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# 시간지연 퍼지 시스템의 보장비용 및 $H_\infty$ 필터링

## (Guaranteed Cost and $H_\infty$ Filtering for Delayed Fuzzy Dynamic Systems)

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### 요약

본 논문은 시간지연을 갖는 퍼지 시스템에 대한 보장 비용과  $H_\infty$  외란감쇄 성능을 갖는 퍼지 필터링 문제를 다룬다. 본 연구는 외란감쇄에 대한 확장  $L_2$  노름 제한조건과 LQ 비용함수의 성능 상한치 제한조건을 만족하는 필터링 설계 방법이다. Lyapunov 함수를 이용하여 필터의 존재성에 대한 충분조건을 유도하고 선형행렬부등식(LMI: linear matrix inequality)으로 나타낸다. 필터 설계는 LMI 해를 구함으로써 바로 구할 수 있다. 제안한 방법의 설계 과정을 설명하기 위한 시뮬레이션 예제를 또한 나타낸다.

### Abstract

This paper presents a method for designing guaranteed cost fuzzy filter with a desired  $H_\infty$  disturbance rejection constraint of delayed fuzzy dynamic systems. This method not only guarantees an induced  $L_2$  norm bound constraint on disturbance attenuation, but also minimizes an upper bound on a linear quadratic performance measure. A sufficient condition for the existence of guaranteed cost fuzzy filter with  $H_\infty$  constraint is then presented in terms of linear matrix inequalities(LMIs). A simulation example is given to illustrate the design procedures and performances of the proposed methods.

**Keywords** : guaranteed cost filtering,  $H_\infty$  filtering, delayed fuzzy systems

### 1. Introduction

In the past few years, there has been rapidly growing interest in fuzzy control of nonlinear

systems, and there have been many successful applications. Recently, stability analysis and systematic design are among the most important issues for fuzzy control systems and there have been significant research efforts on these issues<sup>[1-6]</sup>.

These methods are conceptually simple and straightforward. The nonlinear system is represented by a Takagi-Sugeno(T-S) fuzzy model. And then, the control design is carried out on the basis of the fuzzy model via the so-called parallel distributed compensation(PDC) scheme. Tanaka et al. [4],[5] presented stability analysis for a class of fuzzy dynamic systems. Ma et al. [6] presented the analysis and design of the fuzzy controller and fuzzy

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observer on the basis of T-S fuzzy model using separation property. Chen[9] and Lee et al.[10] presented the design method of fuzzy  $H_\infty$  controller to satisfy an  $H_\infty$  norm bound constraint on disturbance attenuation. Chen et al [12] also considered guaranteed-cost control with a desired  $H_\infty$  disturbance rejection constraint.

Since time delay is frequently a source of instability and encountered in various engineering systems, the  $H_\infty$  control problem for delayed systems has received considerable attention over the last few decades. In fuzzy control systems, Cao et al.[13] presented stability analysis and synthesis of delayed fuzzy dynamic systems. Lee et al. [14] also presented output feedback fuzzy  $H_\infty$  control of delayed fuzzy dynamic systems.

For practical control systems, a fuzzy control with guaranteed cost control performance under a desired  $H_\infty$  disturbance rejection constraint is more appealing for nonlinear systems.

This paper presents a method for designing guaranteed cost fuzzy filter with a desired  $H_\infty$  disturbance rejection constraint of delayed fuzzy dynamic systems. This methods not only guarantees an induced  $L_2$  norm bound constraint on disturbance attenuation, but also an upper bound on a linear quadratic performance measure. A sufficient condition for the existence of guaranteed cost fuzzy filter with  $H_\infty$  constraint is then presented in terms of linear matrix inequalities(LMIs). A simulation example is given to illustrate the design procedures and performances of the proposed methods.

## II. Problem formulation

The continuous fuzzy dynamic model, proposed by Takagi and Sugeno, is described by fuzzy IF-THEN rules which represented local linear input-output relations of nonlinear system. Consider a nonlinear system with time-varying delayed states that can be described by the following T-S fuzzy model with

time-varying delayed states:

Plant Rulei:

IF  $z_1(t)$  is  $M_{i1}$  and  $\dots$  and  $z_r(t)$  is  $M_{ir}$   
 THEN  $\dot{x}(t) = A_i x(t) + A_{di} x(t - d_1(t)) + B_i w(t)$

$$y(t) = C_i x(t) + C_{di} x(t - d_2(t)) + D_i w(t),$$

$$e_z(t) = C_{zi} x(t),$$

$$e_\infty(t) = C_{\infty i} x(t), \quad i = 1, 2, \dots, r$$

$$x(t) = \phi(t), \quad t \in [-\max(d_1(0), d_2(0)), 0] \tag{1}$$

where  $M_{ij}$  is the fuzzy set,  $x(t) \in R^n$  is the state vector,  $\phi(t) \in R^n$  is the continuous initial value function,  $w(t) \in R^p \in L_2(0, T)$  is the square-integrable noise signal,  $y(t) \in R^m$  is the measurement,  $e_L(t) \in R^{q_1}$ , and  $e_\infty(t) \in R^{q_2}$  is the signal to be estimated,  $r$  is the number of IF-THEN rules,  $z_1 \sim z_r$  are some measurable system variables, i.e., the premise variables, and all matrices are constant matrices with appropriate dimensions,  $d_i(t), i = 1, 2,$  are the time-varying delays with following assumptions:

$$0 \leq d_i(t) < \infty, \quad \dot{d}_i(t) \leq \beta_i < 1, \quad i = 1, 2. \tag{2}$$

Given a pair of  $(y(t), u(t))$  by using a center average defuzzifier, product inference, and singleton fuzzifier, the dynamic fuzzy model (1) can be expressed by the following global model:

$$\dot{x}(t) = \sum_{i=1}^r h_i(z(t)) \{A_i x(t) + A_{di} x(t - d_1(t)) + B_i w(t)\}$$

$$y(t) = \sum_{i=1}^r h_i(z(t)) \{C_i x(t) + C_{di} x(t - d_2(t)) + D_i w(t)\}$$

$$e_z(t) = \sum_{i=1}^r h_i(t) C_{zi} x(t)$$

$$e_\infty(t) = \sum_{i=1}^r h_i(t) C_{\infty i} x(t)$$

$$x(t) = \phi(t), \quad t \in [-\max(d_1(0), d_2(0)), 0] \tag{3}$$

where

$$w_i(z(t)) = \prod_{j=1}^n M_{ij}(z_j(t))$$

$$h_i(z(t)) = w_i(z(t)) / \sum_{j=1}^r w_j(z(t))$$

$$z(t) = [z_1(t) \ z_2(t) \ \dots \ z_r(t)]^T$$

where  $M_{ij}(z_j(t))$  is the grade of membership of  $z_j(t)$  in  $M_{ij}$ . It is assumed that

$$\begin{aligned} w_i(z(t)) &\geq 0, \quad i=1,2,\dots,r \\ \sum_{i=1}^r w_i(z(t)) &> 0 \end{aligned} \quad (4)$$

for all  $t$ . Then we can obtain the following conditions:

$$\begin{aligned} h_i(z(t)) &\geq 0, \quad i=1,2,\dots,r \\ \sum_{i=1}^r h_i(z(t)) &= 1 \end{aligned} \quad (5)$$

for all  $t$ . As a fuzzy  $H_\infty$  filter of the fuzzy system (1), we consider the following structure:

Filtering Rule i:

IF  $z_1(t)$  is  $M_{1j}$  and  $\dots$  and  $z_r(t)$  is  $M_{ir}$

THEN  $\hat{x}(t) = F_i \hat{x}(t) + G_i y(t), \quad i=1,2,\dots,r$

$$\hat{e}_z(t) = C_{zi} \hat{x}(t)$$

$$\hat{e}_\infty(t) = C_{\infty i} \hat{x}(t)$$

$$\hat{x}(0) = 0, \quad (6)$$

where the matrix  $F_i$  and  $G_i$  are to be determined. The final output of this fuzzy filter is

$$\hat{x}(t) = \sum_{i=1}^r h_i(z(t)) \{F_i \hat{x}(t) + G_i y(t)\}, \quad \hat{x}(0) = 0 \quad (7)$$

From (3) and (7), we obtain the following estimation error system

$$\begin{aligned} \zeta(t) &= \mathcal{A}(z)\zeta(t) + \mathcal{A}_1(z)x(t-d_1(t)) \\ &\quad + \mathcal{A}_2(z)x(t-d_2(t)) + \mathcal{B}(z)w(t) \end{aligned} \quad (8)$$

$$\zeta(t) = [\phi(t)^T \psi(t)^T]^T,$$

$$t \in [-\max(d_1(0), d_2(0)), 0]$$

$$\tilde{e}_z(t) = \mathcal{C}_z(z)\zeta(t)$$

$$\tilde{e}_\infty(t) = \mathcal{C}_\infty(z)\zeta(t)$$

with the following notations

$$\mathcal{A}(z) = \sum_{i=1}^r \sum_{j=1}^r h_i(z(t)) h_j(z(t)) \mathcal{A}_{ij}, \quad (9)$$

