Research Article

On Uniqueness of Strong Solution of Stochastic Systems

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1. Introduction

As is well known, a great deal of attention has recently been given to systems with stochastic multiplicative noise, due to the fact that the modelling of uncertainties using this kind of formulation has been found of many applications in engineering, finance, and so on. With the appearing of two classic books [1, 2], stochastic stability and stabilization of Itô differential systems were investigated by many researchers for several decades; we refer the readers to [3–5] and the references therein. More specifically, for linear time-invariant stochastic (LTIS) systems, most works are concentrated on the investigation of mean square stabilization, which has important applications in system analysis and design. Some necessary and sufficient conditions for the mean square stabilization of LTIS systems were obtained by many researchers in terms of the generalized algebraic Riccati equation (GARE) in [6], the linear matrix inequality (LMI) in [7], or the spectra of some operators in [4, 8]. In addition, as well known in the linear system theory, complete observability and detectability play important roles in system analysis and design. Complete observability was extended to define exact observability of stochastic Itô systems in [4]. Some of the other works on this subject can be found, for instance, [8–11].

In this paper, we concentrate our attention on the strong solution of GARE related to the critical mean square stabilization of stochastic systems. The main novelty is to analyze the systems with stochastic multiplicative noise in both state and control and to give the condition of uniqueness of the strong solution of GARE. In Section 2, we give some preliminaries including some definitions and theorems we will use. In Section 3, we mainly study the condition of uniqueness of the strong solution of a kind of stochastic systems and give a positive proof of Conjecture 10 in [9]. In Section 4, we give an example where the stochastic system has only a strong solution.

For convenience, we adopt the following traditional notations: $S^n$: the set of all symmetric matrices, their components may be complex; $R^k$: the $k$-dimensional real vector space with the usual inner product $\langle \cdot , \cdot \rangle$ and the corresponding 2-norm $\| \cdot \|$; $R^{m \times n}$: the space of all $m \times n$ matrices with entries in $R$; $A'(\text{Ker}(A))$: the transpose (kernel space) of a matrix $A$; $A \geq 0(A > 0)$: the positive semidefinite (positive definite) symmetric matrix $A$; $I$: the identity matrix; $\sigma(L)$: the spectral set of the operator or matrix $L$; $C^\infty_{\mathbb{R}^+}$: the space of nonanticipative stochastic processes $x(t) \in R^n$ with respect to an increasing $\sigma$-algebra $\{F_t\}_{t \geq 0}$ satisfying $E \int_0^\infty |x(t)|^2 dt < \infty$; $R^+$: the set of all positive real numbers. Finally, we make the assumption throughout this paper that all systems have real coefficients.
2. Preliminaries

In order to illustrate our main results in the next Section 3, first of all, we give some definitions and theorems we will use. Consider the following stochastic Itô differential system:

\[ dx = Ax dt + Cx dw, \]
\[ x(0) = x_0 \in R^n, \tag{1} \]

where \( A \in R^{n \times n} \) and \( C \in R^{n \times n} \) are real constant matrices and \( w(\cdot) \) is a standard one-dimensional Wiener process defined on the filtered probability space \((\Omega, F, P, F_t) \) with \( F_t = \sigma(w(s) \mid 0 \leq s \leq t) \). System (1) or \((A, C)\) is called mean square stable, if \( \lim_{t \to +\infty} E\|x(t)\|^2 = 0 \) for any deterministic initial state \( x(0) \in R^n \). If we set \( X(t) = Ex(t)x^\prime(t) \), by Itô’s formula, \( X(t) \) satisfies the following generalized Lyapunov differential equation:

\[ \dot{X} = AX + XA^\prime + CXC^\prime, \]
\[ X(0) = x(0)x^\prime(0). \tag{2} \]

Here \( \dot{X} \) denotes the time-derivative of \( X \). Motivated by (2), we introduce the following linear Lyapunov operator:

\[ \mathcal{L}_{AC} : X \in S^n \mapsto AX + AX^\prime + CXC^\prime \in S^n. \tag{3} \]

In [10], the following theorem gives a necessary and sufficient condition for the mean square stability of system (1) via the spectrum of \( \mathcal{L}_{AC} \), which is called “spectral criterion.” Other spectral criteria were found in [6, 12].

Theorem 1. System (1) is asymptotical mean square stable if and only if \( \sigma(\mathcal{L}_{AC}) \subset C^- \).

For a state feedback control law \( u = Kx \), one introduces a linear operator \( \mathcal{L}_K \) associated with the closed-loop system:

\[ dx = (A + BK)x dt + (C + DK)x dw, \]
\[ x(0) = x_0 \in R^n, \tag{4} \]

which is defined as \( \mathcal{L}_K : X \in S^n \mapsto (A + BK)X + X(A + BK)^\prime + (C + DK)X(C + DK)^\prime. \) Where \( A \in R^{n \times n}, B \in R^{n \times m}, C \in R^{n \times n}, D \in R^{m \times m} \), and \( K \in R^{m \times n} \) are all real constant matrices.

