## An $H_{\infty}$ Output Feedback Control for Uncertain Singularly Perturbed T-S Fuzzy Systems

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#### Abstract

This paper deals with an  $H_{\infty}$  output feedback controller design for uncertain singularly perturbed T-S fuzzy systems. Integral quadratic constraints are used to describe various kinds of uncertainties of the plant. It is shown that the  $H_{\infty}$  norm of the uncertain singularly perturbed fuzzy system is less than  $\gamma$  for a sufficiently small  $\varepsilon > 0$  if the  $H_{\infty}$  norms of both the slow and fast subsystem are less than  $\gamma$ . Using this fact, we develop a linear matrix inequality based design method which is independent of the singular perturbation parameter  $\varepsilon$ . A numerical example is provided to demonstrate the efficacy of the proposed design method.

**Key Words**: Uncertain Singularly Perturbed Systems, Linear Matrix Inequality,  $H_{\infty}$  Fuzzy Control, Integral Quadratic Constraints, T-S Fuzzy System

### 1. 서 론

There are two main difficulties in controller design for real systems: nonlinearity and uncertainty. This paper deals with an  $H_{\infty}$  output feedback controller design for uncertain singularly perturbed nonlinear systems. The system with both slow mode and fast mode in system dynamics can be modelled as a singularly perturbed system. For example, a mechanical system with very small mass or inertia moment, dynamic system with a fast actuator and a control system with a high gain feedback system can be modelled as a singularly perturbed system.[1]-[3] Uncertain singularly perturbed system is a singularly perturbed system with uncertain factors. Originally the singular perturbation theory has been widely used to find an approximate solution in a field celestial mathematical of mechanics and hydrodynamics. In control problem, P.Kokotovic[2] first used the singular perturbation theory in the optimal control problem to obtain an approximate solution of the Riccati differential equation. After that this effective way to design the controller for linear or nonlinear singularly perturbed system has been used by many researchers.

In past three decades, the  $H_{\infty}$  control problem has been extensively studied by a number of researchers.[4]-[10] This problem can be stated as follows: given a dynamic system with the exogenous input and measured output, design a control law such that the  $L_2$  gain of the operator from the exogenous input to the

접수일자 : 2009년 10월 21일 완료일자 : 2009년 12월 5일 이 논문은 2007학년도 대구대학교 학술연구비 지원에 의한 논문임. regulated output is minimized or no larger than some prescribed level. In nonlinear system case, it is very difficult to solve nonlinear  $H_{\infty}$  control problem. The reason is that it is still very difficult to obtain the solution for the resulting Hamilton–Jacobi–Bellman (HJB) equation.

In the past two decades, there has been rapidly growing interest in approximating a nonlinear system by the Takagi-Sugeno(T-S) fuzzy model.[11]-[13]. Based on this fuzzy model, a model-based fuzzy control has been developed to stabilize the T-S fuzzy model. The T-S fuzzy model provides a powerful tool for modelling complex nonlinear systems. Unlike the conventional model where a single model is used to describe the global behavior of a system, the T-S model is essentially a multi-model approach in which simple sub-models (typically linear models) are combined to describe the global behavior of the system.

In this work, we suggest an output feedback  $H_{\infty}$  controller design method for a singularly perturbed T-S fuzzy system with integral quadratic constraints(IQC's). Many kinds of uncertainties such as time varying parameters, time-delay and unmodelled dynamics etc. can be described via IQC's. Various kinds of uncertainties which can be described via IQC's are presented in [14].

In section 2, we define the uncertain singularly perturbed system and formulate a control problem. In section 3, an  $H_{\infty}$  norm bounded condition for an uncertain singularly perturbed system with IQC's is presented in the form of linear matrix inequalities(LMI's). In section 4, we present an output  $H_{\infty}$  controller design method for a T-S fuzzy system with IQC's. In section 5, a numerical design example is given to demonstrate the efficacy of our method.

The notation is standard. Given matrix A,  $A^T$  is

transposed matrix and  $A^{-1}$  is the inverse matrix of A.  $A^{\perp}$  is a basis vector for null space of  $A^{T}$ . A > 0 means A is a positive definite matrix.  $R^{n}$  is an n dimensional real vector space,  $R^{n \times m}$  is an  $n \times m$  dimensional real matrix set.  $I_{n}$  is the  $n \times n$  unit matrix, and 0 is the zero matrix. diag(A,B) is a diagonal matrix composed by Aand B.

#### II. System Description and Definition

Consider the system described by the following singularly perturbed T-S fuzzy model:

Plant Rule  $R_i \ (i=1,\cdots,r)$  :

 $\text{IF } \rho_1(t) \text{ is } M_{i1} \text{ and } \cdots \text{ and } \rho_g(t) \text{ is } M_{ig},$ 

THEN

$$\begin{bmatrix} \dot{x}_{1}(t) \\ \varepsilon \dot{x}_{2}(t) \end{bmatrix} = A_{i} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + B_{i1}w(t) + B_{i2}u(t) + \sum_{k=1}^{m} F_{ik}p_{k}(t)$$

$$= \begin{bmatrix} A_{i11} A_{i12} \\ A_{i21} A_{i22} \end{bmatrix} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + \begin{bmatrix} B_{i11} \\ B_{i12} \end{bmatrix} w(t)$$

$$+ \begin{bmatrix} B_{i21} \\ B_{i22} \end{bmatrix} u(t) + \sum_{k=1}^{m} \begin{bmatrix} F_{ik1} \\ F_{ik2} \end{bmatrix} p_{k}(t),$$

$$z(t) = C_{i1} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + D_{i12}u(t)$$

$$= \begin{bmatrix} C_{i11} C_{i12} \end{bmatrix} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + D_{i12}u(t),$$

$$y(t) = C_{i2} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + D_{i21}w(t) + \sum_{k=1}^{m} G_{ik}p_{k}(t)$$

$$= \begin{bmatrix} C_{i21} C_{i22} \end{bmatrix} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + D_{i21}w(t) + \sum_{k=1}^{m} G_{ik}p_{k}(t),$$

$$q_{k}(t) = E_{ik} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + J_{ik}w(t) + H_{ik}p(t)$$

$$= \begin{bmatrix} E_{ik1} E_{ik2} \end{bmatrix} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + J_{ik}w(t) + H_{ik}p(t),$$

where r is the number of rules.  $\rho_i(t)$  and  $M_{ij}(1, \dots, g)$ are the premise variables and the fuzzy set respectively.  $x(t) \in \mathbb{R}^{n_1+n_2}$  is the state variable,  $w(t) \in \mathbb{R}^{n_w}$  is the external noise,  $z(t) \in \mathbb{R}^{n_z}$  is the controlled variable,  $u(t) \in \mathbb{R}^{n_u}$  is the input variable and  $y(t) \in \mathbb{R}^{n_y}$  is the output variable. All matrices  $A_i, B_i, \dots, H_{ik}$  are the constant matrices with compatible dimensions.  $p(t) = [p_1(t)^T \cdots p_m(t)^T]^T$ ,  $p_k(t) \in \mathbb{R}^{p_k}$ ,  $(k = 1, \dots, m)$ and  $q(t) = [q_1(t)^T \cdots q_m(t)^T]^T$ ,  $q_k(t) \in \mathbb{R}^{q_k}$   $(k = 1, \dots, m)$ are the uncertainty variables satisfying the following IQC'

$$\int_{0}^{\infty} p_{k}(t)^{T} p_{k}(t) dt \leq \int_{0}^{\infty} q_{k}(t)^{T} q_{k}(t) dt, \ k = 1, \cdots, m.$$
 (2)

