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Stability of Stochastic Systems with Jumps

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This paper deals with stochastic stability of systems with Markovian jumps and Brownian motion. Mainly, we present sufficient conditions for quadratic stabilization of Ito type stochastic linear and nonlinear systems with Markovian jumps and Brownian motion using state feedback control. We also prove the guaranteed cost property of the proposed control strategy for the linear case.

Keywords: Systems with Markovian jumps and Brownian motion; Stochastic Stability; Markov Process; Ito Differential Equation; Lyapunov Function.

Classification Categories: 43E15

1 INTRODUCTION

Due to the need for modeling real world dynamic systems which may experience abrupt changes in their structure and parameters caused by component failures or repairs, changing environmental disturbances and subsystem interconnections, linear systems with Markovian jumping parameters have received recently a remarkable attention (see for example Ji and Chizeck [1], Wonham [2], Swarder [3], Mariton and Bertrand [4], Mariton [5, 6], Boukas [7] and references therein). For systems with

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Brownian motion, there exists in the literature many references dealing with the stochastic stability, see for example the works of Florchinger [8], Kushner [9], Willems and Willems [10], Khasminskii [11] and the references therein.

Most of the presently used analysis and design tools for feedback control systems presuppose that a sufficiently linear model of the process to be controlled is available. This model is not the true representation of the physical system under study since we always make a trade off between the model accuracy and the model complexity. The next step should be the use of a robust control law which enables the designer to explicitly consider the discrepancies between the model and the real process. Theoretically, this approach allows the designer to get results that are not only valid for the approximate model but do hold with certainty for the real process also.

In the control literature, the uncertain systems has received much attention. The quantity of papers and books dealing with this type of problem justifies from the practical point of view the interest of this topics. The Brownian motion, known also as Wiener process is the most studied and the most fascinating process. Many results for linear systems with Brownian motion exist. But to the best of our knowledge no results for linear systems with jumps and Brownian motion exists. Our primary aim is then to study the stochastic stability of this class of systems.

The main idea of this paper consists to use the Lyapunov function techniques to establish a sufficient condition of stability for systems with Markovian jumps and Brownian motion. In linear case, we prove that if the diffusion coefficient is upper bounded by a positive definite matrix (i.e. for two matrices C and D , when we say that the matrix inequality $C < D$ holds, this means that $C - D$ is a nonnegative matrix) then the system under consideration is stochastically stable. We state and prove the guaranteed cost property of the proposed control policy for linear system. We also establish a stochastic stability result for nonlinear system.

The paper is organized as follows. In section 2, we consider the linear system. Three theorems are presented and proved. Theorem 2.1 gives sufficient condition of stability of the system under consideration. Theorem 2.2 is an application of Theorem 2.1. Theorem 2.3 establishes the guaranteed cost property. In section 3, we consider the nonlinear system. A stochastic stability result is given. The contribution of the paper is summarized in section 4.

2 LINEAR SYSTEM

In this section, we will give a description of the class of systems under study. We will also state and prove the stochastic stability results.

2.1 Problem Statement

Let us consider a class of linear systems with Markovian jumps and Brownian motion described by:

$$dx(t) = [A(\xi(t))x(t) + B(\xi(t))u(t)] dt + \sum_{i=1}^k g_i(\xi(t))x(t)dW_i \quad (1)$$

$$x(0) = x_0 \quad (2)$$

where the vector $x(t) \in IR^n$ is the state of the system and the vector $u(t) \in IR^m$ is the control, and the vector $W_i(t)$, ($i = 1, \dots, k$) denotes the disturbance, assumed to be standard Brownian motion. The term $\sum_{i=1}^k g_i(\xi(t))x(t)dW_i$ represents the system uncertainty. The parameter $\xi(t)$ represents a continuous discrete-state Markov process taking values in a finite set $B = \{1, 2, \dots, s\}$ with transition probability, $Pr\{\xi(t + \delta t) = \beta | \xi(t) = \alpha\}$, given by:

$$Pr\{\xi(t + \delta t) = \beta | \xi(t) = \alpha\} = \begin{cases} q_{\alpha\beta}\delta t + o(\delta t), & \text{if } \alpha \neq \beta \\ 1 + q_{\alpha\alpha}\delta t + o(\delta t), & \text{if } \alpha = \beta \end{cases} \quad (3)$$

In this relation, $q_{\alpha\beta}$ stands for the transition probability rate from mode α to mode β and satisfies the following relations:

$$q_{\alpha\beta} \geq 0 \quad (4)$$

$$q_{\alpha\alpha} = - \sum_{\beta \in B, \alpha \neq \beta} q_{\alpha\beta} \quad (5)$$

In Eq. (1), $A(\xi(t))$, $B(\xi(t))$ and $g_i(\xi(t))$ are appropriately dimensioned matrices. These matrices are constant for a given value of $\xi(t)$.

