ROBUST CONTROL FOR BILINEAR TIME-DELAY STOCHASTIC JUMPING SYSTEMS

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Abstract: In this paper, we investigate the stochastic stabilization problem for a class of bilinear continuous time-delay uncertain systems with Markovian jumping parameters. Specifically, the stochastic bilinear jump system under study involves unknown state time-delay, parameter uncertainties, and unknown nonlinear deterministic disturbances. The jumping parameters considered here form a continuous-time discrete-state homogeneous Markov process. The whole system may be regarded as a stochastic bilinear hybrid system which includes both time-evolving and event-driven mechanisms. Our attention is focused on the design of a robust state-feedback controller such that, for all admissible uncertainties as well as nonlinear disturbances, the closed-loop system is stochastically exponentially stable in the mean square, independent of the time delay. Sufficient conditions are established to guarantee the existence of desired robust controllers, which are given in terms of the solutions to a set of either linear matrix inequalities (LMIs), or coupled quadratic matrix inequalities. Copyright © $2002\ IFAC$

Keywords: Bilinear systems, Linear Matrix Inequalities, Markovian jump, Stochastic exponential stability, Time-delay, Uncertainty.

1. INTRODUCTION

A lot of dynamical systems have variable structures subject to random abrupt changes, which may result from abrupt phenomena such as random failures and repairs of the components, changes in the interconnections of subsystems, sudden environment changes, modification of the operating point of a linearized model of a nonlinear systems, etc. The hybrid systems, which involve both time-evolving and event-driven mechanisms, may be employed to model the above problems. A special class of hybrid systems are the so-called Jump Linear Systems (JLSs). The jump linear system has many operation modes,

appeared on this general topic, see (Niculescu

It has been recognized that the time-delays and

and the system mode switching is governed by a Markov process. The parameter jumps among different modes may be seen as discrete events. The control of JLSs has been a research subject and attracted a lot of interest since the mid 1960's. The optimal regulator, controllability, observability, stability and stabilization problems have been extensively studied for JLSs, see (Shi et al., 1999) and references therein.

parameter uncertainties, which are inherent features of many physical processes, are very often the cause for instability and poor performance of systems. In the past few years, considerable attention has been given to the robust and/or H_{∞} controller design problems for linear uncertain state delayed systems. A great many of papers have

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et al., 1998) for a survey. As for the JLSs with parametric uncertainties, the issues of stability, stabilization, H_2 control, H_{∞} control, H_2/H_{∞} control, Kalman filtering have been well investigated, and recent results can be found in (de Farias et al., 2000; Shi et al., 1999) and references therein. In (Mao et al., 2000), the exponential stability analysis problem for a general class of linear/nonlinear stochastic jumping delay systems has been intensively studied, and a number of useful stability criteria have been established. In particular, for the linear case in (Mao et al., 2000), the exponential stability can be easily tested by checking the existence of the solution to a linear matrix inequality. Unfortunately, the parametric uncertainties and the nonlinear exogenous disturbance have not been considered in (Mao et al., 2000) for stabilization problem.

On the other hand, bilinear systems have been of great interest in the past three decades, since many real-world systems can be adequately approximated by a bilinear model. The application areas include nuclear, thermal, chemical processes, biology, socio-economics, immunology, etc., see (Mohler and Kolodziej, 1980). In particular, the stochastic bilinear systems, also called statedependent noise systems or multiplicative noise systems, have been dealt with by many authors. Among them we quote (Bernstein and Haddad, 1987; Skelton et al., 1991; Yaz, 1992; Wang and Burnham, 2001). However, a literature search reveals that the issue of stabilization of jump bilinear systems with or without uncertainty and timedelay has not been fully investigated and remains important and challenging. This situation motivates the present study on the robust stabilization of bilinear continuous time-delay jump systems.

It is now worth pointing out that the essential differences between the Jump Linear System (JLS) which has been extensively studied as mentioned above and the Jump Bilinear Stochastic System (JBSS) which is to be considered in this paper. For JLS, every mode corresponds to a deterministic dynamics, that is, when the mode is fixed, the system state evolves according to the corresponding deterministic dynamics. However, the JBSS can be regarded as the result of several stochastic systems (systems with multiplicative noises) switching from one to the others according to the movement of a Markov chain. For JBSS, every mode corresponds to a stochastic dynamics. Obviously, the JLS is a special case of the JBSS.