The normalized membership function  $\mu_i(\rho(t))$  of the inferred fuzzy set  $h_i(\rho(t))$  is defined as follows:

$$\mu_i(\rho(t)) = \frac{h_i(\rho(t))}{\sum_{i=1}^r h_i(\rho(t))}, \ h_i(\rho(t)) = \prod_{j=1}^q M_{ij}(\rho_j(t)).$$
(3)

For all  $t \ge 0$ , we assume  $h_i(\rho(t)) \ge 0$   $(i=1,\cdots,r)$ and  $\sum_{i=1}^r h_i(\rho(t)) > 0$ . Then we get

$$\mu_i(\rho(t)) \ge 0, (i=1,\cdots,r), \quad \sum_{i=1}^r \mu_i(\rho(t)) = 1.$$

For simplicity, by defining

 $\mu_i = \mu_i(\rho(t)), \ \mu^T = [\mu_1 \cdots \mu_r], \ (1)$  can be written as follows :

$$\begin{split} \begin{bmatrix} \dot{x}_{1}(t) \\ \varepsilon x_{2}(t) \end{bmatrix} &= \begin{bmatrix} A_{11}(\mu) A_{12}(\mu) \\ A_{21}(\mu) A_{22}(\mu) \end{bmatrix} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + \begin{bmatrix} B_{11}(\mu) \\ B_{12}(\mu) \end{bmatrix} w(t) \\ &+ \begin{bmatrix} B_{21}(\mu) \\ B_{22}(\mu) \end{bmatrix} u(t) + \sum_{k=1}^{m} \begin{bmatrix} F_{k1}(\mu) \\ F_{k2}(\mu) \end{bmatrix} p_{k}(t) \\ &= \sum_{i=1}^{r} \mu_{i} \left( \begin{bmatrix} A_{i11} A_{i12} \\ A_{i21} A_{i22} \end{bmatrix} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + \begin{bmatrix} B_{i11} \\ B_{i12} \end{bmatrix} w(t) \\ &+ \begin{bmatrix} B_{i21} \\ B_{i22} \end{bmatrix} u(t) + \sum_{k=1}^{m} \begin{bmatrix} F_{ik1} \\ F_{ik2} \end{bmatrix} p_{k}(t) \right), \\ z(t) &= \begin{bmatrix} C_{11}(\mu) C_{12}(\mu) \end{bmatrix} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + D_{12}(\mu)u(t) \\ &= \sum_{i=1}^{r} \mu_{i} \left( \begin{bmatrix} C_{i11} C_{i12} \end{bmatrix} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + D_{i12}u(t) \right), \\ y(t) &= \begin{bmatrix} C_{21}(\mu) C_{22}(\mu) \end{bmatrix} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + D_{21}(\mu)w(t) \\ &+ \sum_{k=1}^{m} G_{k}(\mu)p_{k}(t) \\ &= \sum_{i=1}^{r} \mu_{i} \left( \begin{bmatrix} C_{i21} C_{i22} \end{bmatrix} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + D_{i21}w(t) \\ &+ \sum_{k=1}^{m} G_{ik}p_{k}(t) \right), \\ q_{k}(t) &= \begin{bmatrix} E_{k1}(\mu) E_{k2}(\mu) \end{bmatrix} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + J_{k}(\mu)w(t) \\ &+ H_{k}(\mu)p_{k}(t) \\ &= \sum_{i=1}^{r} \mu_{i} \left( \begin{bmatrix} E_{ik1} E_{ik2} \end{bmatrix} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \end{bmatrix} + J_{ik}w(t) + H_{ik}p_{k}(t) \right). \end{split}$$

In (4)  $x_2$  is about  $1/\varepsilon$  bigger comparing with  $x_1$  so that we call  $x_2$  the fast variable and  $x_1$  the slow variable. When the singular perturbation parameter  $\varepsilon$  is set to 0, the dimension of the full system (4) changes from  $n_1 + n_2$  to  $n_1$ . When  $\varepsilon = 0$ , assuming that  $A_{22}(\mu)^{-1}$  exists, the singularly perturbed fuzzy system (4) becomes as follows:

$$\dot{x}_{1}(t) = A_{11}(\mu)x_{1}(t) + A_{12}(\mu)x_{2}(t) + B_{11}(\mu)w(t) + B_{21}(\mu)u(t) + \sum_{k=1}^{m} F_{k1}(\mu)p_{k}(t),$$
(5)

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$$\begin{aligned} x_{2}(t) &= -A_{22}(\mu)^{-1}(A_{21}(\mu)x_{1}(t) + B_{12}(\mu)w(t) \\ &+ B_{22}(\mu)u(t) + \sum_{k=1}^{m} F_{k2}(\mu)p_{k}(t)). \end{aligned} \tag{6}$$

Substituting (6) into (5), we obtain

$$\begin{split} \dot{x}_{1}(t) &= A_{s}(\mu)x_{1}(t) + B_{s1}(\mu)w(t) + B_{s2}(\mu)u(t) \\ &+ \sum_{k=1}^{m} F_{sk}(\mu)p_{k}(t), \\ z(t) &= C_{s1}(\mu)x_{1}(t) + D_{s12}(\mu)u(t), \end{split} \tag{7}$$

$$y(t) &= C_{s2}(\mu)x_{1}(t) + D_{s21}(\mu)w(t) + \sum_{k=1}^{m} G_{sk}(\mu)p_{k}(t), \\ q_{k}(t) &= E_{sk}(\mu)x_{1}(t) + J_{sk}(\mu)w(t) + H_{sk}(\mu)p_{k}(t), \end{split}$$

where

$$\begin{split} &A_s(\mu) = A_{11}(\mu) - A_{12}(\mu)A_{22}(\mu)^{-1}A_{21}(\mu),\\ &B_{si}(\mu) = B_{i1}(\mu) - A_{12}(\mu)A_{22}(\mu)^{-1}B_{i2}(\mu) \quad i=1,2,\\ &F_{sk}(\mu) = F_{k1}(\mu) - A_{12}(\mu)A_{22}(\mu)^{-1}F_{k2}(\mu),\\ &C_{si}(\mu) = C_{i1}(\mu) - C_{i2}(\mu)A_{22}(\mu)^{-1}A_{21}(\mu) \quad i=1,2,\\ &D_{sij}(\mu) = D_{ij}(\mu) - C_{i2}(\mu)A_{22}(\mu)^{-1}B_{j2}(\mu) \quad i,j=1,2,\\ &G_{sk}(\mu) = G_k(\mu) - C_{22}(\mu)A_{22}(\mu)^{-1}F_{k2}(\mu),\\ &E_{sk}(\mu) = E_{k1}(\mu) - E_{k2}(\mu)A_{22}(\mu)A_{21}(\mu),\\ &J_{sk}(\mu) = J_k(\mu) - E_{k2}(\mu)A_{22}(\mu)^{-1}B_{12}(\mu),\\ &H_{sk}(\mu) = H_k(\mu) - E_{k2}(\mu)A_{22}(\mu)^{-1}F_{k2}(\mu). \end{split}$$

