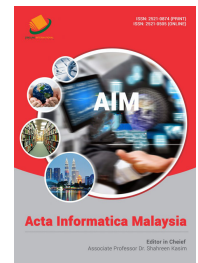




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EXPONENTIAL STABILITY CONTROL FOR NETWORKED SYSTEMS WITH INTERVAL DISTRIBUTION TIME DELAYS

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ABSTRACT

To study the problem of exponential stability control for a class of networked control systems with interval distribution time delays, a new approach is given to model the networked control systems with the stochastic time delays which is assumed to be satisfying a interval Bernoulli distribution. Based on linear matrix inequality approach, the mean-square exponential stability controller design method is presented, and the controller gain matrix is obtain by solving a linear matrix inequality. Moreover, a Lyapunov functional is used, and some stack matrices, which bring much flexibility in solving LMI, are introduced during the proof. A numerical example is given to demonstrate the validity of the results.

KEYWORDS

Networked control systems, Stochastic delays, Linear matrix inequality (LMI)

1. INTRODUCTION

Networked control systems (NCSs) are systems where the feedback loop is closed via a communication network in which information, from various components such as sensors, controllers and actuators, is exchanged with limited bandwidth. NCSs have received increasing attentions in recent years due to their low cost simple installation and maintenance and high reliability [1,2].

However, the network itself is dynamic system that exhibits characteristics such as network-induced delays. The delays come from the time sharing of the communication medium as well as the computation time required for physical signal coding and communication processing. As is known, network-induced delays can degrade a system's performance and even cause system instability. Many researchers have studied stability analysis and controller design for NCSs [3,4]. It is quite common in practice that the time delays occur in a random way, rather than a deterministic way. Based on a similar Bernoulli stochastic model, study NCSs with both sensor-to-controller and controller-to-actuator stochastic delays and design the controller gain [5-6]. One step stochastic delays or one stochastic packet dropout is considered in these papers.

The stabilization problem for a networked control system with Markov communication delays existing in both the system state and the mode signal [7]. The problem of the stabilization of NCSs with packet dropout [8]. A networked predictive control method for networked systems with stochastic delay and data dropout is proposed in to compensate the networked-induced delay [9-10]. This paper considers the problem of mean-square exponential stability control for a class of networked control systems with interval distribution time delay. A new approach is given to model the networked control systems. Based on the LIM approach and Lyapunov stability theorem, the mean-square exponential stability condition is given.

2. PROBLEM FORMULATION

Consider the following control system with delay

$$\begin{aligned} \dot{x}(t) &= Ax(t) + A_d x(t-d) + Bu(t) \\ x(t) &= \phi(t) \end{aligned} \quad t \in [-d, 0] \quad (1)$$

where $x(t) \in R_n$ is the state vector, $u(t) \in R_m$ is the input vector, d is state delay $A, A_d \in R^{n \times n}$ are known real constant matrices, $B \in R^{n \times m}$ is input matrix, $\phi(t) \in R^n$ is given initial state on $[-d, 0]$.

Throughout this note, we suppose that all the system's states are available for a state feedback control. In the presence of the control network, data transfers between the controller and the remote system, e.g., sensors and actuators in a distributed control system will induce network delay in addition to the controller proceeding delay. We introduce stochastic delay $\tau(t)$ to denote the network-induced delay. In this note we make the following assumptions:

Assumption 1: Sensor and controller are clock-driven;

Assumption 2: Actuator is event-driven.

We will design the state feedback controller

$$u(t) = Kx(t - \tau(t)) \quad (2)$$

Inserting the controller (2) into system (1), we obtain the closed system:

$$\begin{aligned} \dot{x}(t) &= Ax(t) + A_d x(t-d) + BKx(t - \tau(t)) \\ x(t) &= \psi(t) \end{aligned} \quad t \in [-\bar{d}, 0] \quad (3)$$

The initial condition of the state is supplemented as $x(t) = \psi(t)$, where $\psi(t)$ is a smooth function on $[-\bar{d}, 0]$, $\bar{d} = \max\{\tau, d\}$. Therefore, there exists a positive constant $\bar{\psi}$ satisfying

$$\|\dot{\psi}(t)\| \leq \bar{\psi} \quad t \in [-\bar{d}, 0]$$

It is assumed that there exists a constant $\tau_1 \in [0, \tau]$ such that the probability of $\tau(t)$ taking values in $[0, \tau_1)$ and $[\tau_1, \tau]$ can be observed. In order to employ the information of the probability distribution of the delays, the following sets are proposed firstly

$$\Omega_1 = \{\tau(t) \in [0, \tau_1)\}, \quad \Omega_2 = \{\tau(t) \in [\tau_1, \tau]\}$$

Obviously, $\Omega_1 \cup \Omega_2 = R^+$ and $\Omega_1 \cap \Omega_2 = \Phi$

Then we define two functions as:

$$h_1(t) = \begin{cases} \tau(t) & t \in \Omega_1 \\ 0 & t \notin \Omega_1 \end{cases}, \quad h_2(t) = \begin{cases} \tau(t) & t \in \Omega_2 \\ \tau_1 & t \notin \Omega_2 \end{cases} \quad (4)$$

Corresponding to $\tau(t)$ taking values in different intervals, a stochastic variable $\beta(t)$ is defined

$$\beta(t) = \begin{cases} 1 & t \in \Omega_1 \\ 0 & t \in \Omega_2 \end{cases} \quad (5)$$

Where we suppose that $\beta(t)$ is a Bernoulli distributed sequence with $\text{Pr}\{ob\{\beta(t)=1\}=E\{\beta(t)\}=\beta$, where $\beta \in [0,1]$ is a constant.

With (4), (5), we know that the systems (3) is equivalent to

$$\begin{aligned} \dot{x}(t) &= Ax(t) + A_d x(t-d) + \beta(t)BKx(t-h_1(t)) + (1-\beta(t))BKx(t-h_2(t)) \\ &= \bar{A}\xi(t) \\ x(t) &= \psi(t) \quad t \in [-\bar{d}, 0] \end{aligned} \quad (6)$$

where

$$\begin{aligned} \bar{A} &= [A \quad A_d \quad \beta(t)BK \quad (1-\beta(t))BK] \\ \xi^T(t) &= [x^T(t), \quad x^T(t-d), \quad x^T(t-h_1(t)), \quad x^T(t-h_2(t))] \end{aligned}$$

3. MAIN RESULTS

Lemma 1 For any vectors ,ab and matrices N,XY,Z with appropriate dimensions, if the following matrix inequality holds [2]

$$\begin{bmatrix} X & Y \\ Y^T & Z \end{bmatrix} \geq 0$$

then we have

$$-2a^T N b \leq \inf_{x,Y,Z} \begin{bmatrix} a \\ b \end{bmatrix}^T \begin{bmatrix} X & Y-N \\ Y^T-N^T & Z \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix}$$

Lemma 2 The LMI $\begin{bmatrix} Y(x) & W(x) \\ * & R(x) \end{bmatrix} > 0$ is equivalent to [4]

$$R(x) > 0, Y(x) - W(x)R^{-1}(x)W^T(x) > 0$$

where $Y(x) = Y^T(x), R(x) = R^T(x)$ depend on x .

Theorem 1 For the given constants $\alpha > 0, 1 \geq \beta \geq 0$, if there exist positive-definite matrices $P, Q, R \in R^{n \times n}$ and matrices $K \in R^{m \times n}$, X, Y with appropriate dimensions, such that the following matrix inequality holds

$$\Theta = \begin{bmatrix} \Theta_{11} & \Theta_{12} \\ * & \Theta_{22} \end{bmatrix} < 0 \quad (7)$$

where

$$\begin{aligned} \Theta_{11} &= \begin{bmatrix} PA + A^T P + Q + 2\alpha P & P\bar{A}_d + \tau_2 X_{12} + \tau_2 A^T R A_d \\ +\tau_2 X_{11} + \tau_2 A^T R A & * \\ * & -e^{-2\alpha d} Q + \tau_2 X_{22} + \tau_2 A_d^T R A_d \end{bmatrix} \\ \Theta_{12} &= \begin{bmatrix} P\beta BK + Y_1 + \tau_2 X_{13} + \tau_2 A^T R \beta BK & P(1-\beta)BK + \tau_2 X_{14} - Y_1 + \tau_2 A^T R(1-\beta)BK \\ Y_2 + \tau_2 X_{23} + \tau_2 A_d^T R \beta BK & \tau X_{24} - Y_2 + \tau_2 A_d^T R(1-\beta)BK \end{bmatrix} \text{ with the controller} \\ \Theta_{22} &= \begin{bmatrix} \tau_2 X_{33} + Y_3 + Y^T + \tau_2 K^T B^T R \beta BK & -Y_3 + Y_4^T + \tau_2 X_{34} \\ * & \tau_2 X_{44} - Y_4 - Y^T + \tau_2 K^T B^T R(1-\beta)BK \end{bmatrix} \end{aligned}$$

(2), the network control systems (6) is mean-square exponentially stable.

Proof : Choose a Lyapunov functional candidate for the system (6) as follows :

$$\begin{aligned} \dot{V}(t) + 2\alpha V(t) &= 2x^T(t)P\dot{x}(t) + x^T(t)Qx(t) - x^T(t-d)Qe^{-2\alpha d}x(t-d) \\ &\quad + \tau x^T(t)R\dot{x}(t) + 2\alpha x^T(t)Px(t) - \int_{t-\tau_2}^t \dot{x}^T(s)Re^{2\alpha(s-t)}\dot{x}(s)ds \end{aligned}$$

