

Trajectory tracking fault tolerant controller design for Takagi-Sugeno systems subject to actuator faults

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Abstract—This paper investigates the problem of fault tolerant control (FTC) design for nonlinear Takagi-Sugeno (T-S) models with measurable premise variables. The idea is to synthesize a fault tolerant controller ensuring state trajectory tracking. Based on Lyapunov theory, new less conservative approaches are proposed in term of Linear Matrix Inequality (LMI). A PI observer is needed to estimate simultaneously the faults and the faulty system states in order to reconfigure the FTC law. A numerical example is considered to compare the conservatism of the proposed FTC approaches with the existing one and to illustrate the effectiveness of the FTC technique vs. the classical controller design methodology.

Keywords- Takagi-Sugeno nonlinear models, PI observer, state and fault estimation, LMI, Lyapunov theory, L_2 norm.

I. INTRODUCTION

The classical control law schemes have shown their interest in the system stabilization framework. Nevertheless, if faults affect the system, the classical controllers may not ensure the system stabilization. In this case, fault tolerant control is introduced to take into account the faults affecting the system components. In literature, two kinds of strategies dealing with the above problem have been proposed. The first one is called robust control or passive FTC. The main idea of this technique is to consider the faults as non structural bounded uncertainties which effect on the system will be minimized by using the L_2 norm. The passive control strategy is designed only for norm bounded faults which constitutes a major drawback of this technique. The second kind is called active FTC strategy. This latter requires the knowledge of the faults to reconfigure the controller to ensure the stability of the faulty system.

The FTC problem has already been studied in the literature. For instance, fault tolerant controller design methodology for linear systems is proposed by [1], [2], [3], [4] and [5]. Recently, this study has been extended to the nonlinear systems given in Takagi-Sugeno [6] representation by [8]. Nevertheless, the proposed approach may be conservative. Moreover, new approaches for trajectory tracking FTC design for T-S models with unmeasurable premise variables have been proposed by [7] and [9].

This paper aims to reduce the conservatism of the results proposed in [8] and to show the effectiveness of the FTC law compared to a classical one when faults affect the system dynamics. Thus, this paper is organized as follows. In the next section, the problem of fault tolerant controller design is presented. In section 3, an active FTC approach is proposed. In the last section, a numerical example is considered to illustrate the efficiency of the proposed active FTC approach compared to a passive one (developed in the appendix). Moreover, the

feasibility areas of the proposed active FTC approach and the one given in [8] are compared.

The following notations are considered to improve the paper readability. The single or double sums can be rewritten as:

$$\phi_\mu = \sum_{i=1}^r \mu_i(\xi(t)) \phi_i \text{ and } \phi_{\mu\mu} = \sum_{i=1}^r \sum_{j=1}^r \mu_i(\xi(t)) \mu_j(\xi(t)) \phi_{ij}.$$

The symbol $*$ denotes the transposed element in the symmetric positions of a matrix and $\text{diag}(M_1, \dots, M_r)$ is a block diagonal matrix which diagonal entries are defined by M_1, \dots, M_r . The following lemma will be needed.

Lemma 1 [10]: Consider two real matrices X and Y with appropriate dimensions, for any positive scalar δ the following inequality holds:

$$X^T Y + Y^T X \leq \delta X^T X + \delta^{-1} Y^T Y \quad (1)$$

II. PROBLEM FORMULATION

Let us consider the following T-S model without faults corresponding to the reference model.

$$\begin{cases} \dot{x}(t) = \sum_{i=1}^r \mu_i(\xi(t)) (A_i x(t) + B_i u(t)) \\ y(t) = \sum_{i=1}^r \mu_i(\xi(t)) (C_i x(t) + D_i u(t)) \end{cases} \quad (2)$$

where r is the number of submodels, $\xi(t)$ is the measurable premise variable, $\mu_i(\xi(t))$ are the membership functions verifying the convex sum property $0 \leq \mu_i(\xi(t)) \leq 1$ and

$\sum_{i=1}^r \mu_i(\xi(t)) = 1$, $x(t) \in \mathbb{R}^n$, $y(t) \in \mathbb{R}^p$ and $u(t) \in \mathbb{R}^m$ represent respectively the state, the output and the input vectors, $\{A_i, B_i, C_i, D_i\}$ are the submodels matrices.

Consider the faulty system given by

$$\begin{cases} \dot{x}_f(t) = \sum_{i=1}^r \mu_i(\xi(t)) (A_i x_f(t) + B_i (u_f(t) + f(t))) \\ y_f(t) = \sum_{i=1}^r \mu_i(\xi(t)) (C_i x_f(t) + D_i (u_f(t) + f(t))) \end{cases} \quad (3)$$

where $x_f(t) \in \mathbb{R}^n$, $y_f(t) \in \mathbb{R}^p$ and $u_f(t) \in \mathbb{R}^m$ represent respectively the faulty state and faulty output vectors and the fault tolerant control signal, $f(t) \in \mathbb{R}^m$ is the fault directly

affecting the input. The fault is supposed to be constant (i.e. $df(t)/dt = 0$).

The objective is to design a fault tolerant controller ensuring the convergence of the faulty state vector $x_f(t)$ to the nominal one $x(t)$. The methodology of controller conception is based on the scheme depicted in Fig.1.

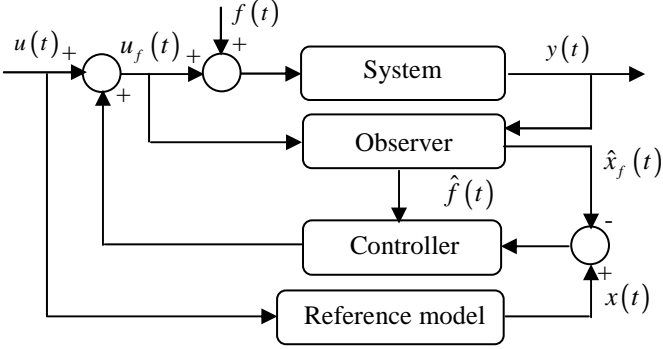


Fig.1. Fault tolerant control strategy

Let us consider the FTC law given by:

$$u_f(t) = \sum_{i=1}^r \mu_i(\xi(t)) K_i(x(t) - \hat{x}_f(t)) + u(t) - \hat{f}(t) \quad (4)$$

where: $K_i \in \mathbb{R}^{m \times n}$ are the state feedback gain matrices to be synthesized. The FTC design simultaneously requires the knowledge of the faulty state vector and the faults affecting the system. In order to estimate $x_f(t)$ and $f(t)$, the following PI observer is considered:

$$\begin{cases} \dot{\hat{x}}_f(t) = \sum_{i=1}^r \mu_i(\xi(t)) (A_i \hat{x}_f(t) + B_i (u_f(t) + \hat{f}(t)) + H_i^1 (y_f(t) - \hat{y}_f(t))) \\ \dot{\hat{f}}(t) = \sum_{i=1}^r \mu_i(\xi(t)) (H_i^2 (y_f(t) - \hat{y}_f(t))) \\ \hat{y}_f(t) = \sum_{i=1}^r \mu_i(\xi(t)) (C_i \hat{x}_f(t) + D_i (u_f(t) + \hat{f}(t))) \end{cases} \quad (5)$$

where $H_i^1 \in \mathbb{R}^{n \times p}$ and $H_i^2 \in \mathbb{R}^{m \times p}$ are the observer's gain matrices to be determined to estimate $f(t)$ and $x_f(t)$. A first solution to this problem was proposed in theorem 5.4 of [8].

III. FAULT TOLERANT CONTROLLER DESIGN

In this section we propose a less conservative approach for fault tolerant controller conception. Let us respectively define the state and fault estimation errors defined by: $e_s(t) = x_f(t) - \hat{x}_f(t)$ and $e_d(t) = f(t) - \hat{f}(t)$. Let us also define the state tracking error $e_p(t) = x(t) - x_f(t)$ and the output estimation error $e_y(t) = y_f(t) - \hat{y}_f(t)$. By adding and subtracting $K_\mu x_f(t)$ in (4), one can obtain:

$$u_f(t) = K_\mu (x(t) - x_f(t)) + K_\mu (x_f(t) - \hat{x}_f(t)) + u(t) - \hat{f}(t) \quad (6)$$

The dynamics of $e_p(t)$ and $e_s(t)$ are given by:

$$\dot{e}_p(t) = (A_\mu - B_\mu K_\mu) e_p(t) - B_\mu K_\mu e_s(t) - B_\mu e_d(t) \quad (7)$$

$$\dot{e}_s(t) = A_\mu e_s(t) + B_\mu e_d(t) - H_\mu^1 e_y(t) \quad (8)$$

According to (8), to avoid the crossing terms resulting from the observer's gains H_i^1 and system matrices (C_i and D_i) multiplication, we introduce a "virtual dynamics" in the output error $e_y(t)$ [11] [12]. This latter can be expressed as:

$$0 \dot{e}_y(t) = C_\mu e_s(t) + D_\mu e_d(t) - e_y(t) \quad (9)$$

where $0 \in \mathbb{R}^{p \times p}$ is a zero matrix.

Since the faults affecting the system are supposed to be constant (i.e. $\dot{f}(t) = 0$), the dynamics of the fault estimation error is given by:

$$\dot{e}_d(t) = -H_\mu^2 C_\mu e_s(t) - H_\mu^2 D_\mu e_d(t) \quad (10)$$

The combination of (7), (8), (9) and (10) allows the formulation of the dynamics of $e_y(t)$, $e_p(t)$, $e_s(t)$ and $e_d(t)$ in a descriptor form:

$$E \dot{\tilde{e}}(t) = \tilde{A}_\mu \tilde{e}(t) \quad (11)$$

where $E = \text{diag}(I \ I \ I \ 0_m)$, $\tilde{e}^T = (\tilde{e}_p^T \ \tilde{e}_s^T \ \tilde{e}_d^T \ \tilde{e}_y^T)$ and

$$\tilde{A}_\mu = \begin{pmatrix} A_\mu - B_\mu K_\mu & -B_\mu K_\mu & -B_\mu & 0 \\ 0 & A_\mu & B_\mu & -H_\mu^1 \\ 0 & -H_\mu^2 C_\mu & -H_\mu^2 D_\mu & 0 \\ 0 & C_\mu & D_\mu & -I \end{pmatrix} \quad (12)$$

The main proposed result can now be established.

Theorem 1: The tracking error $e_p(t)$, the state $e_s(t)$ and fault $e_d(t)$ estimation errors asymptotically converge to zero if there exists some matrices $X = X^T \geq 0$, $P_6 = P_6^T \geq 0$, $P_{11} = I$, P_{13} , P_{14} , P_{15} , P_{16} , \bar{H}_i^1 , \bar{H}_j^2 and K_j such that the following LMI are satisfied for all $i, j = 1, 2, \dots, r$:

$$\begin{pmatrix} \Upsilon_{ij}^{(1,1)} & * \\ \Upsilon_i^{(2,1)} & \Upsilon^{(2,2)} \end{pmatrix} < 0 \quad (13)$$

where:

$$\Upsilon_{ij}^{(1,1)} = \begin{pmatrix} A_i X + X A_i^T & * & * & 0 & * & * \\ -K_j^T B_i^T & \Sigma_i^{(2,2)} & * & * & 0 & 0 \\ -B_i^T & \Sigma_{ij}^{(3,2)} & \Sigma_{ij}^{(3,3)} & * & 0 & 0 \\ 0 & \Sigma_i^{(4,2)} & \Sigma_i^{(4,3)} & -P_{16} - P_{16}^T & 0 & 0 \\ K_j^T B_i^T & 0 & 0 & 0 & -I & 0 \\ X & 0 & 0 & 0 & 0 & -I \end{pmatrix},$$

$$\Upsilon_i^{(2,1)} = \begin{pmatrix} X & 0 & 0 & 0 & 0 & 0 \\ 0 & P_{13}^T C_i & 0 & 0 & 0 & 0 \\ X & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & P_{13}^T D_i & 0 & 0 & 0 \\ X & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & P_{13}^T & 0 & 0 \end{pmatrix}$$

$$\Upsilon^{(2,2)} = \text{diag}(-I \ -I \ -I \ -I \ -I \ -I)$$

$$\Sigma_i^{(2,2)} = P_6 A_i + A_i^T P_6 + P_{14}^T C_i + C_i^T P_{14}$$

$$\Sigma_{ij}^{(3,2)} = P_{15}^T C_i - \bar{H}_j^2 C_i + B_i^T P_6 + D_i^T P_{14}$$

$$\Sigma_{ij}^{(3,3)} = D_i^T P_{15} - \bar{H}_j^2 D_i - D_i^T (\bar{H}_j^2)^T$$

$$\Sigma_i^{(4,2)} = P_{16}^T C_i - \bar{H}_i^1 - P_{14}$$

$$\Sigma_i^{(4,3)} = P_{16}^T D_i - P_{15}$$

Proof: Let us consider the following candidate quadratic Lyapunov function:

$$V(\tilde{e}(t)) = \tilde{e}^T(t) E P \tilde{e}(t) \quad (14)$$

with:

$$E P = P^T E \geq 0 \quad (15)$$

A way to provide easily LMI conditions is to consider the matrix P structure as follows:

$$P = \begin{pmatrix} P_1 & 0 & 0 & 0 \\ 0 & P_6 & 0 & 0 \\ 0 & 0 & P_{11} & 0 \\ P_{13} & P_{14} & P_{15} & P_{16} \end{pmatrix} \quad (16)$$

According to (15), it follows that $P_1 = P_1^T \geq 0$, $P_6 = P_6^T \geq 0$, $P_{11} = P_{11}^T \geq 0$ and P_{13} , P_{14} , P_{15} , P_{16} are free slack matrices.

The tracking error $e_p(t)$, the state $e_s(t)$ and the fault $e_d(t)$ estimation errors converge asymptotically to zero if:

$$\dot{V}(\tilde{e}(t)) = \dot{\tilde{e}}^T(t) E P \tilde{e}(t) + \tilde{e}^T(t) E P \dot{\tilde{e}}(t) < 0 \quad (17)$$

With (11) and (14), the inequality (17) becomes:

$$\tilde{e}^T(t) (\tilde{A}_\mu^T P + P^T \tilde{A}_\mu) \tilde{e}(t) < 0 \quad (18)$$

The inequality (18) is fulfilled if:

$$\tilde{A}_\mu^T P + P^T \tilde{A}_\mu < 0 \quad (19)$$

Indeed, with (12) and (16) the inequality (19) becomes:

$$\begin{pmatrix} \bar{\Psi}_{\mu\mu}^{(1,1)} & * & * & * \\ C_\mu^T P_{13} - K_\mu^T B_\mu^T P_1 & \bar{\Psi}_\mu^{(2,2)} & * & * \\ D_\mu^T P_{13} - B_\mu^T P_1 & \bar{\Psi}_{\mu\mu}^{(3,2)} & \bar{\Psi}_{\mu\mu}^{(3,3)} & * \\ -P_{13} & \bar{\Psi}_\mu^{(4,2)} & P_{16}^T D_\mu - P_{15} & -P_{16} - P_{16}^T \end{pmatrix} < 0 \quad (20)$$

$$\text{where: } \bar{\Psi}_{\mu\mu}^{(1,1)} = P_1 A_\mu + A_\mu^T P_1 - P_1 B_\mu K_\mu - K_\mu^T B_\mu^T P_1$$

$$\bar{\Psi}_\mu^{(2,2)} = P_{14}^T C + C_\mu^T P_{14} + P_6 A_\mu + A_\mu^T P_6$$

$$\bar{\Psi}_{\mu\mu}^{(3,2)} = B_\mu^T P_6 - P_{11} H_\mu^2 C_\mu + P_{15}^T C_\mu + D_\mu^T P_{14}$$

$$\bar{\Psi}_{\mu\mu}^{(3,3)} = P_{15}^T D_\mu + D_\mu^T P_{15} - P_{11} H_\mu^2 D_\mu - D_\mu^T (H_\mu^2)^T P_{11}$$

$$\bar{\Psi}_\mu^{(4,2)} = P_{16}^T C_\mu - (H_\mu^1)^T P_6 - P_{14}$$

Multiplying (20) left and right by $\text{diag}(X \ I \ I \ I)$ where $X = P_1^{-1}$, and considering $P_{11} = P_{11}^T = I > 0$ and the bijective variable changes $(H_\mu^1)^T P_6 = \bar{H}_\mu^1$, $P_{11} H_\mu^2 = \bar{H}_\mu^2$, (20) yields:

$$\begin{pmatrix} \Phi_{\mu\mu}^{(1,1)} & * & * & * \\ C_\mu^T P_{13} X - K_\mu^T B_\mu^T & \bar{\Psi}_\mu^{(2,2)} & * & * \\ D_\mu^T P_{13} X - B_\mu^T & \Phi_{\mu\mu}^{(3,2)} & \Phi_{\mu\mu}^{(3,3)} & * \\ -P_{13} X & \Phi_\mu^{(4,2)} & P_{16}^T D_\mu - P_{15} & -P_{16} - P_{16}^T \end{pmatrix} < 0 \quad (21)$$

where:

$$\Phi_{\mu\mu}^{(1,1)} = A_\mu X + X A_\mu^T - B_\mu K_\mu X - X K_\mu^T B_\mu^T$$

$$\Phi_{\mu\mu}^{(3,2)} = B_\mu^T P_6 - \bar{H}_\mu^2 C_\mu + P_{15}^T C_\mu + D_\mu^T P_{14}$$

$$\Phi_{\mu\mu}^{(3,3)} = P_{15}^T D_\mu + D_\mu^T P_{15} - \bar{H}_\mu^2 D_\mu - D_\mu^T (\bar{H}_\mu^2)^T$$

$$\Phi_\mu^{(4,2)} = P_{16}^T C_\mu - \bar{H}_\mu^1 - P_{14}$$

Applying lemma 1 and considering $\delta_1 = \delta_2 = \delta_3 = \delta_4 = I$, the inequality (21) is implied by:

$$\begin{pmatrix} \bar{\Phi}_{\mu\mu}^{(1,1)} & * & * & 0 \\ -K_\mu^T B_\mu^T & \bar{\Phi}_{\mu\mu}^{(2,2)} & * & * \\ -B_\mu^T & \Phi_{\mu\mu}^{(3,2)} & \bar{\Phi}_{\mu\mu}^{(3,3)} & * \\ 0 & P_{16}^T C_\mu - \bar{H}_\mu^1 - P_{14} & P_{16}^T D_\mu - P_{15} & \bar{\Phi}_{\mu\mu}^{(4,4)} \end{pmatrix} < 0 \quad (22)$$

where:

$$\bar{\Phi}_{\mu\mu}^{(1,1)} = \begin{pmatrix} A_\mu X + X A_\mu^T + \delta_1 B_\mu K_\mu K_\mu^T B_\mu^T + \delta_1^{-1} X X \\ + \delta_2^{-1} X X + \delta_3^{-1} X X + \delta_4^{-1} X X \end{pmatrix}$$

$$\bar{\Phi}_{\mu\mu}^{(2,2)} = P_{14}^T C + C_\mu^T P_{14} + P_6 A_\mu + A_\mu^T P_6 + \delta_2 C_\mu^T P_{13} P_{13}^T C_\mu$$

$$\bar{\Phi}_{\mu\mu}^{(3,3)} = D_\mu^T P_{15} - \bar{H}_\mu^2 D_\mu - D_\mu^T (\bar{H}_\mu^2)^T + \delta_3 D_\mu^T P_{13} P_{13}^T D_\mu$$

$$\bar{\Phi}_{\mu\mu}^{(4,4)} = \delta_4 P_{13} P_{13}^T - P_{16} - P_{16}^T$$

Applying Schur complement [13] on the BMI terms $\bar{\Phi}_{\mu\mu}^{(1,1)}$, $\bar{\Phi}_{\mu\mu}^{(2,2)}$ and $\bar{\Phi}_{\mu\mu}^{(3,3)}$, the sufficient LMI conditions proposed in the theorem 1 hold. \square

Remark 1: New LMI conditions can be provided from the ones given in theorem 1 by considering only the diagonal matrices

of (16) (i.e. P_1 , P_6 , P_{11} and P_{16}). This result is given in corollary 1.

Corollary 1: The tracking error $e_p(t)$, the state $e_s(t)$ and the fault $e_d(t)$ estimation errors convergence asymptotically to zero if there exists the matrices $X = X^T \geq 0$, $P_6 = P_6^T \geq 0$, $P_{11} = I$, P_{16} , \bar{H}_i^1 , \bar{H}_j^2 and K_μ such that the following LMI are satisfied for all $i, j = 1, 2, \dots, r$

$$\begin{pmatrix} \Theta_i^{(1,1)} & * & * & 0 & * & * \\ -K_j^T B_i^T & \Theta_i^{(2,2)} & * & * & 0 & 0 \\ -B_i^T & \Theta_i^{(3,2)} & \Theta_{ij}^{(3,3)} & * & 0 & 0 \\ 0 & P_{16}^T C_i - \bar{H}_i^1 & P_{16}^T D_i & -P_{16} - P_{16}^T & 0 & 0 \\ K_j^T B_i^T & 0 & 0 & 0 & -I & 0 \\ X & 0 & 0 & 0 & 0 & -I \end{pmatrix} < 0 \quad (23)$$

where: $\Theta_i^{(1,1)} = A_i X + X A_i^T$, $\Theta_i^{(2,2)} = P_6 A_i + A_i^T P_6$, $\Theta_i^{(3,2)} = B_i^T P_6 - \bar{H}_j^2 C_i$ and $\Theta_{ij}^{(3,3)} = -\bar{H}_j^2 D_i - D_i^T (\bar{H}_j^2)^T$.

Remark 2: To ensure the stability of (3) even if faults occur, one has to check the existence of $\text{diag}(X P_2 I)$ in theorem 5.4 of [8] or the matrix P given by (16) in the proposed approach. Indeed, the proposed approach (theorem 1) introduces some additional free slack variables to relax the existing LMI conditions. This conservatism reduction can be shown mathematically by considering in theorem 1 that $P_{13} = 0$, $P_{14} = 0$, $P_{15} = 0$ and $P_{16} = 0$. Then, the inequality (22) can be rewritten as:

$$\begin{pmatrix} \bar{\Phi}_{\mu\mu}^{(1,1)} & * & * & 0 \\ -K_\mu^T B_\mu^T & \bar{\Phi}_{\mu}^{(2,2)} & * & * \\ -B_\mu^T & \bar{\Phi}_{\mu\mu}^{(3,2)} & \bar{\Phi}_{\mu\mu}^{(3,3)} & 0 \\ 0 & -\bar{H}_\mu^1 & 0 & 0 \end{pmatrix} < 0 \quad (24)$$

with $\bar{\Phi}_{\mu\mu}^{(1,1)} = A_\mu X + X A_\mu^T + \delta_1 B_\mu K_\mu^T B_\mu^T + \delta_1^{-1} X X$, $\bar{\Phi}_{\mu}^{(2,2)} = P_6 A_\mu + A_\mu^T P_6$ and $\bar{\Phi}_{\mu\mu}^{(3,3)} = -\bar{H}_\mu^2 D_\mu - D_\mu^T (\bar{H}_\mu^2)^T$.

Replacing $e_y(t)$ by its expression given in (24), one obtains the LMI conditions of theorem 5.4 [8].

IV. SIMULATION RESULTS

In order to show the effectiveness and the applicability of the proposed approaches, let us consider the system (2) with

$$A_1 = \begin{pmatrix} a & 1 & 1 \\ 1 & -3 & 0 \\ 2 & 1 & -8 \end{pmatrix}, A_2 = \begin{pmatrix} -3 & 2 & 2 \\ 0 & -3 & 0.2 \\ 0.5 & 2 & -5 \end{pmatrix}, B_1 = \begin{pmatrix} 0 \\ 1 \\ 0.25 \end{pmatrix},$$

$$B_2 = \begin{pmatrix} 1 \\ 1 \\ b \end{pmatrix}, C_1 = \begin{pmatrix} -1 \\ 0.5 \\ 0 \end{pmatrix}^T, C_2 = \begin{pmatrix} -1 \\ -0.5 \\ 0 \end{pmatrix}^T, D_1 = -0.8,$$

$$D_2 = -0.5, \quad \mu_1(u(t)) = \frac{1 - \tanh(0.5 - u(t))}{2} \quad \text{and}$$

$$\mu_2(u(t)) = 1 - \mu_1(u(t)).$$

a and b are two model parameters. Firstly, our aim is to compare the conservatism of the approach given in theorem 5.4 of [8] and the proposed theorem 1 and corollary 1.

Let us consider $a \in [-2 \ -0.6]$ and $b \in [-2 \ 0]$, using Matlab LMI Toolbox the obtained feasibility fields are presented in Fig.2 and show that the proposed approaches are less conservative than in [8].

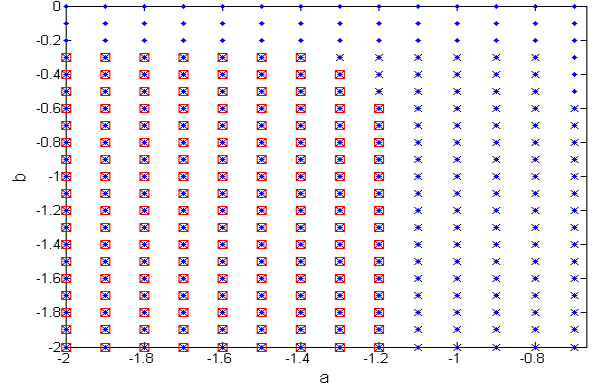


Fig.2. Feasibility fields ; • Theorem 1, × Corollary 1 and □ Theorem 5.4 of [8]

Secondly, in order to illustrate the effectiveness of the fault tolerant controller compared to a classical one, a passive FTC controller is designed as described in appendix, in order to minimize the L_2 -gain from the fault to the tracking error. The obtained results are compared with those issued from the proposed active FTC controller.

In the fault free case, it can be seen on Fig.3 that both passive and active FTC controllers ensure the system stabilization. The simulation is ran for $a = -2$, $b = -0.5$, a nominal input given by $u(t) = \sin(\cos(2t)0.5t)$ and the LMI problem is solved with Matlab LMI Toolbox.

In order to compare passive and active FTC control facing the occurrence of a fault, a piecewise constant fault $f(t)$, occurring at $t = 4$ is considered. The simulation results are displayed on the Fig. 4, 5, 6, 7. The effectiveness of the proposed FTC design can be seen on Fig.4, whereas the passive FTC fails to ensure trajectory tracking when $f(t)$ occurs.

V. CONCLUSION

In this paper, a trajectory tracking fault tolerant controller design approaches have been proposed for faulty T-S models with measurable premise variables. The objective is to ensure the tracking between the faulty system states and one of healthy reference model. The proposed LMI approaches are less conservative. This improvement is due to the considered “virtual dynamics” on the output error allows introducing slack variables in the Lyapunov function and decoupling the observer gains and the system matrices.