We call (7) the slow subsystem. When the slow variable  $x_1(t)$  is set to zero in (4), we obtain the following system :

$$\dot{x}_{2}(t) = A_{22}(\mu)x_{2}(t) + B_{12}(\mu)w(t) + B_{22}(\mu)u(t) + \sum_{k=1}^{m} F_{k2}(\mu)p_{k}(t),$$

$$z(t) = C_{12}(\mu)x_{2}(t) + D_{12}(\mu)u(t), \qquad (8)$$

$$y(t) = C_{22}(\mu)x_{2}(t) + D_{21}(\mu)w(t) + \sum_{k=1}^{m} G_{k}(\mu)p_{k}(t),$$

$$q_{k}(t) = E_{k2}(\mu)x_{2}(t) + J_{k}(\mu)w(t) + H_{k}(\mu)p_{k}(t).$$

We call (8) the fast subsystem.

In this work, we will suggest an  $\varepsilon$  independent  $H_{\infty}$  output feedback controller design method such that the closed loop system is stable and the  $H_{\infty}$  norm from the external noise to the controlled variable is less than a prescribed disturbance attenuation level  $\gamma$ , thus

$$\int_{0}^{\infty} z(\tau)^{T} z(\tau) d\tau < \gamma^{2} \int_{0}^{\infty} w(\tau)^{T} w(\tau) d\tau.$$

# III. Bounded $H_{\infty}$ norm conditions for uncertain singularly perturbed

#### T-S fuzzy system

In this section we introduce an  $H_{\infty}$  norm bounded condition for uncertain singularly perturbed T-S fuzzy

system. We consider an output feedback controller as in (9).

 $R_i$ : IF  $\rho_1(t)$  is  $M_{i1}$  and  $\cdots$  and  $\rho_g(t)$  is  $M_{ig}$ , THEN

$$E_{\varepsilon} \dot{x}_{k}(t) = \sum_{j=1}^{r} \mu_{j} A_{k,ij} x_{k}(t) + B_{k,i} y(t), \qquad (9)$$
$$u(t) = C_{k,i} x_{k}(t),$$

where  $E_{\varepsilon} = diag(I_{n_1}, \varepsilon I_{n_2})$ . Using the same notation as in (4), the controller can be represented as follows:

$$E_{k}x_{k}(t) = A_{k}(\mu)x_{k}(t) + B_{k}(\mu)y(t)$$

$$= \sum_{i=1}^{r}\sum_{j=1}^{r}\mu_{i}\mu_{j}(A_{k,ij}x_{k}(t) + B_{k,i}y(t)),$$

$$u(t) = C_{k}(\mu)x_{k}(t) = \sum_{i=1}^{r}\mu_{i}C_{k,i}x_{k}(t).$$
(10)

**Remark 1** : When  $B_{1,2} = \cdots = B_{r,2}$ , and  $C_{1,2} = \cdots = C_{r,2}$ , the controller state equation in (9) can be simplified as follows:  $E_{\varepsilon} \dot{x}_k(t) = A_{k,i} x_k(t) + B_{k,i} y(t)$ .

Applying the controller (10) to the system (4), the closed loop system becomes

$$\begin{split} E_{\varepsilon\varepsilon}\dot{x}_{c}(t) &= A_{c}(\mu)x_{c}(t) + B_{c}(\mu)w(t) + F_{c}(\mu)p(t), \\ z(t) &= \begin{bmatrix} C_{1}(\mu) & D_{12}(\mu)C_{k}(\mu) \end{bmatrix}x_{c}(t) = C_{c}(\mu)x_{c}(t), \ (11) \\ q(t) &= E_{c}(\mu)x_{c}(t) + J_{c}(\mu)w(t) + H_{c}(\mu)p(t), \end{split}$$

where 
$$E_{\varepsilon\varepsilon} = \begin{bmatrix} E_{\varepsilon} & 0 \\ 0 & E_{\varepsilon} \end{bmatrix}$$
,  $x_{c}(t) = \begin{bmatrix} x(t) \\ x_{k}(t) \end{bmatrix}$  and  
 $A_{c}(\mu) = \begin{bmatrix} A(\mu) & B_{2}(\mu)C_{k}(\mu) \\ B_{k}(\mu)C_{2}(\mu) & A_{k}(\mu) \end{bmatrix}$ ,  
 $B_{c}(\mu) = \begin{bmatrix} B_{1}(\mu) \\ B_{k}(\mu)D_{21}(\mu) \end{bmatrix}$ ,  $C_{c}(\mu) = \begin{bmatrix} C_{1}(\mu) & D_{12}(\mu)C_{k}(\mu) \end{bmatrix}$ ,  
 $F_{c}(\mu) = \begin{bmatrix} F(\mu) \\ B_{k}(\mu)G(\mu) \end{bmatrix}$ ,  $E_{c}(\mu) = \begin{bmatrix} E(\mu) & 0 \end{bmatrix}$ ,  
 $J_{c}(\mu) = J(\mu)$ ,  $H_{c}(\mu) = H(\mu)$ .

Using the coordinate transformation matrix M, the closed loop system (11) can be transformed into (12).

$$\begin{split} E_{2c} \begin{bmatrix} \dot{\zeta}_{1}(t) \\ \dot{\zeta}_{2}(t) \end{bmatrix} &= \overline{A}_{c}(\mu) \begin{bmatrix} \zeta_{1}(t) \\ \zeta_{2}(t) \end{bmatrix} + \overline{B}_{c}(\mu) w(t) + \overline{F}_{c}(\mu) p(t) \\ &= \begin{bmatrix} \overline{A_{c11}}(\mu) & \overline{A_{c12}}(\mu) \\ \overline{A_{c21}}(\mu) & \overline{A_{c22}}(\mu) \end{bmatrix} \begin{bmatrix} \zeta_{1}(t) \\ \zeta_{2}(t) \end{bmatrix} + \begin{bmatrix} \overline{B}_{c1}(\mu) \\ \overline{B}_{c2}(\mu) \end{bmatrix} w(t) \\ &+ \begin{bmatrix} \overline{F_{c1}}(\mu) \\ \overline{F_{c2}}(\mu) \end{bmatrix} p(t), \\ z(t) &= \overline{C}_{c}(\mu) \zeta(t) = \begin{bmatrix} \overline{C}_{c1}(\mu) & \overline{C}_{c2}(\mu) \end{bmatrix} \begin{bmatrix} \zeta_{1}(t) \\ \zeta_{2}(t) \end{bmatrix}, \quad (12) \\ q(t) &= \overline{E}_{c}(\mu) \zeta(t) + \overline{J}_{c}(\mu) w(t) + \overline{H}_{c}(\mu) p(t) \\ &= \begin{bmatrix} \overline{E}_{c1}(\mu) & \overline{E}_{c2}(\mu) \end{bmatrix} \begin{bmatrix} \zeta_{1}(t) \\ \zeta_{2}(t) \end{bmatrix} + \overline{J}_{c}(\mu) w(t) \\ &+ \overline{H}_{c}(\mu) p(t), \end{split}$$