With

$$x(t-h_1(t)) - x(t-h_2(t)) - \int_{t-h_2(t)}^{t-h_1(t)} \dot{x}(s)ds = 0$$

For any $4n \times n$ matrix

$$N = \begin{bmatrix} N_1 \\ N_2 \\ N_3 \end{bmatrix}$$

We know

$$0 = \xi^T(t)N[x(t-h_1(t)) - x(t-h_2(t))] - \int_{t-h_2(t)}^{t-h_1(t)} \dot{x}(s)ds \quad (9)$$

With lemma1 and (9), we obtain

$$\begin{aligned} 0 &\leq 2\xi^T(t)N[x(t-h_1(t)) - x(t-h_2(t))] \\ &\quad + \int_{t-h_2(t)}^{t-h_1(t)} \left[\xi^T(s) \right]^T \begin{bmatrix} X & Y-N \\ Y^T-N^T & Re^{2\alpha(s-t)} \end{bmatrix} \begin{bmatrix} \xi(s) \\ \dot{x}(s) \end{bmatrix} ds \\ &\leq 2\xi^T(t)Y[x(t-h_1(t)) - x(t-h_2(t))] + \tau_2 \xi^T(t)X\xi(t) \\ &\quad + \int_{t-\tau_2}^t \dot{x}^T(s)Re^{2\alpha(s-t)}\dot{x}(s)ds \end{aligned} \quad (10)$$

Inserting (10) into (8), we have :

$$\begin{aligned} \dot{V}(t) + 2\alpha V(t) &\leq x^T(t)[PA + AP + Q + 2\alpha P]x(t) \\ &\quad + 2x^T(t)PA_d x(t-d) + 2x^T(t)P\beta(t)BKx(t-h_1(t)) \\ &\quad + 2x^T(t)P(1-\beta(t))BKx(t-h_2(t)) \\ &\quad - x^T(t-d)Qe^{-2\alpha d}x(t-d) + 2\xi^T(t)Y[0 \quad 0 \quad I \quad -I]\xi(t) \\ &\quad + \tau_2 \xi^T(t)X\xi(t) + \tau_2 \dot{x}^T(t)R\dot{x}(t) \end{aligned} \quad (11)$$

where

$$\begin{aligned} &\dot{x}^T(t)R\dot{x}(t) \\ &= \tau_2 \xi^T(t) \begin{bmatrix} A^T R A & A^T R A_d & A^T R \beta(t)BK & A^T R(1-\beta(t))BK \\ * & A_d^T R A_d & A_d^T R \beta(t)BK & A_d^T R(1-\beta(t))BK \\ * & * & \beta^2(t)K^T B^T R B K & \beta(t)(1-\beta(t))K^T B^T R B K \\ * & * & * & (1-\beta(t))^2 K^T B^T R B \end{bmatrix} \xi(t) \end{aligned} \quad (12)$$

Obviously

$$2\xi^T(t) \begin{bmatrix} Y_1 \\ Y_2 \\ Y_3 \\ Y_4 \end{bmatrix} [0 \quad 0 \quad I \quad -I]\xi(t) = \xi^T(t) \begin{bmatrix} 0 & 0 & Y_1 & -Y_1 \\ * & 0 & Y_2 & -Y_2 \\ * & * & Y_3 + Y_3^T & -Y_3 + Y_4^T \\ * & * & * & -Y_4 - Y_4^T \end{bmatrix} \xi(t) \quad (13)$$

Inserting (12-13) into (11), we obtain

$$E\{\dot{V}(t) + 2\alpha V(t)\} \leq \sum \sum \mu(z(t))\mu(z(t))\xi^T(t)\Theta\xi(t)$$

With matrix inequality (7), we know

$$\{E\dot{V}(t)\} < -2\alpha \{EV(t)\}$$

therefore

$$\{EV(t)\} < \{EV(0)\}e^{-\alpha t} \leq [\lambda_{\max}(P) + d\lambda_{\max}(Q) + \tau\lambda_{\max}(R)]\psi^2 E\{\|\psi(t)\|^2\} e^{-2\alpha t} \quad (14)$$

Obviously

$$E\{V(t)\} \geq \lambda_{\min}(P)E\{\|x(t)\|^2\}$$

From (14-15), we obtain

$$E\{\|x(t)\|\} < \sqrt{\frac{\lambda_{\max}(P) + d\lambda_{\max}(Q) + \tau\lambda_{\max}(R)\bar{\psi}^2}{\lambda_{\min}(P)}} E\{\|\psi(t)\|\} e^{-\alpha t}$$

With the Lyapunov stability theorem and the above inequality, we know that the system (6) is mean-square exponentially stable.

Theorem 2 For the given constants $\alpha > 0, 1 \geq \beta \geq 0$, if there exist positive-definite matrices $\bar{P}, \bar{Q}, \bar{R} \in R^{n \times n}$ and matrix $\bar{K} \in R^{m \times n}$, \bar{X}, \bar{Y} with appropriate dimensions, such that the following linear matrix inequality holds

$$\bar{\Theta} = \begin{bmatrix} \bar{\Theta}_{11} & \bar{\Theta}_{12} \\ * & \bar{\Theta}_{22} \end{bmatrix} < 0 \quad (16)$$