In this paper, we will assume that the Markov process $\xi(t)$ and the noise processes $W_i(t)$ are independent. We also assume that perfect state information is available for feedback.

In order to control the system described by Eq. (1), we use the optimal control approach. The optimization problem consists of minimizing the cost function, J defined by:

$$J = \mathbf{E} \left\{ \int_0^{\infty} x'(t)Q(\xi(t))x(t) + u'(t)R(\xi(t))u(t)dt \right\} \quad (6)$$

subject to the constraints (1)–(2).

In relation (6), the cost weighting matrices $R(\xi(t))$ and $Q(\xi(t))$ are symmetric with $R(\xi(t))$ positive definite and $Q(\xi(t))$ positive semi-definite for each $\xi(t)$.

In the following, we will refer to the jump linear systems described by the following differential equation:

$$dx(t) = [A(\xi(t))x(t) + B(\xi(t))u(t)]dt \quad (7)$$

$$x(0) = x_0 \quad (8)$$

as a nominal system of system (1).

Let $x(t, x_0, \alpha, u)$ represent the corresponding solution of system (1) at time t when the control $u(\cdot)$ is used and the initial conditions are respectively x_0 and α .

The stability is always the first requirement for any control design. For our class of systems, we will use the concept of mean square stochastic stability given by the following definition:

DEFINITION 2.1 The system (1)–(3) is said to be stochastically stabilizable if, for all finite $x_0 \in \mathbb{R}^n$ and $\alpha \in B$, there exists a state feedback control with finite constant matrix $K(\alpha)$, i.e. $u(t) = -K(\alpha)x(t)$, such that the following relation holds:

$$\lim_{T \rightarrow \infty} E_{u(\cdot)} \left\{ \int_0^T x'(t, x_0, e\alpha, u)x(t, x_0, \alpha, u)dt \mid x_0, \alpha \right\} \leq x_0' \tilde{P} x_0 \quad (9)$$

with \tilde{P} is a symmetric and positive-definite matrix

In the remainder of this paper, we assume that the nominal system described by Eq. (1) is stochastically stable and the first component of the state vector, i.e. $x(t)$ in each mode is assumed to be available for feedback.

It was established (see Ji and Chizeck [1] (1990)) that the solution of the optimization problem without the disturbance, $W(t)$, is given by

$$u(\alpha, t) = -L_0(\alpha)x(t), \text{ when } \xi(t) = \alpha \quad (10)$$

where

$$L_0(\alpha) = R^{-1}(\alpha)B'(\alpha)K_0(\alpha) \quad (11)$$

The matrix $K_0(\alpha)$ is the solution of the following coupled algebraic Riccati equations:

$$\begin{aligned} A'(\alpha)K_0(\alpha) + K_0(\alpha)A(\alpha) - K_0(\alpha)B(\alpha)R^{-1}(\alpha)B'(\alpha)K_0(\alpha) + q_{\alpha\alpha}K_0(\alpha) \\ + \sum_{\beta=1, \beta \neq \alpha}^s q_{\alpha\beta}K_0(\beta) + Q(\alpha) = 0, \forall \alpha \in B \end{aligned} \quad (12)$$

In the rest of this section, we will construct the control law that stabilizes the system when the disturbance is upper bounded and also discuss the guaranteed cost property.

2.2 Main Results

In this subsection, we first give a theorem which states the condition for the mean square stability of the system described by Eqs. (1)–(3).

THEOREM 2.1 *If there exist positive definite matrices $Q(\alpha)$ and $R(\alpha)$ such that the following nonlinear coupled matrix equation has a positive definite solution $K(\alpha)$,*

$$\begin{aligned} A'(\alpha)K(\alpha) + K(\alpha)A(\alpha) - K(\alpha)B(\alpha)R^{-1}(\alpha)B'(\alpha)K(\alpha) + q_{\alpha\alpha}K(\alpha) \\ + \sum_{i=1}^k g'_i(\alpha)K(\alpha)g_i(\alpha) + \sum_{\beta=1, \beta \neq \alpha}^s q_{\alpha\beta}K(\beta) + Q(\alpha) = 0, \alpha \in B \end{aligned} \quad (13)$$

