I \\ 0 & V_{21}V_{11}^{-1} \end{bmatrix} \tag{22}$$

Define $J_2 := \text{diag}\{J_V, I, I, J_V, I\}$ and introduce the following matrix variables

$$\begin{aligned}
 E &:= W_{11}, \quad F := V_{11}^{-1}, \quad U := V_{11}^{-T}V_{21}^T W_{21}, \quad \bar{A}_F(\rho) := W_{21}^T A_F(\rho)V_{21}V_{11}^{-1}, \quad \bar{B}_F(\rho) := W_{21}^T B_F(\rho), \\
 \bar{C}_F(\rho) &:= C_F(\rho)V_{21}V_{11}^{-1}, \quad \bar{D}_F(\rho) := D_F(\rho) \\
 \bar{P}(\rho) &:= \begin{bmatrix} \bar{P}_1(\rho) & * \\ \bar{P}_2^T(\rho) & \bar{P}_3(\rho) \end{bmatrix} = J_V^T \begin{bmatrix} P_1(\rho) & * \\ P_2^T(\rho) & P_3(\rho) \end{bmatrix} J_V
 \end{aligned} \tag{23}$$

Then performing congruence transformation to (17) by J_2 , it can be readily established that (19)-(20) are equivalent to (17).

Therefore, from Theorem 3 we can conclude that the filter with a state-space realization

$(A_F(\rho), B_F(\rho), C_F(\rho), D_F(\rho))$ defined in (23) guarantees that the filtering error system (4) has an H_∞ noise attenuation level γ .

Remark 2: Notice that the PLMI conditions (19)-(20) correspond to infinite-dimensional convex problems due to their parametric dependence. Using the gridding technique and the appropriate basis functions [3], infinite-dimensional PLMIs can be transformed to finite-dimensional ones, which can be solved numerically using convex optimization techniques. Hence, by choosing appropriate basis function $\{f_j(\rho)\}_{j=1}^S$ such that

$$\bar{P}(\rho) = \begin{bmatrix} \bar{P}_1(\rho) & * \\ \bar{P}_2^T(\rho) & \bar{P}_3(\rho) \end{bmatrix} = \begin{bmatrix} \sum_{j=1}^{n_f} f_j(\rho) \bar{P}_{1j} & * \\ \sum_{j=1}^{n_f} f_j(\rho) \bar{P}_{2j}^T & \sum_{j=1}^{n_f} f_j(\rho) \bar{P}_{3j} \end{bmatrix} > 0 \quad (24)$$

PLMIs can be approximated.

Remark 3: Theorem 3 casts the full-order robust H_∞ filtering problem for system (1) into PLMIs feasibility test, and any feasible solution to (19) and (24) will yield a suitable filter. If we can find admissible robust H_∞ filters for system (1) in the light of PLMIs (19) and (24) have feasible solutions, then the filter matrices can be calculated from the definition (23). However, there seem to be no systematic ways to de-

termine the matrices V_{21} and V_{22} needed for the filter matrices. To deal with such a problem, first of all, let us denote the filter transfer function from $y(k)$ to $z_F(k)$ by

$$T_{z_F y} = C_F(\rho)(zI - A_F(\rho))^{-1} B_F(\rho) + D_F(\rho) \quad (25)$$

Substituting the filter matrices with (23) and considering the relationship $U = V_{11}^{-T} V_{21}^T W_{21}$ yields

$$\begin{aligned} T_{z_F y} &= \bar{C}_F(\rho) F^{-1} V_{21}^{-1} (zI - W_{21}^{-T} \bar{A}_F(\rho) F^{-1} V_{21}^{-1})^{-1} W_{21}^{-T} \bar{B}_F(\rho) + \bar{D}_F(\rho) \\ &= \bar{C}_F(\rho) [z W_{21}^T V_{21} F - \bar{A}_F(\rho)]^{-1} \bar{B}_F(\rho) + \bar{D}_F(\rho) \\ &= \bar{C}_F(\rho) [zI - U^{-T} \bar{A}_F(\rho)]^{-1} U^{-T} \bar{B}_F(\rho) + \bar{D}_F(\rho) \end{aligned}$$

Therefore, an admissible filter is given by

$$A_F(\rho) = U^{-T} \bar{A}_F(\rho), B_F(\rho) = U^{-T} \bar{B}_F(\rho), C_F(\rho) = \bar{C}_F(\rho), D_F(\rho) = \bar{D}_F(\rho) \quad (26)$$

Remark 4: Note that (19) and (24) are PLMIs not only over the matrix variables, but also over the scalar γ^2 . This implies that the scalar γ^2 can be included as one of the optimization variables for LMI (19) and (23) to obtain the minimum noise attenuation level. Then

the minimum guaranteed cost of robust delay-dependent H_∞ filter can be readily found by solving the following convex optimization problem:

Minimize γ^2 subject to (19) and (23)

$$\text{over } E, U, F, \bar{A}_F(\rho), \bar{B}_F(\rho), \bar{C}_F(\rho), \bar{D}_F(\rho), \gamma^2, \bar{P}_1(\rho), \bar{P}_2(\rho), \bar{P}_3(\rho), Q \quad (27)$$

Remark 5: It can be shown that the time-varying delay of LPV system (1) is constant delay for $d_\Delta = 1$.