This paper is concerned with the stochastic stabilization problem for a class of bilinear continuous time-delay uncertain systems with Markovian jumping parameters. We aim at designing a robust state-feedback controller such that, for all admissible uncertainties as well as nonlinear disturbances,

the closed-loop system is stochastically exponentially stable in the mean square, independent of the time delay. We show that the analysis problem can be tackled in terms of the solutions to a set of linear matrix inequalities (LMIs), see (Gahinet et al., 1995), and the associated synthesis problem can be dealt with by solving a set of coupled quadratic matrix inequalities. Due to space limitation, we omit the numerical simulation example.

Notation. The notations in this paper are quite standard. \mathbb{R}^n and $\mathbb{R}^{n \times m}$ denote, respectively, the n dimensional Euclidean space and the set of all $n \times m$ real matrices. The superscript "T" denotes the transpose and the notation $X \geq Y$ (respectively, X > Y) where X and Y are symmetric matrices, means that X - Y is positive semi-definite (respectively, positive definite). I is the identity matrix with compatible dimension. We let h > 0and $C([-h,0];\mathbb{R}^n)$ denote the family of continuous functions φ from [-h,0] to \mathbb{R}^n with the norm
$$\begin{split} \|\varphi\| &= \sup_{-h \leq \theta \leq 0} |\varphi(\theta)|, \, \text{where} \, |\cdot| \, \text{is the Euclidean} \\ \text{norm in } \mathbb{R}^n. \, \text{If } A \, \text{ is a matrix, denote by } \|A\| \end{split}$$
its operator norm, i.e., $||A|| = \sup\{|Ax| : |x| =$ 1\} = $\sqrt{\lambda_{\max}(A^T A)}$ where $\lambda_{\max}(\cdot)$ (respectively, $\lambda_{\min}(\cdot)$) means the largest (respectively, smallest) eigenvalue of A. $l_2[0,\infty]$ is the space of square integrable vector. Moreover, let $(\Omega, \mathcal{F}, \{\mathcal{F}_t\}_{t>0}, P)$ be a complete probability space with a filtration $\{\mathcal{F}_t\}_{t>0}$ satisfying the usual conditions (i.e., the filtration contains all P-null sets and is right continuous). Denote by $L^p_{\mathcal{F}_0}([-h,0];\mathbb{R}^n)$ the family of all \mathcal{F}_0 -measurable $C([-h,0];\mathbb{R}^n)$ -valued random variables $\xi = \{\xi(\theta) : -h \leq \theta \leq 0\}$ such that $\sup_{-h<\theta<0} \mathbb{E}|\xi(\theta)|^p < \infty$ where $\mathbb{E}\{\cdot\}$ stands for the mathematical expectation operator with respect to the given probability measure P. Sometimes, the arguments of a function will be omitted in the analysis when no confusion can arise.

2. PROBLEM FORMULATION AND ASSUMPTIONS

Let $\{r(t), t \geq 0\}$ be a right-continuous Markov process on the probability space which takes values in the finite space $S = \{1, 2, \dots, N\}$ with generator $\Pi = (\gamma_{ij})$ $(i, j \in S)$ given by

$$P\{r(t + \Delta) = j | r(t) = i\}$$

$$= \begin{cases} \gamma_{ij}\Delta + o(\Delta) & \text{if } i \neq j \\ 1 + \gamma_{ii}\Delta + o(\Delta) & \text{if } i = j \end{cases}$$

where $\Delta > 0$ and $\lim_{\Delta \to 0} o(\Delta)/\Delta = 0$, $\gamma_{ij} \geq 0$ is the transition rate from i to j if $i \neq j$ and $\gamma_{ii} = -\sum_{j \neq i} \gamma_{ij}$.