$$\mathcal{C}_z(z) = \sum_{i=1}^r h_i(z(t)) \mathcal{C}_{zi},$$

$$\mathcal{C}_\infty(z) = \sum_{i=1}^r h_i(z(t)) \mathcal{C}_{\infty i},$$

$$\mathcal{A}_1(z) = \sum_{i=1}^r h_i(z(t)) \mathcal{A}_{1i},$$

$$\mathcal{A}_2(z) = \sum_{i=1}^r \sum_{j=1}^r h_i(z(t)) h_j(z(t)) \mathcal{A}_{2ij},$$

$$\mathcal{B}(z) = \sum_{i=1}^r \sum_{j=1}^r h_i(z(t)) h_j(z(t)) \mathcal{B}_{ij},$$

where  $\zeta(t) = [(x(t) - \hat{x}(t))^T \quad x^T(t)]^T$ ,

$$\tilde{e}(t) = e(t) - \hat{e}(t),$$

$$\mathcal{A}_{ij} = \begin{bmatrix} F_i & A_i - G_i C_j - F_i \\ 0 & A_i \end{bmatrix}, \quad (10)$$

$$\mathcal{A}_{2ij} = \begin{bmatrix} -G_i C_{dj} \\ 0 \end{bmatrix},$$

$$\mathcal{A}_{1i} = [A_{di}^T \quad A_{di}^T]^T,$$

$$\mathcal{B}_{ij} = [(B_i - G_i D_j)^T \quad B_i^T]^T,$$

$$\mathcal{C}_{zi} = [C_{zi} \quad 0],$$

$$\mathcal{C}_{\infty i} = [C_{\infty i} \quad 0].$$

In general, the guaranteed cost filtering is more appealing to achieve a desired filtering performance. Therefore, we consider the following cost function without considering  $w(t)$ .

$$J_2 := \int_0^T \|\tilde{e}_z(t)\|^2 dt < \delta \quad (11)$$

Since  $H_\infty$  control is control design to efficiently eliminate the effect of  $w(t)$ , for given  $\gamma$  we define  $H_\infty$  control performance of the system (10) as the quantity

$$\begin{aligned} J_\infty := & \int_0^T \|\tilde{e}_\infty(t)\|^2 dt \leq \gamma^2 [ \int_0^T \|w(t)\|^2 dt + x(0)^T Q_0 x(0)^T \\ & + \sum_{i=1}^2 \int_{-d_i(0)}^0 x(\tau)^T Q x(\tau) d\tau ] \end{aligned} \quad (12)$$

for all  $T > 0$  and all  $w \in L_2[0, T]$ , where  $\|\cdot\|$  denotes the Euclidean norm. The weighting matrix

$Q_i, i=0,1,2$ , in (12) are measure of the initial state uncertainty at  $t \leq 0$  relative to the uncertainty in  $w(t)$ . A large value of  $Q_i$  indicates that the state at  $t \leq 0$  is very close to zero.

This paper addresses designing fuzzy filter (7) for the system (3) such that the estimation error system is globally exponentially stable and cost function (11) is satisfied with a desired  $H_\infty$  constraint (12).

### III. Fuzzy filter design

Define a Lyapunov functional

$$V(\zeta, t) = \zeta^T(t) P \zeta(t) + \sum_{i=1}^2 \int_{t-d_i(t)}^t x(\tau)^T S_i x(\tau) d\tau, \quad (13)$$

where  $P > 0$  and  $S_i > 0$ . Then there exist positive scalars  $\delta_1$  and  $\delta_2$  such that  $\delta_1 \|\zeta\|^2 \leq V(\zeta, t) \leq \delta_2 \|\zeta\|^2$ . If there exists scalar  $\alpha > 0$  such that  $V(\zeta, t) \leq -\alpha \|x\|$ , then the unforced system of (8) is globally exponentially stable<sup>[15]</sup>. From assumption (2)

$$\begin{aligned} V(\zeta, t) &\leq \zeta^T(t) P \zeta(t) + \zeta^T(t) P \zeta(t) \\ &+ \sum_{i=1}^2 \{x(t)^T S_i x(t) - x(t-d_i(t))^T (1-\beta_i) S_i x(t-d_i(t))\} \\ &:= V_a(\zeta, t), \end{aligned} \quad (14)$$

Lemma 1. Consider the system (3) with assumption (2), and let  $R_i > 0, i=0,1,2$  be a given initial state weighting matrix and  $\gamma > 0$  a given scalar. If there exist matrices  $P > 0, S_1 > 0$  and  $S_2 > 0$ , and positive scalar  $\alpha$  satisfying the following inequalities

$$\Omega_{ii} < 0, \quad i=1,2, \dots, r \quad (15)$$

$$\Omega_{ij} + \Omega_{ji} < 0, \quad i < j < r, \quad (16)$$

$$[I_n \ I_n] P [I_n \ I_n]^T - \gamma^2 Q_0 < 0 \quad (17)$$

$$S_i - \gamma^2 Q_i < 0, \quad i=1,2 \quad (18)$$

then the corresponding estimation error system is globally exponentially stable and achieves  $H_\infty$  control performance for all  $w \in L_2[0, T]$ . In here

$\Omega_{ij} =$

$$\begin{bmatrix} \mathcal{A}_y^T P + P \mathcal{A}_y + \mathcal{S} + S_a + \mathcal{C}_{\infty i}^T \mathcal{C}_{\infty i} & P \mathcal{A}_{1i} & P \mathcal{A}_{2ij} & P B_y \\ * & -\mathcal{S}_1 & 0 & 0 \\ * & * & -\mathcal{S}_2 & 0 \\ * & * & * & -\gamma^2 I \end{bmatrix}, \quad (19)$$

where  $*$  represents the elements below the main diagonal of a symmetric matrix and

$$\begin{aligned} \mathcal{S} &= \begin{bmatrix} 0 & 0 \\ 0 & S_1 + S_2 \end{bmatrix}, \quad S_a = \begin{bmatrix} 0 & 0 \\ 0 & \alpha I \end{bmatrix} \\ \mathcal{S}_i &= (1-\beta_i) S_i, \quad i=1,2. \end{aligned} \quad (20)$$

Proof : Consider  $V_a(\zeta, t) \leq -\alpha \|x\|^2$  and the following condition

$$J_a(t) := V_a(\zeta, t) + \tilde{e}^T(t) \tilde{e}(t) - \gamma^2 w^T(t) w(t) \leq 0. \quad (21)$$

From closed loop system (8)

$$\begin{aligned} J_a(t) &\leq \sum_{i=1}^r \sum_{j=1}^r h_i h_j \begin{bmatrix} \zeta(t) \\ x(t-d_1) \\ x(t-d_2) \\ w(t) \end{bmatrix} \\ &\begin{bmatrix} \mathcal{A}_y^T P + P \mathcal{A}_y + \mathcal{S} + S_a & P \mathcal{A}_{1i} & P \mathcal{A}_{2ij} & P B_y \\ * & -\mathcal{S}_1 & 0 & 0 \\ * & * & -\mathcal{S}_2 & 0 \\ * & * & * & -\gamma^2 I \end{bmatrix} \begin{bmatrix} \zeta(t) \\ x(t-d_1) \\ x(t-d_2) \\ w(t) \end{bmatrix} \\ &+ \sum_{i=1}^r \sum_{j=1}^r h_i h_j (\mathcal{C}_{\infty i} \zeta(t))^T \mathcal{C}_{\infty i} \zeta(t). \end{aligned} \quad (22)$$

By considering the following condition

$$\begin{aligned} \tilde{e}(t)^T e(t) &= \sum_{i=1}^r \sum_{j=1}^r h_i h_j (\mathcal{C}_{\infty i} \zeta(t))^T \mathcal{C}_{\infty j} \zeta(t) \\ &\leq \sum_{i=1}^r h_i \zeta^T(t) \mathcal{C}_{\infty i}^T \mathcal{C}_{\infty i} \zeta(t) \end{aligned} \quad (23)$$

(22) can be presented as follows

$$J_a(t) \leq \sum_{i=1}^r \sum_{j=1}^r h_i h_j \begin{bmatrix} \zeta(t) \\ x(t-d_1) \\ x(t-d_2) \\ w(t) \end{bmatrix} \Omega_{ij} \begin{bmatrix} \zeta(t) \\ x(t-d_1) \\ x(t-d_2) \\ w(t) \end{bmatrix}. \quad (24)$$

The inequality  $\sum_{i=1}^r \sum_{j=1}^r h_i h_j \Omega_{ij} < 0$  is equivalent to

$$\sum_{i=1}^r h_i(z(t)) h_i(z(t)) \Omega_{ii} + \sum_{i \neq j}^r h_i(z(t)) h_j(z(t)) (\Omega_{ij} + \Omega_{ji}) \leq 0. \quad (25)$$

Thus if (15) and (16) are satisfied, then

$$V_a(\xi, t) \leq \alpha \|x\|^2 \text{ and (21) is satisfied. From (21)}$$

$$\begin{aligned} & \int_0^T \|\tilde{e}(t)\|^2 dt - \gamma^2 \int_0^T \|w(t)\|^2 dt \leq \xi^T(0) P \xi(0) \\ & + \sum_{i=1}^2 \int_{-d_i(0)}^0 x^T(\tau) S_i x(\tau) d\tau \end{aligned} \quad (26)$$

because  $V(\xi, T) > 0$ . It follows from initial condition (8) that

$$\begin{aligned} & \int_0^T \|\tilde{e}(t)\|^2 dt - \gamma^2 \left[ \int_0^T \|w(t)\|^2 dt + x_0^T Q_0 x_0 \right. \\ & + \sum_{i=1}^2 \int_{-d_i(0)}^0 x^T(\tau) Q_i x(\tau) d\tau \left. \right] \leq \xi^T(0) P \xi(0) \\ & - \gamma^2 x_0^T Q_0 x_0 + \sum_{i=1}^2 \int_{-d_i(0)}^0 x^T(\tau) S_i x(\tau) d\tau \\ & - \gamma^2 \sum_{i=1}^2 \int_{-d_i(0)}^0 x^T(\tau) Q_i x(\tau) d\tau \leq x_0^T \\ & \{ [I_n \ I_n] P [I_n^T \ I_n^T]^T - \gamma^2 Q_0 \} x_0 \\ & + \sum_{i=1}^2 \int_{-d_i(0)}^0 x^T(\tau) [S_i - \gamma^2 Q_i] x(\tau) d\tau. \end{aligned} \quad (27)$$

Thus if (15)–(18) are satisfied,  $H_\infty$  control performance (12) is satisfied.  $\square$

Lemma 2. Consider the system (3) with assumption (2). If there exist matrices  $P > 0$ ,  $S_1 > 0$ ,  $S_2 > 0$ ,  $R_1 > 0$ ,  $R_2 > 0$  and positive scalar  $\rho$  satisfying the following inequalities

$$\Psi_{ii} < 0, \quad i = 1, 2, \dots, r \quad (28)$$

$$\Psi_{ij} + \Psi_{ji} < 0, \quad i < j < r, \quad (29)$$

$$-\rho + \xi^T(0) P \xi(0) < 0 \quad (30)$$

$$-R_i + N_i^T S_i N_i < 0, \quad i = 1, 2 \quad (31)$$

then the corresponding estimation error system achieves guaranteed cost function  $\delta$  and the guaranteed cost bound  $\delta$  is  $\rho + \text{tr}(R_1) + \text{tr}(R_2)$ .

In here

$$\Psi_{ij} = \begin{bmatrix} \bar{A}_{ij}^T P + P \bar{A}_{ij} + \mathfrak{S} & \mathfrak{C}_{zi}^T \mathfrak{C}_{zi} & P \bar{A}_{1i} & P \bar{A}_{2ij} \\ * & -\mathfrak{S}_i & 0 & \\ * & * & * & -\mathfrak{S}_2 \end{bmatrix}, \quad (32)$$

$$\int_{-d_i(0)}^0 \phi(\tau) \phi(\tau)^T d\tau = N_i N_i^T, \quad i = 1, 2 \quad (33)$$

where

$$\mathfrak{S} = \begin{bmatrix} 0 & 0 \\ 0 & S_1 + S_2 \end{bmatrix}, \quad \mathfrak{S}_i = (1 - \beta_i) S_i, \quad i = 1, 2, \quad (34)$$

and  $*$  represents the elements below the main diagonal of a symmetric matrix.

Proof : From Lyapunov functional (13), we obtain

$$\begin{aligned} J_2 &= \int_0^{t_i} \{ \tilde{e}_z^T(t) \tilde{e}_z(t) \} dt \\ &= V(0) - V(t_i) + \int_0^{t_i} [ \tilde{e}_z^T(t) \tilde{e}_z(t) + V(t) ] dt \\ &\leq V(0) + \int_0^{t_i} [ \tilde{e}_z^T(t) \tilde{e}_z(t) + V(t) ] dt \end{aligned} \quad (35)$$

If

$$V(t) + \tilde{e}_z^T(t) \tilde{e}_z(t) < 0 \quad (36)$$

we obtain the upper bound of the cost function  $J_2$  as follows

$$\begin{aligned} J_2 &\leq V(0) = \xi^T(0) P \xi(0) + \int_{-d_1(0)}^0 x^T(\tau) S_1 x(\tau) d\tau \\ &+ \int_{-d_2(0)}^0 x^T(\tau) S_2 x(\tau) d\tau. \end{aligned} \quad (37)$$

The condition (36) is equivalent to the existence of  $P > 0$  satisfying

$$\begin{aligned} & \sum_{i=1}^r \sum_{j=1}^r h_i h_j \begin{bmatrix} \xi(t) \\ x(t-d_1) \\ x(t-d_2) \end{bmatrix} \\ & \begin{bmatrix} \bar{A}_{ij}^T P + P \bar{A}_{ij} + \mathfrak{S} & P \bar{A}_{1i} & P \bar{A}_{2ij} \\ * & -\mathfrak{S}_{11} & 0 \\ * & * & -\mathfrak{S}_{22} \end{bmatrix} \begin{bmatrix} \xi(t) \\ x(t-d_1) \\ x(t-d_2) \end{bmatrix} \\ & + \sum_{i=1}^r \sum_{j=1}^r h_i h_j (\mathfrak{C}_{zi} \xi(t))^T \mathfrak{C}_{zj} \xi(t) < 0. \end{aligned} \quad (38)$$

By considering the following condition

$$\begin{aligned} & \sum_{i=1}^r \sum_{j=1}^r h_i h_j (\mathcal{C}_{zi} \zeta(t))^T \mathcal{C}_{zj} \zeta(t) \\ & \leq \sum_{i=1}^r h_i \zeta^T(t) \mathcal{C}_{zi}^T \mathcal{C}_{zi} \zeta(t) \end{aligned} \quad (39)$$

it can be easily shown that the inequality (38) is equivalent to

$$\sum_{i=1}^r h_i(z(t)) h_i(z(t)) \Psi_{ii} + \sum_{i < j}^r h_i(z(t)) h_j(z(t)) (\Psi_{ij} + \Psi_{ji}) \leq 0. \quad (40)$$

From (40) we got (28) and (29). The (30) and (31) are related to the upper bound of  $J_2$  performance measure. The first term of right hand side in (37) is equivalent to (30). The second term of right hand side in (37) has the following relations

$$\begin{aligned} & \int_{-d_i(0)}^0 \psi^T(\tau) S_1 \psi(\tau) d\tau = \int_{-d_i(0)}^0 \text{tr}\{\psi(\tau)^T S_1 \psi(\tau)\} d\tau \\ & = \text{tr}\{N_1 N_1^T S_1\} = \sum_{i=1}^2 \text{tr}\{N_i^T S_1 N_i\} < \text{tr}\{R_1\}. \end{aligned} \quad (41)$$

This is equivalent  $-R_1 + N_1^T S_1 N_1 < 0$ . The third term of right hand side in (37) has the following relations

$$\begin{aligned} & \int_{-d_i(0)}^0 \psi^T(\tau) S_2 \psi(\tau) d\tau = \int_{-d_i(0)}^0 \text{tr}\{\psi(\tau)^T S_2 \psi(\tau)\} d\tau \\ & = \text{tr}\{N_2 N_2^T S_2\} = \text{tr}\{N_2^T S_2 N_2\} < \text{tr}\{R_2\}. \end{aligned} \quad (42)$$

From (41) and (42), (31) is obtained and  $\rho + \text{tr}(R_1) + \text{tr}(R_2)$  is an upper bound of guaranteed cost.  $\square$

The next Theorem 1 presents a solution to the guaranteed cost filtering problem with  $H_\infty$  constrain for the delayed fuzzy model in terms of LMIs from lemma 1 and lemma 2.

Theorem 1. Consider the system (3), and let  $R_i > 0, i = 0, 1, 2$  be a given initial state weighting matrix  $\gamma > 0$  a given scalar. If there exist common matrix  $X > 0, Z > 0, S_1 > 0, S_2 > 0, R_1 > 0, R_2 > 0$  and matrices  $W_i, Y_i, i = 1, 2, \dots, r$ , and positive scalar  $\alpha, \rho$  satisfying the following LMIs:

$$\Phi_{ii} < 0, \quad \Lambda_{ii} < 0, \quad i = 1, 2, \dots, r, \quad (43)$$

$$\Phi_{ij} + \Phi_{ji} < 0, \quad \Lambda_{ij} + \Lambda_{ji} < 0, \quad i < j < r, \quad i = 1, 2, \dots, r \quad (44)$$

$$-\rho + \phi^T(0) X \phi(0) + \phi^T(0) Z \phi(0) < 0, \quad (45)$$

$$-R_i + N_i^T S_i N_i < 0, \quad i = 1, 2 \quad (46)$$

$$X + Z - \gamma^2 Q_0 \leq 0 \quad (47)$$

$$S_i - \gamma^2 Q_i \leq 0, \quad i = 1, 2, \quad (48)$$

then there exists a fuzzy filter (7) such that the estimation error system (8) achieves the guaranteed cost function (11) and  $H_\infty$  constraint (12).

In here

$$\delta = \rho + \text{tr}(R_1) + \text{tr}(R_2),$$

$$\Phi_{ii} =$$

$$\begin{bmatrix} W_i + W_i^T + C_{zi}^T C_{zi} & XA_i - Y_i C_i - W_i & XA_{di} & -Y_i C_{di} & XB_i - Y_i D_i \\ XA_i - Y_i C_i - W_i & \Delta_i + \alpha I & ZA_{di} & 0 & ZB_i \\ A_{di}^T X & A_{di}^T Z & -S_1 & 0 & 0 \\ -C_{di}^T Y_i & 0 & 0 & -S_2 & 0 \\ B_i^T X - D_i^T Y_i & B_i^T Z & 0 & 0 & -\gamma^2 I \end{bmatrix}. \quad (49)$$

$$\Lambda_{ii} = \begin{bmatrix} W_i + W_i^T + C_{zi}^T C_{zi} & XA_i - Y_i C_i - W_i & XA_{di} & -Y_i C_{di} \\ XA_i - Y_i C_i - W_i & \Delta_i & ZA_{di} & 0 \\ A_{di}^T X & A_{di}^T Z & -S_1 & 0 \\ -C_{di}^T Y_i & 0 & 0 & -S_2 \end{bmatrix}. \quad (50)$$

where

$$\Delta_i = ZA_i + A_i^T Z + S_1 + S_2, \quad S_r = (1 - \beta_r) S_r, \quad r = 1, 2. \quad (51)$$

Furthermore, filter gains are given by

$$F_i = X^{-1} W_i, \quad G_i = X^{-1} Y_i. \quad (52)$$

Proof: Let  $P = \text{diag}\{X, Z\}$ , where  $X$  and  $Z$  are symmetric positive definite matrices to be found. Denoting  $W_i = X F_i$ , and  $Y_i = X G_i, i = 1, 2, \dots, r$ , and considering (10), we obtained (43)-(48) from lemma 1 and lemma 2.  $\square$

It has been seen that the filter design problem of the fuzzy system (3) can be transformed into a linear algebra problem. This set of LMIs constitutes

a finite-dimensional convex feasibility problem. There are several efficient algorithms to solve the above convex LMIs problem<sup>[17,18]</sup>

#### IV. Design example

We will design a fuzzy filter for the following nonlinear system:

$$\begin{aligned}\dot{x}_1(t) &= -5.125x_1(t) - 0.5x_1(t-d(t)) - 2x_2(t) \\ &\quad - 6.7x_2^3(t) - 0.2x_2(t-d(t)) \\ &\quad - 0.67x_2^3(t-d(t)) + w(t) \\ \dot{x}_2(t) &= x_1(t) \\ y(t) &= x_2(t) - 0.01x_2(t-d(t)) + 0.1w(t) \\ e(t) &= 2x_1(t)\end{aligned}\quad (53)$$

where time-varying delay and initial state is

$$\begin{aligned}d(t) &= 1 + 0.5 \cos(0.1t) \\ \phi(t) &= [e^t \quad -e^t]^T, t \leq 0\end{aligned}\quad (54)$$

$x_1(t)$  is estimated using a fuzzy filter and assume that  $x_2(t)$  is observable. It is also assumed that

$$x_1(t) \in [-1.5 \quad 1.5], \quad x_2(t) \in [-1.5 \quad 1.5]. \quad (55)$$

Using the same procedure as in [6], the nonlinear term can be represented as

$$-6.7x_2^3(t) = M_{11} \cdot 0 \cdot x_2(t) - (1 - M_{11}) \cdot 15.075 x_2(t). \quad (56)$$

By solving the equation,  $M_{11}$  is obtained as follows:

$$\begin{aligned}M_{11}(x_2(t)) &= 1 - \frac{x_2(t)^2}{2.25} \\ M_{12}(x_2(t)) &= \frac{x_2(t)^2}{2.25}.\end{aligned}\quad (57)$$

$M_{11}$  and  $M_{12}$  can be interpreted as membership functions of fuzzy set. By using these fuzzy sets, the nonlinear system can be presented by the following T-S fuzzy model

Plant Rule 1: IF  $x_2(t)$  is  $M_{11}$  THEN}

$$\begin{aligned}\dot{x}(t) &= A_1x(t) + A_{d1}x(t-d(t)) + B_1w(t) \\ y(t) &= C_1x(t) + C_{d1}x(t-d(t)) + D_1w(t) \\ e_z(t) &= C_{z1}x(t), \\ e_\infty(t) &= C_{\infty1}x(t)\end{aligned}\quad (58)$$