5 An Illustrative Example

Consider the following discrete-time LPV system with a state-delay.

$$\begin{aligned} x(k+1) &= \begin{bmatrix} 0 & 0.3 \\ -0.2 & 0.5\rho_1(k) \end{bmatrix} x(k) + \begin{bmatrix} 0 & 0 \\ 0.1 & 0.1\rho_2(k) \end{bmatrix} x(k-d(k)) + \begin{bmatrix} 0 \\ 1 \end{bmatrix} \omega(k) \\ y(k) &= [1 \quad 0] x(k) + [0.2 \quad 0] x(k-d(k)) + \omega(k) \\ z(k) &= [1 \quad 2] x(k) \end{aligned} \quad (28)$$

where $\rho_1 = \sin k$ and $\rho_2 = \cos k$ are time-varying parameters satisfying

$$-1 \leq \rho_1 \leq 1, -1 \leq \rho_2 \leq 1$$

Our objective is to design robust H_∞ controllers. First choose appropriate basis functions $f_1(\rho) = 1$, $f_2(\rho) = \rho_1$, $f_3(\rho) = \rho_2$

Gridding the parameter space uniformly using 9×9 grids. The minimum noise attenuation level obtained by solving convex optimization problem for different $d_\Delta = d_M - d_m + 1$ are shown in Table 1.

From Table 1, we can see that the effect of the delay-varying magnitude on the attainable the minimum

1	0.8169
2	1.1667
3	1.4454
4	1.6902

Table 1 The minimum guaranteed cost for different delay size

$d_\Delta = d_M - d_m + 1$	Minimum Guaranteed Cost γ^*
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guaranteed cost. The minimum noise attenuation level $\gamma^* = 1.6902$ for $d_\Delta = 4$, and the corresponding parameter-dependent filter matrices are given by

$$\begin{aligned}
 A_F(\rho) &= \begin{bmatrix} -0.0275 - 0.0609\rho_1 + 0.0055\rho_2 & 0.2507 + 0.0087\rho_1 + 0.0055\rho_2 \\ -1.2001 + 0.0257\rho_1 + 0.0011\rho_2 & -0.0154 + 0.4710\rho_1 - 0.0126\rho_2 \end{bmatrix} \\
 B_F(\rho) &= \begin{bmatrix} 0.0259 + 0.0841\rho_1 - 0.0129\rho_2 \\ 0.9973 - 0.0180\rho_1 + 0.0059\rho_2 \end{bmatrix}, \\
 C_F(\rho) &= [1.0042 + 0.0828\rho_1 - 0.0016\rho_2 \quad 2.0392 - 0.0030\rho_1 - 0.0240\rho_2], \\
 D_F(\rho) &= -0.0144 - 0.0685\rho_1 + 0.1098\rho_2
 \end{aligned} \tag{29}$$

The minimum noise attenuation level $\gamma^* = 0.8169$ for $d_\Delta = 1$, and the corresponding parameter-dependent filter matrices are given by

$$\begin{aligned}
 A_F(\rho) &= \begin{bmatrix} -0.0221 - 0.0160\rho_1 + 0.0011\rho_2 & 0.2878 + 0.0083\rho_1 + 0.0011\rho_2 \\ -0.1962 + 0.0059\rho_1 + 0.0002\rho_2 & -0.0187 + 0.4920\rho_1 - 0.0034\rho_2 \end{bmatrix}, \\
 B_F(\rho) &= \begin{bmatrix} 0.0197 + 0.0215\rho_1 - 0.0030\rho_2 \\ 0.9969 - 0.0039\rho_1 + 0.0014\rho_2 \end{bmatrix}, \\
 C_F(\rho) &= [1.0016 + 0.0245\rho_1 - 0.0009\rho_2 \quad 2.0125 - 0.0002\rho_1 - 0.0056\rho_2], \\
 D_F(\rho) &= -0.0047 - 0.0221\rho_1 + 0.0277\rho_2
 \end{aligned} \tag{30}$$

Then we analyze the disturbance attenuation level of the filtering error system by connecting the two obtained filters to the original system. Figures 1 and 2 present the simulation curves of estimating the signal $z(k)$ by the two filters respectively. Here we assume $\omega(k)$ to be

$$\omega(k) = \begin{cases} 2, & 20 \leq k \leq 30 \\ -2, & 50 \leq k \leq 60 \\ 0, & \text{else} \end{cases} \tag{31}$$