In this paper, we consider the following class of bilinear uncertain continuous-time state delayed stochastic systems of the Itô type

$$dx(t) = [A(r(t)) + \Delta A(t, r(t))] x(t) dt$$

$$+ \sum_{k=1}^{m} J_k(r(t)) x(t) dw_k(t)$$

$$+ [A_d(r(t)) x(t-h) + B(r(t)) u(t)$$

$$+ D(r(t)) f(x(t, r(t)))] dt,$$
 (1)

(2)

 $x(t) = \varphi(t), \quad r(t) = r(0), \ t \in [-h, \ 0]$

where $x(t) \in \mathbb{R}^n$ is the state, $u(t) \in \mathbb{R}^m$ is the control input, $f(\cdot) : \mathbb{R}^n \to \mathbb{R}^{n_f}$ is an unknown nonlinear exogenous disturbance input, h denotes the unknown state delay, $\varphi(t)$ is a continuous vector valued initial function. Here, $w(t) := [w_1(t) \ w_2(t) \cdots \ w_m(t)]^T \in \mathbb{R}^m$ is an m-dimensional Brownian motion, and it is assumed that the Markov process $r(\cdot)$ is independent of $w_k(\cdot)$ $(k = 1, 2, \cdots, m)$. For fixed system mode, A(r(t)), $J_k(r(t))$ $(k = 1, 2, \cdots, n)$, $A_d(r(t))$, B(r(t)), D(r(t)) are known constant matrices with appropriate dimensions. $\Delta A(t, r(t))$ is a real-valued matrix function representing norm-

Assumption 1. The uncertain matrix $\Delta A(t, r(t))$ satisfies

bounded parameter uncertainty.

$$\Delta A(t, r(t)) = M(r(t))F(t, r(t))N(r(t)) \tag{3}$$

where for fixed system mode, $M(r(t)) \in \mathbb{R}^{n \times i}$ and $N(r(t)) \in \mathbb{R}^{j \times n}$ are known real constant matrices which characterize how the deterministic uncertain parameter in F(t, r(t)) enters the nominal matrix A(r(t)); and $F(t, r(t)) \in \mathbb{R}^{i \times j}$ is an unknown time-varying matrix function meeting

$$F^{T}(t, r(t))F(t, r(t)) \le I, \ \forall \ t \ge 0; r(t) = i \in \mathcal{S}. \tag{4}$$

Assumption 2. For fixed system mode, there exists a known real constant matrix $H(r(t)) \in \mathbb{R}^{n \times n}$ such that the unknown nonlinear vector function $f(\cdot)$ satisfies the following boundedness condition

$$|f(x(t,r(t)))| \le |H(r(t))x(t)|, \quad \forall \ x(t,r(t)) \in \mathbb{R}^n.$$
(5)

Assumption 3. For all $\delta \in [-h, 0]$, there exists a scalar $\sigma > 0$ such that $|x(t + \delta)| \leq \sigma |x(t)|$.

Remark 1. It is noted that, in the system model (1)-(2), there are two kinds of uncertainties acting on the nominal matrix A(r(t)), that is, the deterministic uncertainty $\Delta A(t,r(t))$ which can be regarded as the energy-bounded noise, and the stochastic perturbation $\sum_{k=1}^{m} J_k(r(t))dw_k(t)$ which is the multiplicative noise with known statistics. Both kinds of uncertainties have been extensively studied in the literature. If the multiplicative noise disappears and there are no time-delay and nonlinear exogenous disturbance, the system model (1)-(2) will be reduced to the usual

jump linear system which has received considerable attention. Note that when the mode is fixed, the system (1)-(2) corresponds to a bilinear stochastic time-delay uncertain system.

Remark 2. The parameter uncertainty structure as in (3)-(4) has been widely used in the problems of robust control and robust filtering of uncertain systems, see, e.g., (Wang and Huang, 2000). We point out that the exogenous nonlinear timevarying disturbance term f(x(t,r(t))) in the system model (1)-(2) has not been taken into account in the research literature concerning jump systems. Such kind of disturbance may result from the linearization process of an originally highly nonlinear plant or may be an actual external nonlinear input. Also, as mentioned in (Cao amd Lam, 2000), Assumption 3 is not restrictive as the scalar $\sigma > 0$ can be chosen arbitrarily.

Observe the system (1)-(2) and let $x(t;\xi)$ denote the state trajectory from the initial data $x(\theta) = \xi(\theta)$ on $-h \leq \theta \leq 0$ in $L^2_{\mathcal{F}_0}([-h,0];\mathbb{R}^n)$. Clearly, the system (1)-(2) admits a trivial solution $x(t;0) \equiv 0$ corresponding to the initial data $\xi = 0$.