where  $\zeta(t) = Mx_c(t)$  and

$$M = \begin{bmatrix} I_{n_1} & 0 & 0 & 0 \\ 0 & 0 & I_{n_1} & 0 \\ 0 & I_{n_2} & 0 & 0 \\ 0 & 0 & 0 & I_{n_2} \end{bmatrix} \qquad E_{2c} = \begin{bmatrix} I_{2n_1} & 0 \\ 0 & cI_{2n_2} \end{bmatrix},$$
  
$$\overline{A}_c(\mu) = MA_c(\mu)M^{-1}, \ \overline{B}_c(\mu) = MB_c(\mu),$$
  
$$\overline{C}_c(\mu) = C_c(\mu)M^{-1}, \ \overline{E}_c(\mu) = E_c(\mu)M^{-1},$$
  
$$\overline{F}_c(\mu) = MF_c(\mu), \ \overline{J}_c(\mu) = J_c(\mu), \ \overline{H}_c(\mu) = H_c(\mu).$$

The slow subsystem of the system (12) can be expressed as follows:

$$\begin{split} \zeta_1(t) &= A_s(\mu)\zeta_1(t) + B_s(\mu)w(t) + F_s(\mu)p(t), \\ z(t) &= C_s(\mu)\zeta_1(t), \\ q(t) &= E_s(\mu)\zeta_1(t) + J_s(\mu)w(t) + H_s(\mu)p(t), \end{split} \tag{13}$$

where

$$\begin{split} A_{s}(\mu) &= A_{c11}(\mu) - A_{c12}(\mu) A_{c21}^{-1}(\mu) A_{c21}(\mu), \\ B_{s}(\mu) &= \overline{B}_{c1}(\mu) - \overline{A}_{c12}(\mu) \overline{A}_{c22}^{-1}(\mu) \overline{B}_{c2}(\mu), \\ C_{s}(\mu) &= \overline{C}_{c1}(\mu) - \overline{C}_{c2}(\mu) \overline{A}_{c22}^{-1}(\mu) \overline{A}_{c21}(\mu), \\ F_{s}(\mu) &= \overline{F}_{c1}(\mu) - \overline{A}_{c12}(\mu) \overline{A}_{c22}^{-1}(\mu) \overline{F}_{c2}(\mu), \\ E_{s}(\mu) &= \overline{E}_{c1}(\mu) - \overline{E}_{c2}(\mu) \overline{A}_{c22}^{-1}(\mu) \overline{A}_{c21}(\mu), \\ J_{s}(\mu) &= \overline{J}_{c}(\mu) - \overline{E}_{c2}(\mu) \overline{A}_{c22}^{-1}(\mu) \overline{B}_{c2}(\mu), \\ H_{s}(\mu) &= \overline{H}_{c}(\mu) - \overline{E}_{c2}(\mu) \overline{A}_{c22}^{-1}(\mu) \overline{F}_{c2}(\mu). \end{split}$$

The fast subsystem of system (12) can be described as follows:

$$\begin{split} &\zeta_{2}(t) = \overline{A}_{c22}(\mu)\zeta_{2}(t) + \overline{B}_{c2}(\mu)w(t) + \overline{F}_{c2}(\mu)p(t), \\ &z(t) = \overline{C}_{c2}(\mu)\zeta_{2}(t), \\ &q(t) = \overline{E}_{c2}(\mu)\zeta_{2}(t) + \overline{J}_{c}(\mu)w(t) + \overline{H}_{c}(\mu)p(t). \end{split}$$

Let  $T_{zw}^s$  be the operator from w(t) to z(t) in the slow subsystem (14). If there exist  $P_s = P_s^T > 0$ , and  $S = S^T > 0$ ,  $R = R^T > 0$  satisfying LMI (15) for all  $\mu$ , then  $||T_{zw}^s||_{\infty} < \gamma$  for all uncertainties satisfying the IQC's.

$$\begin{bmatrix} A_{s}(\mu)^{T}P_{s} + P_{s}A_{s}(\mu) & * & * & * & * \\ B_{s}(\mu)^{T}P_{s} & -\gamma I & * & * & * \\ F_{s}(\mu)^{T}P_{s} & 0 & -S & * & * \\ C_{s}(\mu) & 0 & 0 & -\gamma I & * \\ E_{s}(\mu) & J_{s}(\mu) & H_{s}(\mu) & 0 & -R \end{bmatrix} < 0, \quad (15)$$

$$SR = I.$$

Let  $T_{zw}^{f}$  be the operator from w(t) to z(t) in the fast subsystem (14). If there exist  $P_{f} = P_{f}^{T} > 0$ , and  $S = S^{T} > 0$ ,  $R = R^{T} > 0$  satisfying LMI (16) for all  $\mu$ , then  $\|T_{zw}^{f}\|_{\infty} < \gamma$  for all uncertainties satisfying the IQC's.

$$\begin{bmatrix} \overline{A}_{c22}(\mu)^T P_f + P_f \overline{A}_{c22}(\mu) & * & * & * & * \\ \overline{B}_{c2}(\mu)^T P_f & -\gamma I & * & * & * \\ \overline{F}_{c2}(\mu)^T P_f & 0 & -S & * & * \\ \overline{C}_{c2}(\mu) & 0 & 0 & -\gamma I & * \\ \overline{C}_{c2}(\mu) & \overline{J}_c(\mu) \overline{H}_c(\mu) & 0 & -R \end{bmatrix} < 0, (16)$$

$$SR = I.$$

**Lemma 1**[7]: When the symmetric matrix  $\Omega$  and the matrix U, V is given, condition (i) and (ii) are equivalent.

(i) There exists K satisfying the following LMI.  $\Omega + UKV + (UKV)^T < 0.$ 

(ii)  $U^{\perp T} \Omega U^{\perp} < 0, V^{T \perp T} \Omega V^{T \perp} < 0.$ 

Lemma 2: The next two conditions are equivalent.

- (i) There exist symmetric positive definite matrices P<sub>s</sub>, P<sub>f</sub>, S and R satisfying LMI's (15) and (16) for all μ(t).
- (ii) There exist a matrix *P* and symmetric positive definite matrices S and R satisfying LMI (17) for all μ(t).

$$\begin{bmatrix} \overline{A}_{c}(\mu)^{T} \overline{P} + \overline{P}^{T} \overline{A}_{c}(\mu) & * & * & * & * \\ \overline{B}_{c}(\mu)^{T} \overline{P} & -\gamma I & * & * & * \\ \overline{F}_{c}(\mu)^{T} \overline{P} & 0 & -S & * & * \\ \overline{C}_{c}(\mu) & 0 & 0 & -\gamma I & * \\ \overline{C}_{c}(\mu) & \overline{J}_{c}(\mu) & \overline{H}_{c}(\mu) & 0 & -R \end{bmatrix} < 0. (17)$$

where  $\begin{bmatrix} \overline{P}_1 & 0 \\ \overline{P}_2 & \overline{P}_3 \end{bmatrix}$ ,  $\overline{P}_1 = \overline{P}_1^T > 0$ ,  $\overline{P}_3 = \overline{P}_3^T > 0$ .

(proof) LMI (17) can be expressed as follows:

$$\Omega + U\overline{P}_2 V + (U\overline{P}_2 V)^T < 0, \tag{18}$$

where

$$\Omega = \begin{bmatrix} \Omega_{11} & * & * & * & * & * & * \\ \Omega_{21} & \Omega_{22} & * & * & * & * \\ \overline{B}_{c1}(\mu)^{T} \overline{P}_{1}(\mu) & \overline{B}_{c2}(\mu)^{T} \overline{P}_{3} & -\gamma I & * & * & * \\ \overline{F}_{c1}(\mu)^{T} \overline{P}_{1}(\mu) & \overline{F}_{c2}(\mu)^{T} \overline{P}_{3} & 0 & -S & * & * \\ \overline{C}_{c1}(\mu) & \overline{C}_{c2}(\mu) & 0 & 0 & -\gamma I & * \\ \overline{E}_{c1}(\mu) & \overline{E}_{c2}(\mu) & \overline{J}_{c}(\mu) & \overline{H}_{c}(\mu) & 0 & -R \end{bmatrix} < 0$$

$$U^{T} = \begin{bmatrix} \overline{A}_{c21}(\mu) & \overline{A}_{c22}(\mu) & \overline{B}_{c21}(\mu) & \overline{F}_{c21}(\mu) & 0 & 0 \end{bmatrix},$$

$$\begin{split} & \mathcal{O} = [I^{1}\mathcal{O}_{21}(\mu), I^{2}\mathcal{O}_{22}(\mu), D^{2}\mathcal{O}_{21}(\mu), I^{2}\mathcal{O}_{21}(\mu), 0] \\ & V = [I^{0}0000], \\ & \Omega_{11} = \overline{A}_{c11}(\mu)^{T}\overline{P}_{1} + \overline{P}_{1}\overline{A}_{c11}(\mu), \\ & \Omega_{21} = \overline{A}_{c12}(\mu)^{T}\overline{P}_{1} + \overline{P}_{3}\overline{A}_{c21}(\mu), \\ & \Omega_{22} = \overline{A}_{c22}(\mu)^{T}\overline{P}_{3} + \overline{P}_{3}\overline{A}_{c22}(\mu). \end{split}$$

We can write  $U^{\perp}$  and  $V^{T_{\perp}}$  as follows:

$$U^{\perp} = \begin{bmatrix} I & 0 & 0 & 0 & 0 \\ u_{21} & u_{22} & u_{23} & 0 & 0 \\ 0 & I & 0 & 0 & 0 \\ 0 & 0 & I & 0 & 0 \\ 0 & 0 & 0 & I & 0 \\ 0 & 0 & 0 & 0 & I \end{bmatrix}, \qquad V^{T\perp} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ I & 0 & 0 & 0 \\ 0 & I & 0 & 0 \\ 0 & 0 & I & 0 \\ 0 & 0 & 0 & I & 0 \\ 0 & 0 & 0 & I & 0 \end{bmatrix},$$
(19)

where

$$\begin{split} u_{21} =& -\overline{A}_{\,c21}(\mu)^{\,T} (\overline{A}_{\,c22}(\mu)^{-1})^{\,T}, \\ u_{22} =& -\overline{B}_{\,c2}(\mu)^{\,T} (\overline{A}_{\,c22}(\mu)^{-1})^{\,T}, \\ u_{23} =& -\overline{F}_{\,c2}(\mu)^{\,T} (\overline{A}_{\,c22}(\mu)^{-1})^{\,T} \end{split}$$

Using  $U^{\perp}$  and  $V^{T_{\perp}}$  defined in (19), we obtain

$$\begin{array}{c} U^{\perp T} \Omega U^{\perp} = \\ \begin{bmatrix} A_{s}(\mu)^{T} \overline{P}_{1} + \overline{P}_{1} A_{s}(\mu) & * & * & * & * \\ B_{s}(\mu)^{T} \overline{P}_{1} & -\gamma I & * & * & * \\ F_{s}(\mu)^{T} \overline{P}_{1} & 0 & -S & * & * \\ C_{s}(\mu) & 0 & 0 & -\gamma I & * \\ E_{s}(\mu) & J_{s}(\mu) H_{s}(\mu) & 0 & -R \end{bmatrix} < 0 \quad (20) \\ V^{T \perp T} \Omega V^{T \perp} = \\ \begin{bmatrix} \overline{A}_{c22}(\mu)^{T} \overline{P}_{3} + \overline{P}_{3} \overline{A}_{c22}(\mu) & * & * & * & * \\ \overline{B}_{c2}(\mu)^{T} \overline{P}_{3} & -\gamma I & * & * & * \\ \overline{B}_{c2}(\mu)^{T} \overline{P}_{3} & 0 & -S & * & * \\ \overline{C}_{c2}(\mu) & 0 & 0 & -\gamma I & * \\ \overline{E}_{c2}(\mu) & \overline{J}_{c}(\mu) \overline{H}_{c}(\mu) & 0 & -R \end{bmatrix} < 0 \quad (21)$$

From lemma 1, we complete the proof.

**Lemma 3:** If LMI (18) has a feasible solution, there exists an  $\overline{e}$  such that  $||T_{zw}||_{\infty} < \gamma$  for all  $0 < \varepsilon < \overline{e}$ , where  $T_{zw}$  is the operator from w(t) to z(t) in the full order singularly perturbed system (12).

(Proof) In (12) we define

$$\begin{split} \overline{A}_{c}(\mu,\varepsilon) &= \begin{bmatrix} \overline{A}_{c11}(\mu) & \overline{A}_{c12}(\mu) \\ \overline{A}_{c21}(\mu)/\varepsilon & \overline{A}_{c22}(\mu)/\varepsilon \end{bmatrix}, \ \overline{B}_{c}(\mu,\varepsilon) &= \begin{bmatrix} \overline{B}_{c1}(\mu) \\ \overline{B}_{c2}(\mu)/\varepsilon \end{bmatrix}, \\ \overline{F}_{c}(\mu,\varepsilon) &= \begin{bmatrix} \overline{F}_{c1}(\mu) \\ \overline{F}_{c2}(\mu)/\varepsilon \end{bmatrix}. \end{split}$$

The solvability of LMI (22) guarantees  $\|T_{zw}\|_{\infty} < \gamma$  in (12).

$$\begin{bmatrix} \overline{A}_{c}(\mu,\varepsilon)^{T}P_{\varepsilon} + P_{\varepsilon}\overline{A}_{c}(\mu,\varepsilon) & * & * & * & * \\ \overline{B}_{c}(\mu)^{T}P_{\varepsilon} & -\gamma I & * & * & * \\ \overline{F}_{c}(\mu)^{T}P_{\varepsilon} & 0 & -S_{\varepsilon} & * & * \\ \overline{C}_{c}(\mu) & 0 & 0 & -\gamma I & * \\ \overline{C}_{c}(\mu) & \overline{J}_{c}(\mu) \overline{H}_{c}(\mu) & 0 & -R_{\varepsilon} \end{bmatrix} < 0. (22)$$

We seek  $P_{\varepsilon}$ ,  $S_{\varepsilon}$  and  $R_{\varepsilon}$  from solutions of LMI (18). Let's  $\overline{P} = \begin{bmatrix} \overline{P}_1 & 0 \\ \overline{P}_2 & \overline{P}_3 \end{bmatrix}$ , S and R be any feasible solutions of LMI (18). We define  $P_{\varepsilon} = \begin{bmatrix} \overline{P}_1 & \varepsilon \overline{P}_2 \\ \varepsilon \overline{P}_2 & \varepsilon \overline{P}_3 \end{bmatrix}$ ,  $S_{\varepsilon} = S$  and  $R_{\varepsilon} = R$ . Since  $\overline{P}_1$  and  $\overline{P}_3$  are positive definite matrices there exists  $\varepsilon_1$  such that  $P_{\varepsilon}$  is positive definite for all  $0 < \varepsilon < \varepsilon_1$ . Substituting  $P_{\varepsilon}$ ,  $S_{\varepsilon}$  and  $R_{\varepsilon}$  into (22) leads to

$$\begin{bmatrix} \overline{A}_{c}(\mu,\varepsilon)^{T}P_{\varepsilon} + P_{\varepsilon}\overline{A}_{c}(\mu,\varepsilon) & * & * & * & * \\ & \overline{B}_{c}(\mu)^{T}P_{\varepsilon} & -\gamma I & * & * & * \\ & \overline{F}_{c}(\mu)^{T}P_{\varepsilon} & 0 & -S_{\varepsilon} & * & * \\ & \overline{C}_{c}(\mu) & 0 & 0 & -\gamma I & * \\ & \overline{E}_{c}(\mu) & \overline{J}_{c}(\mu)\overline{H}_{c}(\mu) & 0 & -R_{\varepsilon} \end{bmatrix} = \begin{bmatrix} \overline{A}_{c}(\mu)^{T}\overline{P} + \overline{P}^{T}\overline{A}_{c}(\mu) & * & * & * & * \\ & \overline{B}_{c}(\mu)^{T}\overline{P} & -\gamma I & * & * & * \\ & \overline{B}_{c}(\mu)^{T}\overline{P} & 0 & -S & * & * \\ & \overline{C}_{c}(\mu) & 0 & 0 & -\gamma I & * \\ & \overline{C}_{c}(\mu) & 0 & 0 & -\gamma I & * \\ & \overline{E}_{c}(\mu) & \overline{J}_{c}(\mu)\overline{H}_{c}(\mu) & 0 & -R \end{bmatrix} + \\ \varepsilon \begin{bmatrix} \overline{P}_{2}\overline{A}_{c11}(\mu) & \overline{A}_{c12}(\mu)^{T}\overline{P}_{2}^{T} + \overline{P}_{2}\overline{A}_{c12}(\mu) & * & * & * \\ & 0 & \overline{B}_{c1}(\mu)^{T}\overline{P}_{2}^{T} & 0 & * & * \\ & 0 & \overline{B}_{c1}(\mu)^{T}\overline{P}_{2}^{T} & 0 & 0 & * & * \\ & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} . (23)$$

Since the first term of the right hand side of (23) is negative definite, there exists an  $\varepsilon_2$  such that  $P_{\varepsilon}$ ,  $S_{\varepsilon}$  and  $R_{\varepsilon}$  satisfy LMI (22) for all  $0 < \varepsilon < \varepsilon_2$ . By choosing  $\overline{\varepsilon} = \min(\varepsilon_1, \varepsilon_2)$  the proof is completed.

#### IV. Design of Output Feedback Controller

In this section, we suggest an output feedback controller design method based on lemma 3. By pre and post-multiplying LMI (17) by  $diag(M^T, I, I)$  and diag(M, I, I), we obtain

$$\begin{bmatrix} A_{c}(\mu)^{T}P + PA_{c}(\mu) & * & * & * & * \\ B_{c}(\mu)^{T}P & -\gamma I & * & * & * \\ F_{c}(\mu)^{T}P & 0 & -S & * & * \\ C_{c}(\mu) & 0 & 0 & -\gamma I & * \\ E_{c}(\mu) & J_{c}(\mu) & H_{c}(\mu) & 0 & -R \end{bmatrix} < 0, \quad (24)$$

where

$$P = M^T \overline{P} M = \begin{bmatrix} P_1 & P_2 \\ P_3 & P_4 \end{bmatrix}.$$
(25)

From the relationship of  $\overline{P}$  and P defined in (17) and (25), the sub-matrices  $P_i(i=1,\dots,4)$  must have the structure as in (26) and (27).

$$P_{i} = \begin{bmatrix} P_{i1} & 0 \\ P_{i2} & P_{i3} \end{bmatrix}, \quad (i = 1, \cdots, 4),$$
(26)

$$\begin{bmatrix} P_{11} P_{21} \\ P_{31} P_{41} \end{bmatrix} = \begin{bmatrix} P_{11} P_{21} \\ P_{31} P_{41} \end{bmatrix}^T > 0, \ \begin{bmatrix} P_{13} P_{23} \\ P_{33} P_{43} \end{bmatrix} = \begin{bmatrix} P_{13} P_{23} \\ P_{33} P_{43} \end{bmatrix}^T > 0. \ (27)$$

We introduce a technical lemma necessary for controller design.

**Lemma 4:** Suppose that  $P_1$ ,  $Q_1$  have the same matrix structure defined in (26) and (27) and  $I - P_{11}Q_{11} < 0$ ,  $I - P_{13}Q_{13} < 0$ . Then there exist P and Q satisfying the following two conditions.

 (i) All the sub-matrices have the same structure as in (25)-(27).

(ii) PQ = I.

(proof) We choose an arbitrary invertible matrices  $P_2$  and  $P_3$  with the following structures.

$$P_{2} = \begin{bmatrix} P_{21} & 0 \\ P_{22} & P_{23} \end{bmatrix}, P_{3} = \begin{bmatrix} P_{31} & 0 \\ P_{32} & P_{33} \end{bmatrix}, P_{21} = P_{31}^{T}, P_{23} = P_{33}^{T}.$$
 (28)

We define  $P_4$ ,  $Q_2$ ,  $Q_3$  and  $Q_4$  as follows:

$$Q_{2} = (I - Q_{1}P_{1})P_{3}^{-1}, \ Q_{3} = P_{2}^{-1}(I - P_{1}Q_{1}),$$

$$P_{4} = -P_{3}Q_{1}(I - P_{1}Q_{1})^{-1}P_{2}, \ Q_{4} = (P_{4} - P_{3}P_{1}^{-1}P_{2})^{-1}.$$

$$[P, P] \qquad [Q, Q]$$

$$[P, P] \qquad [Q, Q]$$

Let *P* and *Q* be  $P = \begin{bmatrix} P_1 & P_2 \\ P_3 & P_4 \end{bmatrix}$ ,  $Q = \begin{bmatrix} Q_1 & Q_2 \\ Q_3 & Q_4 \end{bmatrix}$ . One can easily know PQ = I. This completes the proof.

From (11) we have

$$\begin{split} A_{c}(\mu) &= \sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} A_{cij} = \sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} \begin{bmatrix} A_{i} & B_{i2}C_{kj} \\ B_{kj}C_{i2} & A_{kij} \end{bmatrix} \\ B_{c}(\mu) &= \sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} B_{cij} = \sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} \begin{bmatrix} B_{i1} \\ B_{kj}D_{i21} \end{bmatrix}, \\ C_{c}(\mu) &= \sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} C_{cij} = \sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} \begin{bmatrix} C_{i1} & D_{i12}C_{kj} \end{bmatrix}, \\ F_{c}(\mu) &= \sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} F_{cij} = \sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} \begin{bmatrix} F_{j} \\ B_{kj}G_{i} \end{bmatrix}, \\ E_{c}(\mu) &= \sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} E_{cij} = \sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} [E_{i} & 0], \\ J_{c}(\mu) &= \sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} J_{cij} = \sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} J_{i}, \\ H_{c}(\mu) &= \sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} H_{cij} = \sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} H_{i}. \\ \text{LMI (24) can be written as follows:} \end{split}$$

$$\sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} \Phi_{ij} = \sum_{i=1}^{r} \sum_{j=1}^{r} \mu_{i} \mu_{j} \begin{bmatrix} A_{cij}^{T} P + P A_{cij} & * & * & * \\ B_{cij}^{T} P & -\gamma I & * & * \\ F_{cij}^{T} P & 0 & -S & * & * \\ C_{cij} & 0 & 0 & -\gamma I & * \\ E_{cij} & J_{cij} & H_{cij} & 0 & -R \end{bmatrix} < 0.$$

$$(30)$$

It is well known that the solvability of LMI (31) guarantees the solvability of LMI (30) for all permis-

$$\begin{array}{l} \mbox{sible } \mu(t). \\ \Phi_{ii} < 0 \; (i=1, \cdots, r) \qquad \Phi_{ij} + \Phi_{ji} < 0, \; (i=1, \cdots, r, \; j > i). \; (31) \end{array}$$

**Theorem 5:** Suppose that  $P_1, Q_1$  with matrix structure defined in lemma 4 and  $S = S^T = R^{-1} > 0$ ,  $\widehat{A}_{ij}, \widehat{B}_j, \widehat{C}_j (i, j = 1, \dots, r)$  solve LMI's (32). There exists a sufficiently small  $\overline{\varepsilon}$  such that the design objective is achieved in an uncertain singularly perturbed system (1) for all  $0 < \varepsilon < \overline{\varepsilon}$ .

$$\Psi_{ii} < 0, \ \Psi_{ij} + \Psi_{ji} < 0, \ i, j = 1, \cdots, r, \ j >, \eqno(32)$$

where

$$\Psi_{ij} = \begin{bmatrix} A_1 & * & * & * & * & * \\ A_i^T + \hat{A}_{ij} & A_2 & * & * & * & * \\ B_{i1}^T & B_{i1}^T P_1 + D_{i21}^T \hat{B}_j^T - \gamma I & * & * & * \\ F_j^T & F_j^T P_1 + G_i^T \hat{B}_j^T & 0 & -S & * & * \\ C_{i1}Q_1 + D_{i12}\hat{C}_j & C_{i1} & 0 & 0 & -\gamma I & * \\ E_iQ_1 & E_i & J_i & H_i & 0 & -R \end{bmatrix}.$$

$$A_1 = A_iQ_1 + Q_1^T A_i^T + \hat{C}_j^T B_{i2}^T + B_{kj}\hat{C}_j,$$
(33)

$$A_2 = A_i^{\ T} P_1 + P_1^{\ T} A_i + C_{i2}^{\ T} \hat{B}_j^{\ T} + \hat{B}_j C_{i2}.$$

(proof) From  $P_1$  and  $Q_1$ , P and Q can be constructed from lemma 4. We define  $\Pi_1$  and  $\Pi_2$  as follows :

$$\Pi_1 = \begin{bmatrix} Q_1 & I \\ Q_3 & 0 \end{bmatrix}, \ \Pi_2 = \begin{bmatrix} I & P_1 \\ 0 & P_3 \end{bmatrix}.$$
(34)

Note that  $P\Pi_1 = \Pi_2$ . By direct matrix manipulation we obtain

$$\begin{split} \Psi_{ij} &= \begin{bmatrix} \Pi_1^{T****} \\ 0 & I^{***} \\ 0 & 0 & I^* \\ 0 & 0 & 0 & I^* \\ 0 & 0 & 0 & I \end{bmatrix} \Phi_{ij} \begin{bmatrix} \Pi_1 * * * * \\ 0 & I^{***} \\ 0 & I^{***} \\ 0 & 0 & I^* \\ 0 & 0 & 0 & I \end{bmatrix} \\ &= \begin{bmatrix} A_1 & * & * & * & * & * \\ A_i^T + \hat{A}_{ij} & A_2 & * & * & * & * \\ A_i^T + \hat{A}_{ij} & A_2 & * & * & * & * \\ B_{i1}^T & B_{i1}^T P_1 + D_{i21}^T \hat{B}_j^T - \gamma I & * & * & * \\ B_{i1}^T & F_j^T P_1 + G_i^T \hat{B}_j^T & 0 & -S & * & * \\ F_j^T & F_j^T P_1 + G_i^T \hat{B}_j^T & 0 & -S & * & * \\ C_{i1} Q_1 + D_{i12} \hat{C}_j & C_{i1} & 0 & 0 & -\gamma I & * \\ E_i Q_1 & E_i & J_i & H_i & 0 & -R \end{bmatrix}, \end{split}$$

where  $\hat{A}_{ij} = P_1^T A_i Q_1 + \hat{B}_j C_{i2} Q_1 + P_1^T B_{i2} \hat{C}_j + P_3^T A_{kij} Q_3,$  $\hat{B}_j = P_3^T B_{kj}, \quad \hat{C}_j = C_{kj} Q_3.$ 

Now the state space realization

 $(A_{kij}, B_{kj}, C_{kj}), (i, j = 1, \dots, r)$  of desired output feedback controller can be calculated from  $\hat{A}_{ij}, \hat{B}_j, \hat{C}_j$  in (35). This completes the proof.

**Lemma 6:** When  $B_{12} = \cdots = B_{r2}$ ,  $C_{12} = \cdots = C_{r2}$ , Suppose

that  $P_1, Q_1$  with matrix structure defined in lemma 4 and  $S = S^T = R^{-1} > 0, \ \hat{A}_i, \ \hat{B}_i, \ \hat{C}_i \ (i = 1, \cdots, r)$  solve LMI's (33). Then for a sufficiently small  $\varepsilon > 0$ , there exists an output feedback controller guaranteeing  $\|T_{zw}\|_{\infty} < \gamma$  in the full system (4).

$$\varPsi_i < 0, \ i = 1, \cdots, r \tag{33}$$

where

$$\Psi_{i} = \begin{bmatrix} \Xi_{1} & * & * & * & * & * & * \\ A_{i}^{T} + \hat{A}_{i} & \Xi_{2} & * & * & * & * \\ B_{i1}^{T} & B_{i1}^{T}P_{1} + D_{i21}^{T}\hat{B}_{i}^{T} - \gamma I & * & * & * \\ F_{i}^{T} & F_{i}^{T}P_{1} + G_{i}^{T}\hat{B}_{i}^{T} & 0 & -S & * & * \\ C_{i1}Q_{1} + D_{i12}\hat{C}_{i} & C_{i1} & 0 & 0 & -\gamma I & * \\ E_{i}Q_{1} & E_{i} & J_{i} & H_{i} & 0 & -R \end{bmatrix},$$
$$\Xi_{1} = A_{i}Q_{1} + Q_{1}^{T}A_{i}^{T} + \hat{C}_{i}^{T}B_{i2}^{T} + B_{ki}\hat{C}_{i}, \qquad (34)$$

$$\begin{split} \Xi_2 &= A_i^{\ T} P_1 + P_1^{\ T} A_i + C_{i2}^{\ T} \hat{B}_i^{\ T} + \hat{B}_i C_{i2}, \\ \hat{A}_i &= P_1^{\ T} A_i Q_1 + \hat{B}_i C_{i2} Q_1 + P_1^{\ T} B_{i2} \hat{C}_i + P_3^{\ T} A_{ki} Q_3 \\ \hat{B}_i &= P_3^{\ T} B_{ki}, \ \hat{C}_i &= C_{ki} Q_3. \end{split}$$

The state space realization  $(A_{ki}, B_{ki}, C_{ki})$ ,  $(i = 1, \dots, r)$  of the output feedback controller can be calculated from  $\hat{A}_i, \hat{B}_i, \hat{C}_i$ .

**Remark 2:** The main advantage of singular perturbation method in control problem is to avoid numerical ill-conditioning problems resulting from a small parasitic term  $\varepsilon$ . In this work, we present an  $\varepsilon$ -independent controller design method to avoid numerical ill-conditioning problems.

#### V. A numerical Example

In this section, we present a numerical example to illustrate the efficacy of the proposed design method. We consider an inverted pendulum controlled by a dc motor via a gear train which can be described by the following singularly perturbed T-S fuzzy system [15]-[16]:

$$R_1$$
: If  $x_1(t)$  is  $M_1$ ,

THEN

$$\begin{aligned} E_{\varepsilon} \dot{x}(t) &= A_{1} x(t) + B_{1,1} w(t) + B_{1,2} u(t) + F_{1} p(t), \\ z(t) &= C_{1,1} x(t) + D_{1,12} u(t), \\ y(t) &= C_{1,2} x(t) + D_{1,21} w(t) + G_{1} p(t), \\ q(t) &= E_{1} x(t). \end{aligned}$$
(35)

 $R_2$ : If  $x_1(t)$  is  $M_2$ ,

THEN

$$E_{\varepsilon}\dot{x}(t) = A_{2}x(t) + B_{2,1}w(t) + B_{2,2}u(t) + F_{2}p(t),$$
  

$$z(t) = C_{2,1}x(t) + D_{2,12}u(t),$$
  

$$y(t) = C_{2,2}x(t) + D_{2,21}w(t) + G_{2}p(t),$$
  

$$q(t) = E_{2}x(t),$$
  
(36)

where the membership function  $M_1 = 1 - |x_1(t)|$ ,  $M_2 = 1 - M_1$  and

$$\begin{split} E_{\varepsilon} &= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & \varepsilon \end{bmatrix}, \qquad A_{1} = \begin{bmatrix} 0 & 1 & 0 \\ 9.8 & 0 & 1 \\ 0 & -1 & -1 \end{bmatrix}, \qquad A_{2} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & -1 & -1 \end{bmatrix}, \\ B_{1,1} &= B_{2,1} = \begin{bmatrix} 0.1 \\ 0 \\ 0 \end{bmatrix}, B_{1,2} = B_{2,2} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \qquad C_{1,1} = C_{2,1} = \begin{bmatrix} 0.1 \\ 0 \\ 0 \end{bmatrix}^{T}, \\ C_{1,2} &= C_{2,2} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}^{T}, \quad F_{1} = F_{2} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \qquad G_{1} = G_{2} = 0, \end{split}$$

$$D_{1,12} = D_{2,12} = D_{1,21} = D_{2,21} = 0.1, \ E_1 = E_2 = \begin{bmatrix} 0 & 0 & 0.1 \end{bmatrix}$$

When  $\gamma = 0.5$  and R = 1, solving LMI's (33) gives

$$\begin{split} P_1 = \begin{bmatrix} 2.5172e+1 & -6.4124e+0 & 0 \\ -6.4124e+0 & 2.4170e+0 & 0 \\ 5.9783e+0 & 3.9826e+2 & 3.9804e+2 \end{bmatrix},\\ Q_1 = \begin{bmatrix} 2.0705e+0 & -5.7032e+0 & 0 \\ -5.7032e+0 & 1.9998e+1 & 0 \\ -1.3214e+0-1.0295e+2\,1.1155e+2 \end{bmatrix}, \end{split}$$

so that we obtain a controller as

$$\begin{split} A_{k,1} &= \begin{bmatrix} -3.1807e+3 & -1.0289e+4 & -3.7080e-2\\ -1.9710e+5 & -6.3860e+5 & -1.3896e+0\\ -1.9704e+5 & -6.3840e+5 & -1.3949e+0 \end{bmatrix}, \\ A_{k,2} &= \begin{bmatrix} -2.8389e+3 & -9.1858e+3 & -3.7066e-2\\ -1.9704e+5 & -6.3842e+5 & -1.3886e+0\\ -1.9681e+5 & -6.3765e+5 & -1.3939e+0 \end{bmatrix}, \\ B_{k,1} &= \begin{bmatrix} -6.7547e+1\\ 4.9099e+0\\ 3.5250e-1 \end{bmatrix}, \quad B_{k,2} &= \begin{bmatrix} -7.2651e+1\\ 5.8382e+0\\ 3.2536e-1 \end{bmatrix}, \\ C_{k,1} &= \begin{bmatrix} 4.9514e+2 & 1.6042e+3 & 9.9208e-4 \end{bmatrix}, \\ C_{k,2} &= \begin{bmatrix} 4.9455e+2 & 1.6023e+3 & 9.8962e-4 \end{bmatrix}. \end{split}$$

In order to see if the above controller meets the design specification, we check the solvability of LMI's (22) for the full order closed loop system. We can obtain a positive definite solution  $P_{\varepsilon}$  when  $\gamma = 0.5$ ,  $\varepsilon \leq 4.4 \times 10^{-5}$ .

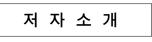
#### VI. Conclusion

In this work we have suggested a robust  $H_{\infty}$  output feedback controller design method for uncertain singularly perturbed T-S fuzzy systems. When the  $H_{\infty}$ norm of both slow and fast sub-systems of uncertain singularly perturbed system are less than  $\gamma$ , we proved that the  $H_{\infty}$  norm of the full system is also less than  $\gamma$ for a sufficiently small parasitic term  $\varepsilon$ . A sufficient condition guaranteing the existence of an output feedback controller satisfying the design specification has been presented in terms of  $\varepsilon$  independent linear matrix inequalities which may alleviate a numerical ill- conditioning problems resulting from the extremely small  $\varepsilon$ .

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