Plant Rule 2: IF  $x_2(t)$  is  $M_{12}$  THEN}

$$\begin{aligned}\dot{x}(t) &= A_2x(t) + A_{d2}x(t-d(t)) + B_2w(t) \\ y(t) &= C_2x(t) + C_{d2}x(t-d(t)) + D_2w(t) \\ e_z(t) &= C_{z2}x(t) \\ e_\infty(t) &= C_{\infty2}x(t)\end{aligned}\quad (59)$$

where  $x(t) = [x_1(t) \quad x_2(t)]^T$ ,

$$\begin{aligned}A_1 &= \begin{bmatrix} -5.125 & -2 \\ 1 & 0 \end{bmatrix}, \quad A_{d1} = \begin{bmatrix} -0.5 & -0.2 \\ 0 & 0 \end{bmatrix}, \\ A_2 &= \begin{bmatrix} -5.125 & -17.075 \\ 1 & 0 \end{bmatrix}, \quad A_{d2} = \begin{bmatrix} -0.5 & -1.71 \\ 0 & 0 \end{bmatrix}\end{aligned}\quad (60)$$

$$B_1 = B_2 = [1 \quad 0]^T, \quad C_1 = C_2 = [0 \quad 1],$$

$$C_{d1} = C_{d2} = [0 \quad -0.01], \quad D_1 = D_2 = 0.1,$$

$$C_{\infty1} = C_{\infty2} = \text{diag}[1.5 \quad 1.5], \quad C_{z1} = C_{z2} = [1 \quad 0].$$

$$\text{Let } \gamma = 3, \quad \beta_1 = \beta_2 = 0.05, \quad R_0 = \text{diag}[2.4 \quad 2.4],$$

$R_1 = R_2 = \text{diag}[2.7, 2.7]$ , then filter gains obtained from Theorem 1 are

$$\begin{aligned}F_1 &= \begin{bmatrix} -6.2112 & -3.8335 \\ -1.9171 & -29.2087 \end{bmatrix}, \\ F_2 &= \begin{bmatrix} -4.9771 & -4.0592 \\ -2.1546 & -29.4116 \end{bmatrix}, \\ G_1 &= \begin{bmatrix} -2.7060 \\ 29.1623 \end{bmatrix}, \quad G_2 = \begin{bmatrix} -15.1443 \\ 20.3392 \end{bmatrix}\end{aligned}\quad (61)$$

The upper bound of guaranteed cost is 72.8945. The simulation results of nonlinear systems with time-varying delays are shown in <Fig. 1(a)> and <Fig. 1(b)>, respectively.

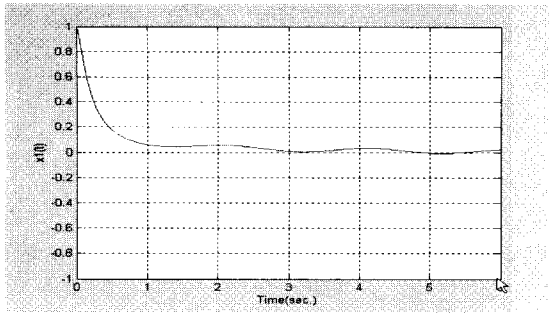
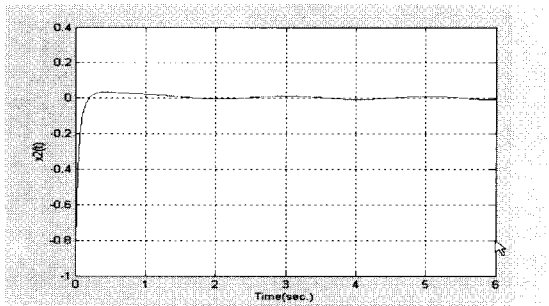
(a)  $x_1(t) - \hat{x}_1(t)$ (b)  $x_2(t) - \hat{x}_2(t)$ 

그림 1. 시간지연 비선형시스템에 대한 시뮬레이션 결과  
Fig. 1. The simulation result of nonlinear system with time delays.

For these simulation, the noise signal  $w(t)$  is

$$w(t) = 0.1 \cdot \cos(\pi t) \quad (62)$$

and the initial value of the state is assumed by

$$[x_1^T(t) \ x_2^T(t)]^T = [1 \ -1]^T, \quad t \leq 0. \quad (63)$$

The designed fuzzy filter estimates the states of the nonlinear system without the steady state errors.

## V. Conclusion

In this paper, we have developed guaranteed cost fuzzy  $H_\infty$  filter design method for delayed fuzzy dynamic systems. We have obtained sufficient conditions for the existence of fuzzy filters such that the estimation error system is globally exponentially stable and achieves guaranteed cost and  $H_\infty$  performance simultaneously. The filter design has utilized the concept of parallel distributed

compensation and the filter gains can also be directly obtained from the LMI solutions.

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