We now introduce the following stability concepts.

Definition 1. For the uncertain time-delay bilinear jump system (1)-(2) with $u(t) \equiv 0$ and every $\xi \in L^2_{\mathcal{F}_0}([-h,0];\mathbb{R}^n)$, the trivial solution is exponentially stable in the mean square if there exist scalars $\alpha > 0$ and $\beta > 0$ such that

$$\mathbb{E}|x(t;\xi)|^2 \le \alpha e^{-\beta t} \sup_{-h \le \theta \le 0} \mathbb{E}|\xi(\theta)|^2.$$
 (6)

Definition 2. We say that the system (1)-(2) is exponentially stabilizable in the mean square if, for every $\xi \in L^2_{\mathcal{F}_0}([-h,0];\mathbb{R}^n)$, there exists a linear state feedback control law u(t) = G(r(t))x(t) (the feedback gain G(r(t)) is constant for each fixed mode) such that the closed-loop system is exponentially stable in mean square.

In this paper, we assume that all jump states $r(t) = i \in \mathcal{S}$ $(t \geq 0)$ and the system states x(t) $(t \geq 0)$ are accessible, i.e., they are measurable for feedback.

The purpose of this paper is to design a delay-independent memoryless state feedback controller of the form

$$G(r(t)): \quad u(t) = G(r(t))x(t) \tag{7}$$

based on the state x(t) and the system mode r(t), such that the following closed-loop system of (1)-(2) with $\mathcal{G}(r(t))$

$$dx(t) = \left[A(r(t)) + B(r(t))G(r(t)) + \Delta A(t, r(t)) \right] x(t) dt$$

$$+ \sum_{k=1}^{m} J_k(r(t))x(t) dw_k(t)$$

$$+ \left[A_d(r(t))x(t-h) + D(r(t))f(x(t, r(t))) \right] dt \qquad (8)$$

is exponentially stable in the mean square.

3. MAIN RESULTS AND PROOFS

Lemma 1. Let M, N and F be real matrices of appropriate dimensions with $F^TF \leq I$. Then for any scalar $\mu \neq 0$, we have

$$MFN + N^T F^T M^T \le \mu^2 M M^T + \mu^{-2} N^T N.$$

Recall that the Markov process $\{r(t), t \geq 0\}$ takes values in the finite space $S = \{1, 2, \dots, N\}$. For the sake of simplicity, we write

$$\begin{split} A(i) &:= A_i, \quad A_d(i) := A_{di}, \quad B(i) := B_i, \\ J_k(i) &:= J_{ki}, \quad D(i) := D_i, \quad M(i) := M_i, \\ N(i) &:= N_i, \quad H(i) := H_i, \quad G(i) := G_i, \quad \forall \ i \in \mathcal{S}. \end{split}$$

and

$$A_{ci} := A(i) + B(i)G(i) = A_i + B_iG_i,$$
 (9)

and then for the mode r(t) = i, the closed-loop system (8) becomes

$$dx(t) = [A_{ci} + M_i F(t, i) N_i] x(t) dt$$

$$+ \sum_{k=1}^{m} J_{ki} x(t) dw_k(t)$$

$$+ [A_{di} x(t - h) + D_i f(x(t, i))] dt. (10)$$

In the following theorem, we establish the analysis results, i.e., for a given controller, we derive the sufficient conditions under which the closed-loop system (10) is exponentially stable in the mean square.

Theorem 1. Let the controller gain G(r(t)) be given. If there exist a positive scalar $\mu > 0$ such that the following N matrix inequalities

$$A_{ci}^{T}P_{i} + P_{i}A_{ci} + \sum_{k=1}^{m} J_{ki}^{T}P_{i}J_{ki} + \sum_{j=1}^{N} \gamma_{ij}P_{j}$$
$$+P_{i}(A_{di}A_{di}^{T} + D_{i}D_{i}^{T} + \mu^{2}M_{i}M_{i}^{T})P_{i}$$
$$+\mu^{-2}N_{i}^{T}N_{i} + H_{i}^{T}H_{i} + I < 0 \tag{11}$$

have positive definite solutions $P_i > 0$ $(i \in \mathcal{S})$, then the system (10) is exponentially stable in the mean square.

Proof. First, we let $C^{2,1}(\mathbb{R}^n \times \mathbb{R}_+ \times \mathcal{S}; \mathbb{R}_+)$ denote the family of all nonnegative functions Y(x,t,i)

on $\mathbb{R}^n \times \mathbb{R}_+ \times \mathcal{S}$ which are continuously twice differentiable in x and once differentiable in t.

Fix $\xi \in L^2_{\mathcal{F}_0}([-h,0];\mathbb{R}^n)$ arbitrarily and write $x(t;\xi) = x(t)$. Define a Lyapunov function candidate $Y(x,t,i) \in C^{2,1}(\mathbb{R}^n \times \mathbb{R}_+ \times \mathcal{S};\mathbb{R}_+)$ by

$$Y(x(t), t, i) = x^{T}(t)P_{i}x(t) + \int_{t-h}^{t} x^{T}(s)x(s)ds.$$
(12)

By Itô's formula (see, e.g., Mao 1997), the stochastic derivative of Y along a given trajectory is obtained as

$$dY(x(t), t, i) := \mathcal{L}Y(x(t), t, i)dt + 2x^{T}(t)P_{i}$$

$$\cdot \left(\sum_{k=1}^{m} J_{ki}x(t)dw_{k}(t)\right), \quad (13)$$

where

$$\mathcal{L}Y(x(t), t, i) = x^{T}(t) \left[(A_{ci} + \Delta A(t, i))^{T} P_{i} + P_{i} (A_{ci} + \Delta A(t, i)) + \sum_{k=1}^{m} J_{ki}^{T} P_{i} J_{ki} + \sum_{j=1}^{N} \gamma_{ij} P_{j} + I \right] x(t) + x^{T}(t - h) A_{di}^{T} P_{i} x(t) + x^{T}(t) P_{i} A_{di} x(t - h) + x^{T}(t) P_{i} D_{i} f(x(t, i)) + f^{T}(x(t, i)) D_{i}^{T} P_{i} x(t) - x^{T}(t - h) x(t - h).$$

$$(14)$$

Note that $\Delta A(t,i) = M_i F(t,i) N_i$ and $F^T F \leq I$. Lemma 1 shows that, for any scalar $\mu > 0$,

$$P_{i}(\Delta A(t,i)) + (\Delta A(t,i))^{T} P_{i}$$

$$= (P_{i}M_{i})F(t,i)N_{i} + N_{i}^{T} F^{T}(t,i)(P_{i}M_{i})^{T}$$

$$\leq \mu^{2} P_{i}M_{i}M_{i}^{T} P_{i} + \mu^{-2}N_{i}^{T} N_{i}.$$
(15)

Moreover, it results from (5) and the inequality

$$(f^T(x(t,i)) - x^T(t)P_iD_i)$$
$$\cdot (f^T(x(t,i)) - x^T(t)P_iD_i)^T \ge 0$$

that

$$x^{T}(t)P_{i}D_{i}f(x(t,i)) + f^{T}(x(t,i))D_{i}^{T}P_{i}x(t)$$

$$\leq f^{T}(x(t,i))f(x(t,i)) + x^{T}(t)P_{i}D_{i}D_{i}^{T}P_{i}x(t)$$

$$\leq x^{T}(t)(H_{i}^{T}H_{i} + P_{i}D_{i}D_{i}^{T}P_{i})x(t).$$
(16)

Denote

$$\Theta_{i} := A_{ci}^{T} P_{i} + P_{i} A_{ci} + \mu^{2} P_{i} M_{i} M_{i}^{T} P_{i}$$

$$+ \mu^{-2} N_{i}^{T} N_{i} + \sum_{k=1}^{m} J_{ki}^{T} P_{i} J_{ki} + \sum_{j=1}^{N} \gamma_{ij} P_{j}$$

$$+ H_{i}^{T} H_{i} + P_{i} D_{i} D_{i}^{T} P_{i} + I,$$

$$(17)$$

$$S_i := \begin{bmatrix} \Theta_i & P_i A_{di} \\ A_{di}^T P_i & -I \end{bmatrix}. \tag{18}$$

Then, substituting (15)(16) into (14) results in

$$\mathcal{L}Y(x(t),t,i) \leq x^{T}(t)\Theta x(t) + x^{T}(t-h)A_{di}^{T}P_{i}x(t) + x^{T}(t)P_{i}A_{di}x(t-h) - x^{T}(t-h)x(t-h) = \left[x^{T}(t) \ x^{T}(t-h)\right]S_{i} \cdot \begin{bmatrix}x(t) \\ x(t-h)\end{bmatrix}$$

$$(19)$$

From the Schur Complement Lemma, we know that $S_i < 0$ if and only if

$$\Theta_i + P_i A_{di} A_{di}^T P_i < 0 \tag{20}$$

which is the same as the inequality (11). Therefore, we have $\mathcal{L}Y(x(t),t,i) < 0$.

Note that $|x(t)| \leq |x_e(t)|$, $S_i < 0$, and $P_i > 0$. Then, based on $\mathcal{L}Y(x(t),t,i) < 0$, following Assumption 3 and the line of the proof of Theorem 1 in (Cao and Lam, 2000), we can prove that, the uncertain time-delay bilinear jump system (10) is asymptotically stable in the mean square.

With the inequality (19), the exponential stability (in the mean square) of the system (10) can be proved as follows by using the techniques developed in (Mao, 1997; Mao et al., 2000).

Define

$$\lambda_P = \max_{i \in \mathcal{S}} \lambda_{max}(P_i), \ \lambda_S = \min_{i \in \mathcal{S}} (-\lambda_{max}(S_i),$$
$$\lambda_p = \min_{i \in \mathcal{S}} \lambda_{min}(P_i),$$

where $P_i > 0$ is the solution to (11) and S_i is defined in (18). Let δ be the unique root to the equation

$$\delta(\lambda_P + he^{\delta h}) = \lambda_S + \min(1, \lambda_S e^{\delta h}).$$

To prove the mean square exponential stability, we modify the Lyapunov function candidate (12) as

$$Y_1(x(t), t, i) = e^{\delta t} \left(x^T(t) P_i x(t) + \int_{t-h}^t |x(s)|^2 ds \right). \tag{21}$$

Along the similar line for the proof of Theorem 3.1 in (Mao *et al.*, 2000), we can show that

$$e^{\delta t} \lambda_p \mathbb{E}|x(t)|^2 \le (\lambda_P + h(1 + e^{\delta h})) \mathbb{E}||\xi||^2,$$

or

$$\lim_{t \to \infty} \sup \frac{1}{t} \log(\mathbb{E}|x(t,\xi)|^2) \le -\delta.$$

This indicates that the trivial solution of the system (10) is exponentially stable in the mean square. This completes the proof of this theorem.

Remark 3. Theorem 1 provides the analysis results for the exponential stability in mean square of the system (10). It can be seen from (11) that we need to check whether there exist a positive scalar μ and N positive definite matrices $P_i > 0$ $(i = 1, 2 \cdots, N)$ meeting the N coupled matrix inequalities. This may be done by converting the N coupled nonlinear (on P_i and μ) inequalities into the associated Linear Matrix Inequalities (LMIs), and then we are able to determine exponential stability of the system (10) readily by checking the solvability of the LMIs (Gahinet et al., 1995).

Theorem 2. Let the controller gain G(r(t)) be given. If there exist a positive scalar $\varepsilon > 0$ and N positive definite matrices $P_i > 0$ $(i \in \mathcal{S})$ satisfying the following LMIs

$$\begin{bmatrix} \Lambda_{i} & P_{i}A_{di} & P_{i}D_{i} & \varepsilon N_{i}^{T} & P_{i}M_{i} \\ A_{di}^{T}P_{i} & -I & 0 & 0 & 0 \\ D_{i}^{T}P_{i} & 0 & -I & 0 & 0 \\ \varepsilon N_{i} & 0 & 0 & -\varepsilon I & 0 \\ M_{i}^{T}P_{i} & 0 & 0 & 0 & -\varepsilon I \end{bmatrix} < 0 \quad (22)$$

where Λ_i is defined by

$$\begin{split} \Lambda_i := A_{ci}^T P_i + P_i A_{ci} + \sum_{k=1}^m J_{ki}^T P_i J_{ki} \\ + \sum_{j=1}^N \gamma_{ij} P_j + H_i^T H_i + I, \end{split}$$

then the system (10) is exponentially stable in the mean square.