From the figure we can see that $\omega(k)$ drives $z_f(k)$ to deviate from $z(k)$. However, when $\omega(k)$ is zero, the deviation tends to be zero due to the asymptotically stability of the filter error system. Now we will further analyze the H_∞ performance. Fig. 3 and 4 give the changing curves of the disturbance signal and the filtering error signal. From (31) and Fig. 3, we obtain that

$$\begin{aligned}
 \|\omega\|_2 &= \sqrt{\sum_{k=0}^{\infty} \omega^T(k)\omega(k)} = 9.3808 \quad \text{and} \\
 \|e\|_2 &= \sqrt{\sum_{k=0}^{\infty} e^T(k)e(k)} = 1.5058, \text{ then it can be easily established that } \frac{\|e\|_2}{\|\omega\|_2} = 0.1605 < \gamma^* = 1.6902; \text{ And}
 \end{aligned}$$

Fig.4 shows that $\|e\|_2 = \sqrt{\sum_{k=0}^{\infty} e^T(k)e(k)} = 1.9536$, then it can be easily established that

$\frac{\|e\|_2}{\|\omega\|_2} = 0.2083 < \gamma^* = 0.8169$, therefore, the H_∞ filters can guarantee the prescribed noise disturbance attenuation level.

6 Concluding Remarks

In this paper, robust H_∞ filters design is proposed for LPV discrete-time systems with constant and time-varying state delay. The filtering problems have been solved and cast into convex optimization problems in terms of PLMIs, which can be solved via efficient interior-point algorithms. A numerical example has shown the feasibility applicability of the proposed designs.

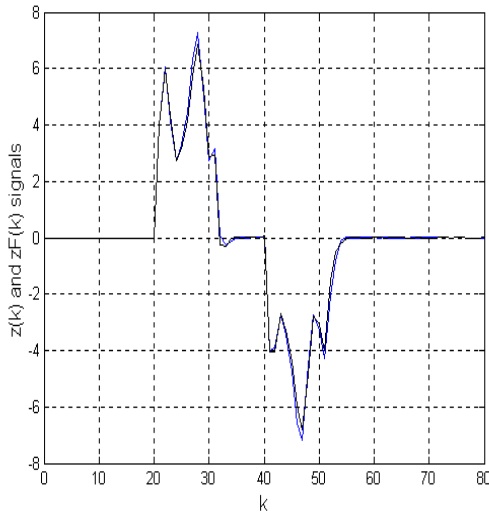


Figure 1 $z(k)$ and $z_F(k)$ signals of filter error system with time-varying state delay

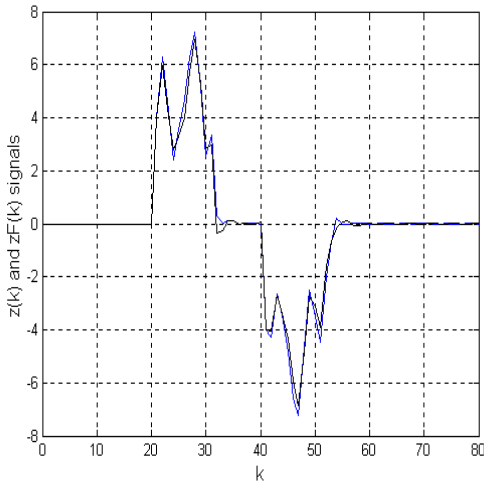


Figure 2 $z(k)$ and $z_F(k)$ signals of filter error system with constant delay

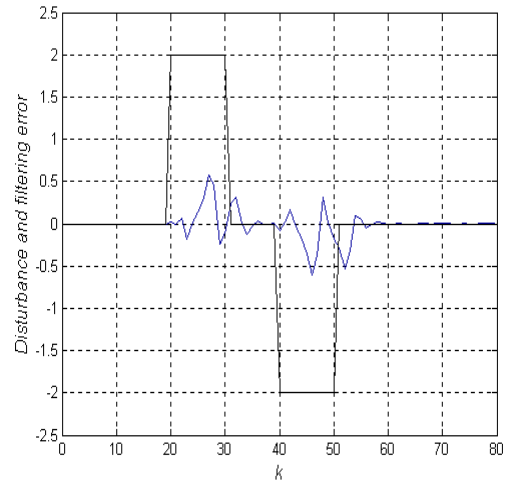


Figure 3 Disturbance and filtering error (Time-varying delay case)

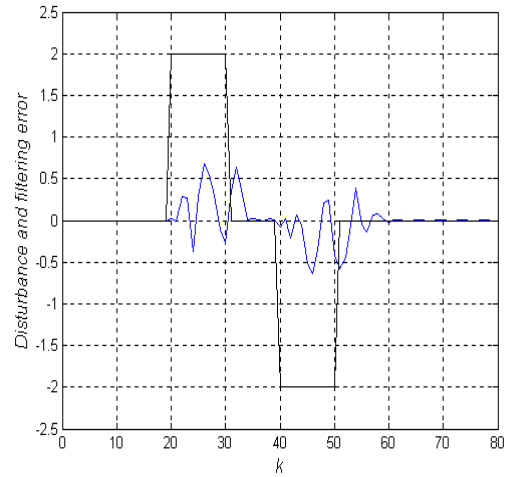


Figure 4 Disturbance and filtering error (Constant delay case)

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