Then the system (1)–(3) is stochastically stable under the feedback control law $u = -L(\alpha)x(t)$ with

$$L(\alpha) = R^{-1}(\alpha)B'(\alpha)K(\alpha) \quad (14)$$

Proof Let the Lyapunov function $V(x, \alpha)$ be:

$$V(x, \alpha) = x'(t)K(\alpha)x(t) \quad (15)$$

Consider weak infinitesimal operator \tilde{A} of the joint process $(\xi(t), x(t))$ which is the natural analogue of the deterministic derivative of the Lyapunov function and is defined as follows:

$$\begin{aligned} \tilde{A}V(x, \alpha) &= \lim_{h \rightarrow 0} \frac{1}{h} [E\{V(x(t+h), \xi(t+h)) | x(t), \xi(t)\} \\ &= \alpha\} - V(x(t), \xi(t) = \alpha)] \end{aligned} \quad (16)$$

The weak infinitesimal operator is then given by:

$$\begin{aligned} \tilde{A}V(x, \alpha) &= x'(t) \{ [A(\alpha) - B(\alpha)R^{-1}(\alpha)B'(\alpha)K(\alpha)]' K(\alpha) \\ &+ K(\alpha)[A(\alpha) - B(\alpha)R^{-1}(\alpha)B'(\alpha)K(\alpha)] + \sum_{\beta \in B} q_{\alpha\beta} K(\beta) \} x(t) \\ &+ x'(t) \left[\sum_{i=1}^k g'_i(\alpha) K(\alpha) g_i(\alpha) \right] x(t) \end{aligned} \quad (17)$$

It is easy to see that

$$\tilde{A}V(x, \alpha) = x'(t)[Q(\alpha) + L'(\alpha)R(\alpha)L(\alpha)]x(t) \quad (18)$$

Therefore,

$$\frac{\tilde{A}V(x, \alpha)}{V(x, \alpha)} \leq - \frac{\lambda_{\min}[Q(\alpha) + K(\alpha)B(\alpha)R^{-1}(\alpha)B'(\alpha)K(\alpha)]}{\lambda_{\max}(K(\alpha))} \quad (19)$$

Then by Dynkin's formula and the Bellman-Gronwall lemma for all $\alpha \in B$, it follows that:

$$E[V(x, \alpha)] \leq \exp(-\gamma t)V(x_0, \alpha) \quad (20)$$

where

$$\gamma = \min_{\alpha \in B} \left\{ \frac{\lambda_{\min}[Q(\alpha) + K(\alpha)B(\alpha)R^{-1}(\alpha)B'(\alpha)K(\alpha)]}{\gamma_{\max}[K(\alpha)]} \right\} \quad (21)$$

Thus, after some manipulations (see Ji and Chizeck [1]) the Eq. (20) gives:

$$\lim_{T \rightarrow \infty} E \left[\int_0^T x'(t)K(\alpha)x(t)dt \middle| x_0, \xi(0) = \alpha \right] \leq \frac{1}{\gamma} x'_0 K(\alpha) x_0 \quad (22)$$

Since $K(\alpha) > 0$ for each $\alpha \in B$, thus

$$\lim_{T \rightarrow \infty} E \left[\int_0^T x'(t)x(t)dt \middle| x_0, \xi(0) = \alpha \right] \leq x'_0 \max_{\alpha \in B} \frac{K(\alpha)}{\gamma \|K(\alpha)\|} x_0 \quad (23)$$

which proves the theorem. \square

Notice that Eq. (13) is nonlinear and its solution is not easy to obtain. Next theorem gives a sufficient condition that avoids the solution of Eq. (13).

THEOREM 2.2 *Let $K_0(\alpha)$ denote the solution of the coupled algebraic Riccati equation (12). If the nominal system (7) is stochastically stable and there exists a positive definite matrices $Q^*(\alpha)$ such that the following relation holds:*

$$\sum_{i=1}^k g'_i(\alpha) K_0(\alpha) g_i(\alpha) \leq Q^*(\alpha) \quad (24)$$

Then the system (1) is stochastically stable.

Proof Let $K_0(\alpha)$ be the solution of the coupled algebraic Riccati equations (12), with $Q(\alpha) = Q^*(\alpha) + \epsilon Q_1$, where Q_1 is a positive definite symmetric matrix and ϵ is a positive number. It is easy to see that $K_0(\alpha)$ is solution of the following equation:

