

Decoupled Nested LMI conditions for Takagi-Sugeno Observer Design

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Abstract: This work extends recent investigations on control design of continuous-time nonlinear models to non-quadratic observer design. The models under consideration are exactly rewritten in the Takagi-Sugeno form. By means of the Finsler's Lemma or a Tustin-like transformation, the progressively complex observer gains and the Lyapunov function are decoupled, thus providing the flexibility of using a quadratic Lyapunov functions while preserving the non-quadratic nature of the observer. Conditions obtained are expressed as linear matrix inequalities which are efficiently solved by convex optimization techniques. Examples are provided to show the effectiveness of the proposed approach.

1. INTRODUCTION

For many years, Takagi-Sugeno fuzzy models (TS) (Takagi and Sugeno 1985) have been an important topic of many research works in the control community due to their capacity to exactly represent an important class of nonlinear systems in a compact set of the state space. The TS representation is usually obtained via the sector nonlinearity approach (Taniguchi et al. 2001): it is an exact rewriting of the nonlinear model, not an approximation. A TS fuzzy model is composed of a set of linear models blended together with memberships functions (MFs) which contain the model nonlinearities and hold the convex sum property (Tanaka and Wang 2001). Taking advantage of their convex structure along with the direct Lyapunov method, TS models permit to obtain linear matrix inequality (LMI) conditions for stability analysis as well as for controller and observer design (Wang et al. 1996), (Tanaka et al. 1998). Getting LMI conditions is convenient since they are efficiently solved via convex optimization techniques (Boyd et al. 1994).

The aforementioned conditions are only sufficient due to several reasons: the way the MFs are dropped off or considered in the analysis, the model construction as well as the choice of the Lyapunov function (Feng et al. 2005), (Sala et al. 2005). Several works have been developed in order to tackle these sources of conservatism: diverse ways to obtain LMIs from nested convex sums (Tuan et al. 2001), (Liu and Zhang 2003), (Sala and Ariño 2007); different representations of the convex models such as descriptor (Guelton et al. 2009) or polynomial forms (Tanaka et al. 2009); more general Lyapunov functions such as piecewise (Johansson et al. 1999), line-integral (Rhee and Won 2006) and fuzzy (Tanaka et al. 2003), (Guerra and Vermeiren 2004).

In the continuous-time framework, non-quadratic Lyapunov functions have not met the development of the discrete-time domain (Guerra et al. 2009). The latter is due to the fact that the use of non-quadratic Lyapunov functions obliges to deal with the time-derivatives of the MFs (Blanco et al. 2001), a problem that has been considering in several works (Mozelli et al. 2009), (Bernal and Guerra 2010), (Lee et al. 2012).

This paper proposes an observer design scheme based on two former results: the Finsler's Lemma approach (Jaadari et al. 2012) and a Tustin-like transformation (Shaked 2001), (Márquez et al. 2013); it breaks the link between observer gains and the Lyapunov function.

This paper is organized as follows. Section 2 presents the TS model obtained by the sector nonlinearity methodology, provides basic notation and useful properties. In section 3 the main result in this paper is developed: it considers Finsler-based and Tustin-like approaches via a quadratic Lyapunov function for TS observers; moreover, thanks to a scheme of nested convex sums, it produces progressively more relaxed results. Section 4 gives some examples to illustrate the effectiveness of the proposed approaches, and finally, Section 5 briefs the paper results and discusses future work on the subject.

2. DEFINITIONS AND NOTATIONS

Consider a nonlinear model of the form

$$\begin{aligned}\dot{x}(t) &= f(z(t))x(t) + g(z(t))u(t) \\ y(t) &= e(z(t))x(t)\end{aligned}\tag{1}$$

with $f(\cdot)$, $g(\cdot)$ and $e(\cdot)$ being nonlinear functions, $x(t) \in \mathbb{R}^n$ the state vector, $u(t) \in \mathbb{R}^m$ the input vector,

$y(t) \in \mathbb{R}^o$ the output of the system, and $z(x(t)) \in \mathbb{R}^p$ the premise vector assumed to be bounded and smooth in a compact set C of the state space including the origin.

Let $z_j(\cdot) \in [\underline{z}_j, \overline{z}_j]$, $j \in \{1, \dots, p\}$ be the set of bounded nonlinearities in (1) belonging to C . Employing the sector nonlinearity approach (Taniguchi et al. 2001), the following weighting functions can be constructed

$$w_0^j(\cdot) = \frac{\overline{z}_j - z_j(\cdot)}{\overline{z}_j - \underline{z}_j}, \quad w_1^j(\cdot) = 1 - w_0^j(\cdot), \quad j \in \{1, \dots, p\}. \quad (2)$$

From the previous weights, the following MFs are defined:

$$h_i = h_{1+i_1+2+i_2+\dots+i_p, 2^{p-1}} = \prod_{j=1}^p w_{i_j}^j(z_j) \quad (3)$$

with $i \in \{1, \dots, 2^p\}$, $i_j \in \{0, 1\}$. These MFs satisfy the convex sum property $\sum_{i=1}^r h_i(\cdot) = 1$, $h_i(\cdot) \geq 0$ in C . Where convenient, convex sums will be denoted as $\Upsilon_z = \sum_{i=1}^r h_i(z(t)) \Upsilon_i$, their inverse as $\Upsilon_z^{-1} = \left(\sum_{i=1}^r h_i(z(t)) \Upsilon_i \right)^{-1}$, and with extended indexes as $\Upsilon_{\bar{z}} = \sum_{i_1=1}^r \sum_{i_2=1}^r \dots \sum_{i_q=1}^r h_{i_1}(z(t)) h_{i_2}(z(t)) \dots h_{i_q}(z(t)) \Upsilon_{i_1 i_2 \dots i_q}$.

Based on the previous definitions, an exact representation of (1) in C is given by the following continuous-time T-S model:

$$\dot{x}(t) = \sum_{i=1}^r h_i(z(t)) (A_i x(t) + B_i u(t)) = A_z x(t) + B_z u(t) \quad (4)$$

$$y(t) = \sum_{i=1}^r h_i(z(t)) C_i x(t)$$

with $r = 2^p \in \mathbb{N}$ representing the number of linear models and (A_i, B_i, C_i) , $i = 1, \dots, r$ a set of matrices of proper dimensions.

For brevity, an asterisk (*) for inline expressions denotes the transpose of the terms on its left-hand side; for matrix expressions denotes the transpose of its symmetric block-entry. Should a matrix expression be involved with symbols "<" and ">", they will stand for negative and positive-definiteness, respectively. When convenient, arguments will be omitted.

The following properties will be used to develop the main results:

Property 1 (Schur complement): Let $P \in R^{n \times n} : P = P^T > 0$, $X \in R^{m \times n}$ a full rank matrix, and $Q \in R^{n \times n}$, then (Boyd et al. 1994):

$$\begin{cases} Q - X^T P^{-1} X > 0 \\ P > 0 \end{cases} \Leftrightarrow \begin{bmatrix} Q & (*) \\ X & P \end{bmatrix} > 0 \quad (5)$$

Property 2: Given $P = P^T > 0$, then

$$Q^T P^{-1} Q \geq Q^T + Q - P \quad (6)$$

Property 3 (Finsler's Lemma): Let $x \in \mathbb{R}^n$, $Q = Q^T \in \mathbb{R}^{n \times n}$, and $R \in \mathbb{R}^{m \times n}$ such that $rank(R) < n$; the following expressions are equivalent:

- a) $x^T Q x < 0, \forall x \in \{x \in \mathbb{R}^n : x \neq 0, R x = 0\}$
- b) $\exists X \in \mathbb{R}^{n \times m} : Q + X R + R^T X^T < 0$.

It is well-known that TS-LMI based controller design usually leads to inequalities containing multiple nested convex sums. For instance, given matrix expressions $\Upsilon_{i_0 i_1 \dots i_q}$, $i_0, i_1, \dots, i_q \in \{1, \dots, r\}$, the following inequality may arise:

$$\sum_{i_0=1}^r \sum_{i_1=1}^r \sum_{i_2=1}^r \dots \sum_{i_q=1}^r h_{i_0}(z(t)) h_{i_1}(z(t)) \dots h_{i_q}(z(t)) \Upsilon_{i_0 i_1 \dots i_q} < 0 \quad (7)$$

The sign of such expressions should be established via LMIs, which implies that the MFs therein should be adequately dropped-off: conditions thus obtained will be therefore only sufficient. This is why selecting a proper way to perform this task is important to reduce conservatism. When double sums are involved ($q = 1$), a good compromise for guaranteeing (7) without adding slack variables is given by the following lemma:

Relaxation 1 (Tuan et al. 2001): Let $\Upsilon_{i_0 i_1}$, $i_0, i_1 \in \{1, \dots, r\}$ be matrices of the same size. Condition (7) is verified for $q = 1$ if:

$$\begin{aligned} \Upsilon_{i_0 i_0} < 0, \quad \forall i_0 \in \{1, \dots, r\} \\ \frac{2}{r-1} \Upsilon_{i_0 i_0} + \Upsilon_{i_0 i_1} + \Upsilon_{i_1 i_0} < 0, \quad \forall (i_0, i_1) \in \{1, \dots, r\}^2, i_0 \neq i_1 \end{aligned} \quad (8)$$

Should more than two nested convex sums be involved, a generalization of the sum relaxation in (Tanaka et al. 1998) will be used (Sala and Ariño 2007):

Relaxation 2 (Sala and Ariño 2007): Let $\Upsilon_{i_0 i_1 \dots i_q}$, $i_0, i_1, \dots, i_q \in \{1, \dots, r\}$ be matrices of the same size and $P(i_0, i_1, \dots, i_q)$ be the set of all permutations of the indexes i_0, i_1, \dots, i_q . Condition (7) is verified if:

$$\sum_{i_0 i_1 \dots i_q \in P(i_0, i_1, \dots, i_q)} \Upsilon_{i_0 i_1 \dots i_q} < 0, \quad \forall (i_0, i_1, \dots, i_q) \in \{1, \dots, r\}^{q+1}. \quad (9)$$

3. OBSERVER DESIGN

As in (Jaadari et al. 2012) and (Márquez et al. 2013), Finsler's Lemma and Tustin-like transformation will be used to relax the link between the Lyapunov function and the observer design. A quadratic Lyapunov function will be considered; it will outline the way to involve the approaches in the desired decoupling.

The proposed observer of the TS model in (4) has the following structure:

$$\begin{aligned} \dot{\hat{x}}(t) &= A_z \hat{x}(t) + B_z u(t) + H_{\bar{z}}^{-1} K_{\bar{z}} (y(t) - \hat{y}(t)) \\ \hat{y}(t) &= C_z \hat{x}(t) \end{aligned} \quad (10)$$

with $\hat{x}(t) \in \mathbb{R}^n$ as the observer state, $e(t) = x(t) - \hat{x}(t)$ as the estimation error, $H_{\bar{z}} \in \mathbb{R}^{n \times n}$ and $K_{\bar{z}} \in \mathbb{R}^{o \times n}$ matrix functions of the premise vector to be designed in the sequel. Therefore, the estimation error dynamics is described as:

$$\dot{e}(t) = (A_z - H_{\bar{z}}^{-1} K_{\bar{z}} C_z) e(t). \quad (11)$$

3.1 Finsler's lemma

Consider the quadratic Lyapunov function (QLF) candidate with $P = P^T > 0$

$$V(x(t)) = e(t)^T P e(t). \quad (12)$$

Theorem 1 (QLF, generalized observer design via Finsler's lemma): The estimation error model (11) with $H_{\bar{z}}$ and $K_{\bar{z}}$ is asymptotically stable if $\exists \varepsilon > 0$, and matrices $P = P^T > 0$, $H_{i_1 i_2 \dots i_q}$, and $K_{i_1 i_2 \dots i_q}$, $i_1, \dots, i_q \in \{1, \dots, r\}$ of proper dimensions such that (9) holds with

$$\Upsilon_{i_1 i_2 \dots i_q} = \begin{bmatrix} H_{i_1 i_2 \dots i_q} A_{i_0} - K_{i_1 i_2 \dots i_q} C_{i_0} + (*) & (*) \\ \left(P - H_{i_1 i_2 \dots i_q}^T \right) & \left(H_{i_1 i_2 \dots i_q} \right) \\ \left(+ \varepsilon (H_{i_1 i_2 \dots i_q} A_{i_0} - K_{i_1 i_2 \dots i_q} C_{i_0}) \right) & -\varepsilon \left(+ H_{i_1 i_2 \dots i_q}^T \right) \end{bmatrix}. \quad (13)$$

Proof: Consider the quadratic Lyapunov function in (12); its time-derivative will thus be negative if:

$$\dot{V} = \begin{bmatrix} e \\ \dot{e} \end{bmatrix}^T \begin{bmatrix} 0 & P \\ P & 0 \end{bmatrix} \begin{bmatrix} e \\ \dot{e} \end{bmatrix} < 0 \quad (14)$$

all together with the following restriction

$$\begin{bmatrix} A_z - H_{\bar{z}}^{-1} K_{\bar{z}} C_z & -I \end{bmatrix} \begin{bmatrix} e \\ \dot{e} \end{bmatrix} = 0, \quad (15)$$

arising from (11). Inequality (14) under equality constraint (15) holds is equivalent to through Finsler's Lemma:

$$\begin{bmatrix} 0 & P \\ P & 0 \end{bmatrix} + \begin{bmatrix} U \\ W \end{bmatrix} \begin{bmatrix} A_z - H_{\bar{z}}^{-1} K_{\bar{z}} C_z & -I \end{bmatrix} + (*) < 0. \quad (16)$$

Let $U = H_{\bar{z}}$ and $W = \varepsilon H_{\bar{z}}$ with $\varepsilon > 0$, then (16) yields

$$\begin{bmatrix} H_{\bar{z}} A_z - K_{\bar{z}} C_z + (*) & (*) \\ P - H_{\bar{z}}^T + \varepsilon (H_{\bar{z}} A_z - K_{\bar{z}} C_z) & -\varepsilon (H_{\bar{z}} + H_{\bar{z}}^T) \end{bmatrix} < 0. \quad (17)$$

Conditions (9) with $\Upsilon_{i_1 i_2 \dots i_q}$ defined as in (13) guarantee the inequality above, thus producing the desired result. \square

Remark 1: Parameter ε has been introduced in the aforementioned development to naturally include the

quadratic case. Effectively, with $H_{\bar{z}} = P$ and $K_{\bar{z}} = K_z$, the Schur complement of (17) satisfies the following property:

$$P A_z - K_z C_z + (*) + \frac{1}{2} \varepsilon (P A_z - K_z C_z)^T P^{-1} (*) < 0, \quad (18)$$

which proves the referred inclusion when $\varepsilon > 0$ is enough small.

3.2 Tustin-like transformation

Theorem 2 (QLF, generalized observer design via Tustin-like transformation): The estimation error model (11) with $H_{\bar{z}}$ and $K_{\bar{z}}$ is asymptotically stable if $\exists \varepsilon > 0$, and matrices $P = P^T > 0$, $H_{i_1 i_2 \dots i_q}$, and $K_{i_1 i_2 \dots i_q}$, $i_1, \dots, i_q \in \{1, \dots, r\}$ of proper dimensions such that (9) holds with

$$\Upsilon_{i_1 i_2 \dots i_q} = \begin{bmatrix} -P & (*) \\ \left(-H_{i_1 i_2 \dots i_q} \right) & \left(P - H_{i_1 i_2 \dots i_q} \right) \\ \left(-\varepsilon (H_{i_1 i_2 \dots i_q} A_{i_0} - K_{i_1 i_2 \dots i_q} C_{i_0}) \right) & \left(-H_{i_1 i_2 \dots i_q}^T \right) \end{bmatrix}. \quad (19)$$

Proof: Consider the quadratic Lyapunov function in (12); derivative $\dot{V}(x(t)) < 0$ is satisfied if:

$$P(A_z - H_{\bar{z}}^{-1} K_{\bar{z}} C_z) + (*) < 0. \quad (20)$$

Considering a small enough $\varepsilon > 0$, it is clear that the following condition is equivalent to (20):

$$P(A_z - H_{\bar{z}}^{-1} K_{\bar{z}} C_z) + (*) + \varepsilon (A_z - H_{\bar{z}}^{-1} K_{\bar{z}} C_z)^T P (*) < 0 \quad (21)$$

from which the next rewriting can be done multiplying by ε and adding $P - P$:

$$\begin{aligned} \varepsilon P(A_z - H_{\bar{z}}^{-1} K_{\bar{z}} C_z) + (*) \\ + \varepsilon^2 (A_z - H_{\bar{z}}^{-1} K_{\bar{z}} C_z)^T P (*) + P - P < 0 \end{aligned}$$

or rewritten:

$$\left(I + \varepsilon (A_z - H_{\bar{z}}^{-1} K_{\bar{z}} C_z)^T \right) P (*) - P < 0 \quad (22)$$

Thus by Schur complement (22) is equivalent to:

$$\begin{bmatrix} P & (*) \\ I + \varepsilon (A_z - H_{\bar{z}}^{-1} K_{\bar{z}} C_z)^T & P^{-1} \end{bmatrix} > 0, \quad (23)$$

which after pre-multiplication by $\begin{bmatrix} I & 0 \\ 0 & H_{\bar{z}} \end{bmatrix}$ and post-

multiplication by $\begin{bmatrix} I & 0 \\ 0 & H_{\bar{z}}^T \end{bmatrix}$ gives:

$$\begin{bmatrix} P & (*) \\ H_{\bar{z}} + \varepsilon (H_{\bar{z}} A_z - K_{\bar{z}} C_z) & H_{\bar{z}} P^{-1} H_{\bar{z}}^T \end{bmatrix} > 0. \quad (24)$$

Using the property (6) with $Q = H_{\bar{z}}$, it is clear that $H_{\bar{z}} P^{-1} H_{\bar{z}}^T \geq H_{\bar{z}} + H_{\bar{z}}^T - P$, which allows guaranteeing (24) if the following holds:

$$\begin{bmatrix} P & (*) \\ H_{\bar{z}} + \varepsilon(H_{\bar{z}}A_z - K_{\bar{z}}C_z) & H_{\bar{z}} + H_{\bar{z}}^T - P \end{bmatrix} > 0. \quad (25)$$

But (25) holds if relaxation (9) is applied with $\Upsilon_{i_0 i_1 \dots i_q}$ defined as in (19), which concludes the proof. \square

Remark 2: The inclusion of the quadratic case is also guaranteed. Proof is direct with a Schur complement using $H_{\bar{z}} = P$, $K_{\bar{z}} = K_z$ and the fact that ε can be as small as possible.

Remark 3: Results in this work are parameter-dependent LMI; their result depend on the choice of ε . Nevertheless, it has been proved in (de Oliveira and Skelton 2001) and (Oliveira et al. 2011) that a logarithmically spaced family of values, for instance $\varepsilon \in \{10^{-6}, 10^{-5}, \dots, 10^6\}$, is adequate to avoid an exhaustive search of feasible solutions, thus outperforming existing results.

4. EXAMPLES

In this section some examples are presented to show the effectiveness of the proposed observer design.

4.1 Example 1

Consider the following TS model:

$$\dot{x}(t) = \sum_{i=1}^2 h_i(z(t))(A_i x(t) + B_i u(t)) \quad (26)$$

$$\text{with } A_1 = \begin{bmatrix} 1 & 0 \\ 5+5a & 10-10b \end{bmatrix}, \quad A_2 = \begin{bmatrix} 1 & 0 \\ 5-5a & 10+10b \end{bmatrix},$$

$$C_1 = \begin{bmatrix} 1+a \\ 1+b \end{bmatrix}^T, \quad C_2 = \begin{bmatrix} 1-a \\ 1-b \end{bmatrix}^T, \quad i=1,2, \quad z_1 = \sin x_1,$$

$$w_0^1 = \frac{1-\sin x_1}{2}, \quad w_1^1 = 1-w_0^1, \quad h_1 = w_0^1, \quad h_2 = w_1^1 \text{ with } a \in [-1,1],$$

and $b \in [-1,1]$.

The following results were obtained considering $\varepsilon \in E = \{10^{-6}, \dots, 10^6\}$.

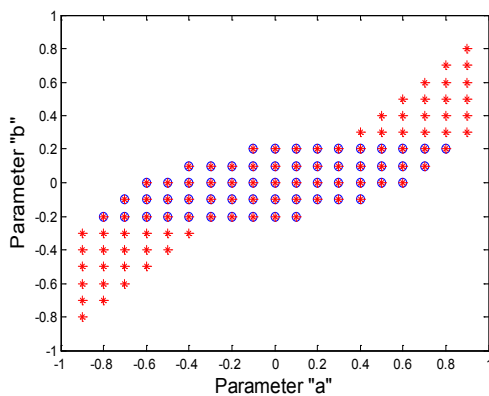


Fig. 1. Comparison: "*" for (13) with $q=3$, "o" for (13) with $q=1$.

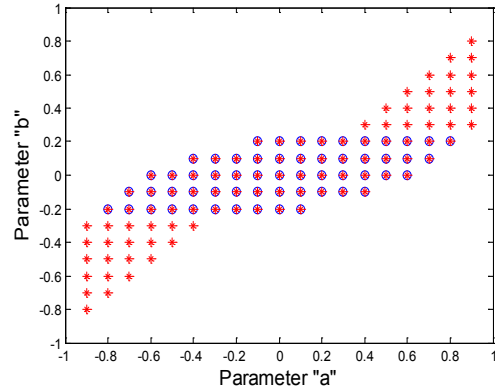


Fig. 2. Comparison: "*" for (19) with $q=3$, "o" for (19) with $q=1$.

Theorems 1 was compared considering two different values for q ($q=1$ and $q=3$). Fig. 1 shows that feasibility points of (13) with $q=1$ are included in the solutions presented considering $q=3$. Fig. 2 illustrate that solutions of (19) in theorem 2 considering $q=3$ overcome solutions with $q=1$.

Comparing results between Fig. 1 and Fig. 2, it is possible to observe that both approaches present the same feasibility region. Nevertheless, it is not possible to show theoretically that both approaches are equivalent or one included the other. They remain two different approaches to solve the problem.

4.2 Example 2

Consider the following TS model:

$$\dot{x}(t) = \sum_{i=1}^2 h_i(z(t))(A_i x(t) + B_i u(t)) \quad (27)$$

$$\text{with } A_1 = \begin{bmatrix} -1 & 1.5+a \\ 1.5 & -0.5-b \end{bmatrix}, \quad A_2 = \begin{bmatrix} -1 & 1.5-a \\ 1.5 & -0.5+b \end{bmatrix},$$

$$C_1 = \begin{bmatrix} 1-b \\ 0 \end{bmatrix}^T, \quad C_2 = \begin{bmatrix} 1+b \\ 0 \end{bmatrix}^T, \quad i=1,2, \quad z_1 = x_1^2, \quad w_0^1 = 1-x_1^2,$$

$$w_1^1 = 1-w_0^1, \quad h_1 = w_0^1, \quad \text{with } a \in [-3,3], \text{ and } b \in [-3,3].$$

A non-PDC control law $u(t) = F_z G_z^{-1} x(t) + 0.1 \sin(t)$ will be employed in order to stabilize the nonlinear system; the gains are calculated as in (Jaadari et al. 2012).

Theorem 2 with $q=3$ was compared with conditions in (Bergsten et al. 2002) considering parameter $\varepsilon \in E = \{10^{-6}, \dots, 10^6\}$. Fig. 3 shows that solutions of conditions in (Bergsten et al. 2002) are included in the solutions of (19) with $q=3$.

Using the Tustin-like approach in theorem 2 for state estimation (measured premises) and selecting the following parameters $q=3$, $a=1$, and $b=-2$, with $\varepsilon=0.1$, a feasible solution has been found, it is not possible with conditions in (Bergsten et al. 2002).

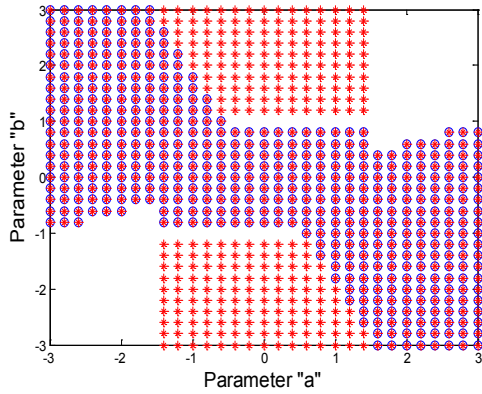


Fig. 3. Comparison: "*" for (19) with $q = 3$, "o" for conditions in (Bergsten et al. 2002).

The gains for the observer and Lyapunov matrix are given by

$$\begin{aligned}
 P &= \begin{bmatrix} 3.263 & -1.446 \\ -1.446 & 1.691 \end{bmatrix}, K_{111} = \begin{bmatrix} 12.680 \\ -0.444 \end{bmatrix}, K_{112} = \begin{bmatrix} 23.2469 \\ 1.3349 \end{bmatrix}, \\
 K_{121} &= \begin{bmatrix} 5.635 \\ -0.490 \end{bmatrix}, K_{122} = \begin{bmatrix} -59.069 \\ -3.511 \end{bmatrix}, K_{211} = \begin{bmatrix} 5.635 \\ -0.490 \end{bmatrix}, \\
 K_{212} &= \begin{bmatrix} 74.642 \\ -4.791 \end{bmatrix}, K_{221} = \begin{bmatrix} -74.642 \\ -4.791 \end{bmatrix}, K_{222} = \begin{bmatrix} -41.483 \\ 0.383 \end{bmatrix}, \\
 H_{111} &= \begin{bmatrix} 4.132 & -0.718 \\ -1.718 & 1.547 \end{bmatrix}, H_{112} = \begin{bmatrix} 14.767 & 4.902 \\ -26.003 & 121.072 \end{bmatrix}, \\
 H_{121} &= \begin{bmatrix} -7.699 & -2.406 \\ 11.170 & -33.948 \end{bmatrix}, H_{122} = \begin{bmatrix} 278.828 & -42.130 \\ -6.069 & -334.907 \end{bmatrix}, \\
 H_{211} &= \begin{bmatrix} -2.003 & -2.406 \\ 11.170 & -83.110 \end{bmatrix}, H_{212} = \begin{bmatrix} -112.002 & 20.350 \\ 1.741 & 143.536 \end{bmatrix}, \\
 H_{221} &= \begin{bmatrix} -161.947 & 20.350 \\ 1.741 & 195.714 \end{bmatrix}, H_{222} = \begin{bmatrix} 4.344 & 0.067 \\ -1.792 & 2.044 \end{bmatrix}.
 \end{aligned}$$

The estimation error for a trajectory of the states is presented in Fig 4.

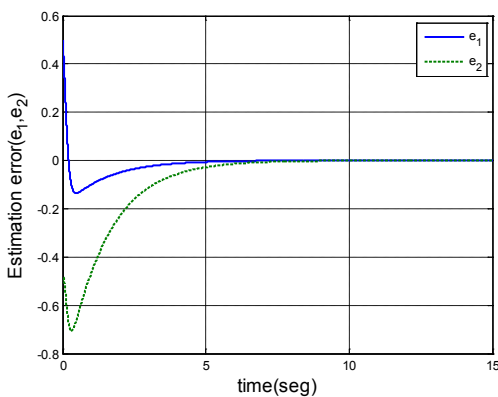


Fig. 4. Time evolution of the estimation error in Example 2.

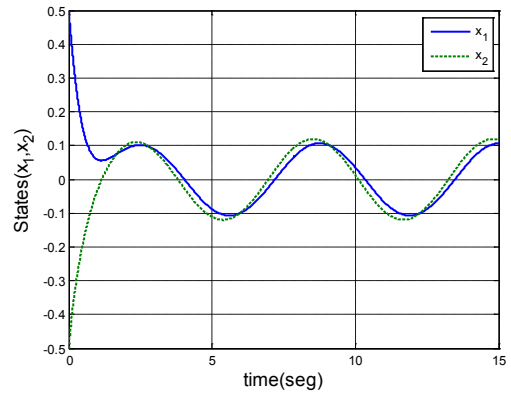


Fig. 5. Time evolution of the states in Example 2.

The initial condition are $[0.5 \ -0.5]^T$, while the estimated ones are $[0 \ 0]^T$. The time evolution of the states is shown in Fig. 5. It is clear to observe that the estimation error goes to zero despite the fact that the states remain oscillating.

5. CONCLUSIONS

Novel approaches for observer design for continuous-time nonlinear models have been reported. Taking advantage of a convex rewriting of the model (TS form) as well as the Finsler's Lemma or a Tustin-like transformation, the observer design has been decoupled from the quadratic Lyapunov function it is based on. It has been shown that the proposed decoupling introduces progressively better results thanks to a nested convex structure. Some examples have been given to illustrate the usefulness of the new schemes.

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REFERENCES

- Bergsten, P., Palm, R., and Driankov, D. (2002). "Observers for Takagi-Sugeno fuzzy systems." *IEEE Transactions on Systems, Man, and Cybernetics, Part B*, 32(1), 114–121.
- Bernal, M., and Guerra, T. M. (2010). "Generalized non-quadratic stability of continuous-time Takagi-Sugeno models." *IEEE Transactions on Fuzzy Systems*, 18(4), 815–822.
- Blanco, Y., Perruquetti, W., and Borne, P. (2001). "Stability and stabilization of nonlinear systems and Tanaka-Sugeno fuzzy models." *European Control Conference*.

- Boyd, S., El Ghaoul, L., Feron, E., and Balakrishnan, V. (1994). *Linear matrix inequalities in system and control theory*. Society for Industrial Mathematics.
- Feng, G., Chen, C. L., Soun, D., and Zhu, Y. (2005). “ H_∞ controller synthesis of fuzzy dynamic systems based on piecewise Lyapunov functions and bilinear matrix inequalities.” *IEEE Transactions on Fuzzy Systems*, 13(1), 94–103.
- Guelton, K., Bouarar, T., and Manamanni, N. (2009). “Robust dynamic output feedback fuzzy Lyapunov stabilization of Takagi-Sugeno systems: A descriptor redundancy approach.” *Fuzzy Sets and Systems*, 160(19), 2796–2811.
- Guerra, T. M., Kruszewski, A., and Bernal, M. (2009). “Control law proposition for the stabilization of discrete Takagi-Sugeno models.” *IEEE Transactions on Fuzzy Systems*, 17(3), 724–731.
- Guerra, T. M., and Vermeiren, L. (2004). “LMI-based relaxed non-quadratic stabilization conditions for nonlinear systems in the Takagi-Sugeno’s form.” *Automatica*, 40(5), 823–829.
- Jaadari, A., Guerra, T. M., Sala, A., Bernal, M., and Guelton, K. (2012). “New controllers and new designs for continuous-time Takagi-Sugeno models.” *IEEE International Conference on Fuzzy Systems*, 1–7.
- Johansson, M., Rantzer, A., and Arzen, K. E. (1999). “Piecewise quadratic stability of fuzzy systems.” *IEEE Transactions on Fuzzy Systems*, 7(6), 713–722.
- Lee, D. H., Park, J. B., and Joo, Y. H. (2012). “A fuzzy Lyapunov function approach to estimating the domain of attraction for continuous-time Takagi-Sugeno fuzzy systems.” *Information Sciences*, 185(1), 230–248.
- Liu, X., and Zhang, Q. (2003). “Approaches to quadratic stability conditions and H_∞ control designs for TS fuzzy systems.” *IEEE Transactions on Fuzzy Systems*, 11(6), 830–839.
- Márquez, R., Guerra, T. M., Kruszewski, A., and Bernal, M. (2013). “Improvements on Non-PDC Controller Design for Takagi-Sugeno Models.” *IEEE International Conference on Fuzzy Systems*, 1–7.
- Mozelli, L. A., Palhares, R. M., and Avellar, G. S. (2009). “A systematic approach to improve multiple Lyapunov function stability and stabilization conditions for fuzzy systems.” *Information Sciences*, 179(8), 1149–1162.
- De Oliveira, M., and Skelton, R. (2001). *Perspectives in robust control*. Chapter 15, Vol. 268, Springer.
- Oliveira, R. C. L. F., de Oliveira, M. C., and Peres, P. L. D. (2011). “Robust state feedback LMI methods for continuous-time linear systems: Discussions, extensions and numerical comparisons.” *IEEE International Symposium on Computer-Aided Control System Design*, 1038–1043.
- Rhee, B. J., and Won, S. (2006). “A new fuzzy Lyapunov function approach for a Takagi-Sugeno fuzzy control system design.” *Fuzzy Sets and Systems*, 157(9), 1211–1228.
- Sala, A., and Ariño, C. (2007). “Asymptotically necessary and sufficient conditions for stability and performance in fuzzy control: Applications of Polyá’s theorem.” *Fuzzy Sets and Systems*, 158(24), 2671–2686.
- Sala, A., Guerra, T. M., and Babuška, R. (2005). “Perspectives of fuzzy systems and control.” *Fuzzy Sets and Systems*, 156(3), 432–444.
- Shaked, U. (2001). “Improved LMI representations for the analysis and the design of continuous-time systems with polytopic type uncertainty.” *Automatic Control, IEEE Transactions on*, 46(4), 652–656.
- Takagi, T., and Sugeno, M. (1985). “Fuzzy identification of systems and its applications to modeling and control.” *IEEE Transactions on Systems, Man and Cybernetics*, 15(1), 116–132.
- Tanaka, K., Hori, T., and Wang, H. O. (2003). “A multiple Lyapunov function approach to stabilization of fuzzy control systems.” *IEEE Transactions on Fuzzy Systems*, 11(4), 582–589.
- Tanaka, K., Ikeda, T., and Wang, H. O. (1998). “Fuzzy regulators and fuzzy observers: relaxed stability conditions and LMI-based designs.” *IEEE Transactions on Fuzzy Systems*, 6(2), 250–265.
- Tanaka, K., Ohtake, H., and Wang, H. O. (2009). “Guaranteed cost control of polynomial fuzzy systems via a sum of squares approach.” *IEEE Transactions on Systems, Man, and Cybernetics*, 39(2), 561–567.
- Tanaka, K., and Wang, H. O. (2001). *Fuzzy control systems design and analysis: a linear matrix inequality approach*. Wiley-Interscience.
- Taniguchi, T., Tanaka, K., Ohtake, H., and Wang, H. O. (2001). “Model construction, rule reduction, and robust compensation for generalized form of Takagi-Sugeno fuzzy systems.” *IEEE Transactions on Fuzzy Systems*, 9(4), 525–538.
- Tuan, H. D., Apkarian, P., Narikiyo, T., and Yamamoto, Y. (2001). “Parameterized linear matrix inequality techniques in fuzzy control system design.” *IEEE Transactions on Fuzzy Systems*, 9(2), 324–332.
- Wang, H. O., Tanaka, K., and Griffin, M. F. (1996). “An approach to fuzzy control of nonlinear systems: Stability and design issues.” *IEEE Transactions on Fuzzy Systems*, 4(1), 14–23.