The efficiency of the FTC law comparing with classical one is illustrated with a numerical T-S model whose input is corrected by a fault.

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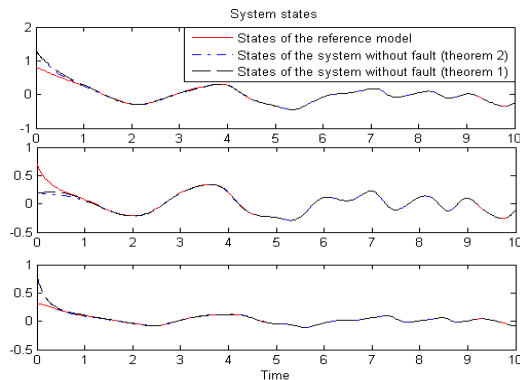


Fig.3. Comparison of the reference model states (no fault), the system states with FTC (theorem 1) and system states with classical control law (theorem 2).

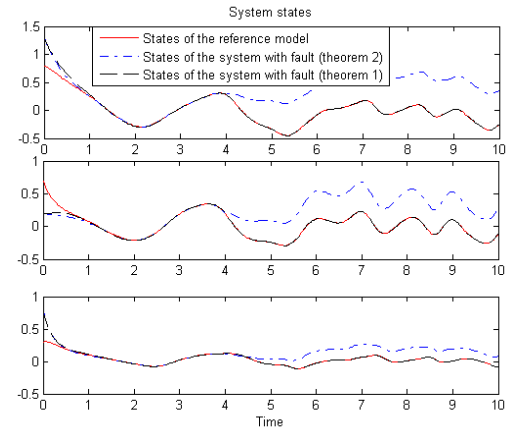


Fig.4. Comparison of the reference model state (no fault), the faulty system state with FTC (theorem 1) and the faulty system state with classical control law (theorem 2).

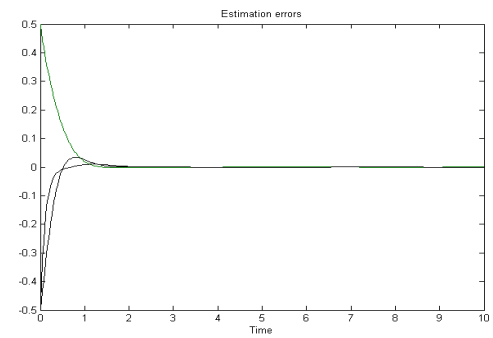


Fig.5. Estimation errors

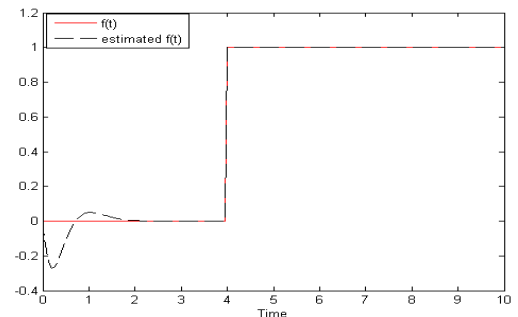


Fig.6. Fault and its estimation

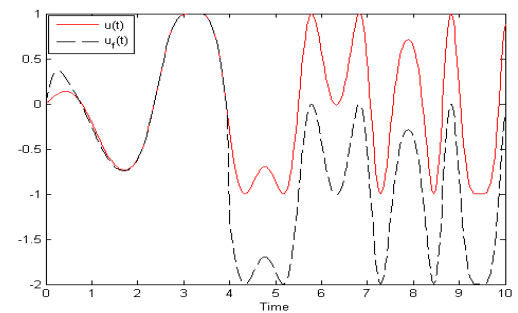


Fig.7. Nominal control input and FTC input

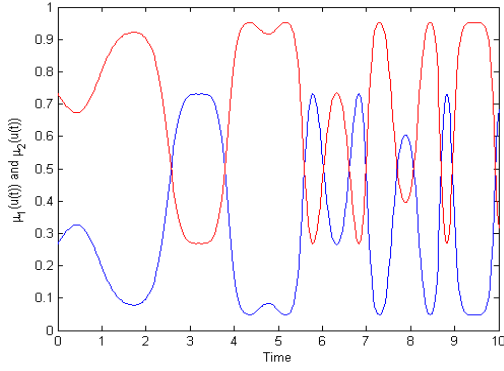


Fig.8. Membership function evolution

VI. APPENDIX. CLASSICAL CONTROLLER DESIGN APPROACH

The classical controller design methodology is based on the following scheme.