Proof. To begin with, we let

$$\Lambda_{1i} := \left[P_i A_{di} \ P_i D_i \ \mu^{-1} N_i^T \ \mu P_i M_i \right]$$

and then rewrite (11) as

$$\Lambda_i + \Lambda_{1i} \Lambda_{1i}^T < 0. (23)$$

If follows from the Schur Complement Lemma that, the above inequality holds if and only if

$$\begin{bmatrix} \Lambda_{i} & P_{i}A_{di} & P_{i}D_{i} & \mu^{-1}N_{i}^{T} & \mu P_{i}M_{i} \\ A_{di}^{T}P_{i} & -I & 0 & 0 & 0 \\ D_{i}^{T}P_{i} & 0 & -I & 0 & 0 \\ \mu^{-1}N_{i} & 0 & 0 & -I & 0 \\ \mu M_{i}^{T}P_{i} & 0 & 0 & 0 & -I \end{bmatrix} < 0.$$

$$(24)$$

Note that (24) is not linear in μ . Let $\varepsilon := \mu^{-2}$. Pre- and post-multiplying the inequality (24) by diag $\{I, I, I, \mu^{-1}I, \mu^{-1}I\}$ yield (22). The proof follows from Theorem 1 immediately.

Remark 4. It is observed that the inequality (22) is linear in ε and $P_i > 0$ $(i = 1, 2 \cdots, N)$, and thus the standard LMI techniques can be exploited to check the exponential stability of the closed-loop system (10) when the controller is given. The

analysis result given in Theorem 3 is also useful to determine the exponential stability of the free system (1)-(2) (i.e., $u(t) \equiv 0$).

Finally, the following result solves the addressed stochastic stabilization problem of bilinear continuous time-delay jump uncertain systems in terms of quadratic matrix inequalities.

Theorem 3. Consider the system (1)-(2) satisfying the Assumption 1 and Assumption 2. Let $\rho > 0$ be a positive scalar. If there exist a scalar $\varepsilon > 0$ and N positive definite matrices $P_i > 0$ $(i \in \mathcal{S})$ such that the following quadratic matrix inequalities

$$A_{i}^{T}P_{i} + P_{i}A_{i} + \sum_{k=1}^{m} J_{ki}^{T}P_{i}J_{ki} + \sum_{j=1}^{N} \gamma_{ij}P_{j}$$
$$+P_{i}(-2\rho B_{i}B_{i}^{T} + A_{di}A_{di}^{T} + D_{i}D_{i}^{T} + \mu^{2}M_{i}M_{i}^{T})P_{i}$$
$$+\mu^{-2}N_{i}^{T}N_{i} + H_{i}^{T}H_{i} + I < 0$$
(25)

hold, then the uncertain time-delay jump system (1)-(2) with nonlinear disturbances can be exponentially stabilized (in the mean square) by the memoryless state feedback controller of the form (7) with the gain matrix

$$G_i = -\rho B_i^T P_i \tag{26}$$

for all admissible parameter uncertainty.

Proof. The proof follows from Theorem 1 immediately by substituting (26) into (11).

Remark 5. If the matrices $A_d(r(t))$ and B(r(t)) in system (1)-(2) contain parameter uncertainties, say $\Delta A_d(r(t))$ and $\Delta B(r(t))$, similar results to Theorem 3 can also be obtained.

4. CONCLUSIONS

This paper has introduced an algebraic matrix inequality approach to the robust stabilization for a class of bilinear continuous time-delay uncertain systems with Markovian jumping parameters. We have focused on the design of a robust state-feedback controller such that, for all admissible uncertainties as well as nonlinear disturbances, the closed-loop system is stochastically exponentially stable in the mean square, independent of the time delay. Sufficient conditions have been derived to ensure the existence of desired robust controllers, which are given in terms of the solutions to a set of either linear matrix inequalities (LMIs), or coupled quadratic matrix inequalities.

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