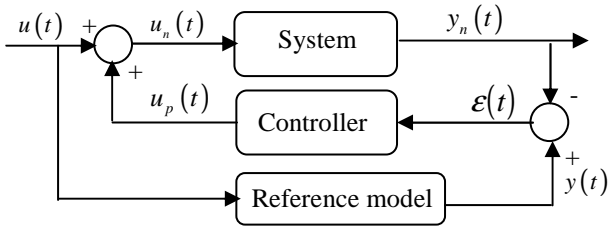


Fig.9. Classical controller design scheme

The system state representation is given by:

$$\begin{cases} \dot{x}_n(t) = \sum_{i=1}^r \mu_i(\xi(t)) (A_i x_n(t) + B_i (u_n(t) + f(t))) \\ y_n(t) = \sum_{i=1}^r \mu_i(\xi(t)) (C_i x_n(t) + D_i (u_n(t) + f(t))) \end{cases} \quad (25)$$

Let us define the state and output tracking errors between (2) and (25) by $e(t) = x(t) - x_n(t)$ and $\varepsilon(t) = y(t) - y_n(t)$ respectively. To ensure the tracking of the reference model, we consider the following control law $u_p(t) = K_p \varepsilon(t)$.

Introducing a “virtual dynamic” on $\varepsilon(t)$, one can obtain:

$$\bar{E} \dot{\bar{e}}(t) = \Gamma_\mu \bar{e}(t) - \Lambda_\mu f(t) \quad (26)$$

where $\bar{E} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$, $\bar{e}(t) = \begin{pmatrix} e(t) \\ \varepsilon(t) \end{pmatrix}$, $\Lambda_\mu = \begin{pmatrix} B_\mu \\ D_\mu \end{pmatrix}$ and

$$\Gamma_\mu = \begin{pmatrix} A_\mu & -B_\mu K_p \\ C_\mu & -(D_\mu K_p + I) \end{pmatrix}.$$

The LMI conditions leading to synthesize the controller K_p under the L_2 norm bound are given in the following theorem 2.

Theorem 2: The tracking error $e_p(t)$ asymptotically converges to zero if there exists some matrices $P_1 = P_1^T \geq 0$, P_3 and K_p and a positive scalar $\bar{\gamma}$ such that the following LMI are satisfied for all $i = 1, 2, \dots, r$

$$\begin{pmatrix} \Theta_i & * & * & * & 0 & 0 & 0 \\ P_2^T C_i & P_2 + P_2^T & * & 0 & * & * & * \\ -B_i^T P_1 & -D_i^T P_2 & -\bar{\gamma} & 0 & 0 & 0 & 0 \\ P_1 & 0 & 0 & -I & 0 & 0 & 0 \\ 0 & B_i K_p & 0 & 0 & -I & 0 & 0 \\ 0 & P_2 & 0 & 0 & 0 & -I & 0 \\ 0 & D_i K_p & 0 & 0 & 0 & 0 & -I \end{pmatrix} < 0 \quad (27)$$

where $\Theta_i = P_1 A_i + A_i^T P_1 + I$.

Proof: Let us consider the following candidate quadratic Lyapunov function:

$$V(e(t)) = e^T(t) \bar{E} \bar{P} e(t) \quad (28)$$

with $\bar{E} \bar{P} = \bar{P}^T \bar{E} \geq 0$ (29)

we consider $\bar{P} = \begin{pmatrix} P_1 & 0 \\ 0 & P_2 \end{pmatrix}$. According to (29), one can find

that $P_1 = P_1^T \geq 0$. It is well known that the L_2 -gain from $f(t)$ to $e(t)$ is bounded by γ if [13]:

$$e^T(t) \bar{E} \bar{P} e(t) + e^T(t) \bar{E} \bar{P} \dot{e}(t) + e^T(t) \dot{e}(t) - \gamma^2 f^T(t) f(t) < 0 \quad (30)$$

Considering (29) and substituting (26) in (30), one can obtain:

$$\begin{pmatrix} \bar{P}^T \Gamma_\mu + \Gamma_\mu^T \bar{P} + \bar{E} & * \\ -\Lambda_\mu^T \bar{P} & -\gamma^2 \end{pmatrix} < 0 \quad (31)$$

The mathematical development of (31) with (26) and (29) leads to:

$$\begin{pmatrix} P_1 A_\mu + A_\mu^T P_1 & * & * \\ P_2^T C_\mu - K_p^T B_\mu^T P_1 & -(K_p^T D_\mu^T - I) P_2 - P_2^T (D_\mu K_p - I) & * \\ -B_\mu^T P_1 & -D_\mu^T P_2 & -\gamma^2 \end{pmatrix} < 0 \quad (32)$$

Applying Lemma 1 then Schur complement on (32), the sufficient LMI conditions proposed in theorem 2 holds. \square