Definition 2. System (4) or \((A, B, C, D)\) is mean square stabilizable if and only if \( \sigma(\mathcal{L}_K) \subset C^- \) for some \( K \in R^{m \times n} \).

Definition 3. System \((A, B, C, D)\) is critical mean square stabilizable if and only if \( \mathcal{L}_K \subset C^{-0} \) for some \( K \in R^{m \times n} \).

The following theorems give the relations between the Lyapunov operator spectrum of deterministic system and the Lyapunov operator spectrum of stochastic system (see [9]), which will be used in the proof of the main results.

Theorem 4. For system (1), if \( \sigma(\mathcal{L}_{AC}) \subset C^{-0} \), then \( \sigma(A) \subset C^- \).

Theorem 5. For system (1), if \( \sigma(\mathcal{L}_{AC}) \subset C^- \), then \( \sigma(A) \subset C^- \).

In infinite horizon linear quadratic optimal control, stochastic stability and filtering (see, e.g. [3, 13, 14]), the following general algebra Ricatti equation

\[ PA + A^\prime P + PC + Q - (PB + C^\prime PD) \]
\[ \times (R + D^\prime PD)^{-1} (B^\prime P + D^\prime PC) = 0, \tag{5} \]

\[ R + D^\prime PD > 0, \quad Q \geq 0, \quad R > 0 \]

has many applications. In fact, GARE (5) is a generalized version of the following deterministic algebraic Ricatti equation (DARE):

\[ PA + A^\prime P + Q - PBR^{-1}B^\prime P = 0, \tag{6} \]

\[ R > 0, \quad Q \geq 0. \]

In [9], the following definitions and theorem are given.

Definition 6. A solution \( P \in S^n \) of GARE (5) is called a feedback stabilizing solution if \( \sigma(\mathcal{L}_K) \subset C^- \); \( P \) is called a strong solution if \( \sigma(\mathcal{L}_K) \subset C^{-0} \), where \( K = -(R + D^\prime PD)^{-1}(B^\prime P + D^\prime PC) \).

Definition 7. A solution \( \hat{P} \in S^n \) of GARE (5) is called the maximal solution if \( \hat{P} - P \geq 0 \) for any solution \( P \in S^n \) of GARE (5).

Theorem 8. Suppose system \((A, B, C, D)\) is mean square stabilizable, the weighting real matrices \( Q \geq 0, R > 0 \). Let \( \hat{P} \) be any real symmetric solution of the GARE

\[ PA + A^\prime P + C^\prime PC + Q - (PB + C^\prime PD) \]
\[ \times (\hat{R} + D^\prime PD)^{-1} (B^\prime P + D^\prime PC) = 0, \tag{7} \]

\[ \hat{R} + D^\prime PD > 0. \]

If \( R \geq \hat{R} \) and \( Q \geq \hat{Q} \), then GARE (5) has maximal solution \( P \geq 0 \) and \( \hat{P} \geq \hat{P} \). Moreover, \( P \) is a strong solution.

Remark 9. We know that if system \((A, B, C, D)\) is mean square stabilizable, then the maximal solution of (5) is a feedback stabilizing solution, but a strong solution of (5) may not be a feedback stabilizing solution.

In the following section, we will study the uniqueness of the strong solution of a kind of stochastic system; namely, the strong solution of (5) must be the maximal solution. Accordingly, we give a positive proof of Conjecture 10 in [9]. To end this section, we give this conjecture.

Conjecture 10. Assume system \((A, B, C, D)\) is mean square stabilizable, if GARE (5) admits a strong solution \( P \in S^n \), then \( P \) is also the maximal solution.

3. Main Results

In this section, we will give the condition of uniqueness of the strong solution of GARE (5). Here we only investigate the
strong solution of GARE (5) related to a class of stochastic systems.

Consider the following stochastic system:
\[
dx = (Ax + Bu) \, dt + Cx \, dw, \quad x(0) = x_0 \in R^n.
\] (8)

GARE (5) may become
\[
PA + A^T P + C^T PC + Q - PBR^{-1}B^T P = 0, \quad Q \geq 0, \quad R > 0.
\] (9)

Now we present a very useful lemmas which helps us give the condition of uniqueness of the strong solution.

**Lemma 11.** Let \( P_1 \in S^n \) be a solution of GARE (9). Then \( P_2 \in S^n \) satisfies GARE (9) if and only if \( D_0 = P_1 - P_2 \) satisfies \( A_0^T D_0 + D_0 A_0 + C^T D_0 C + D_0 B R^{-1} B^T D_0 = 0 \), where \( A_0 = A - BR^{-1}B^T P_1 \).

**Proof.** We know that the solutions \( P_1 \) and \( P_2 \) of GARE (9) satisfy the following two equations:
\[
P_1 A + A^T P_1 + C^T P_1 C + Q - P_1 B R^{-1} B^T P_1 = 0, \quad Q \geq 0, \quad R > 0.
\]
\[
P_2 A + A^T P_2 + C^T P_2 C + Q - P_2 B R^{-1} B^T P_2 = 0, \quad Q \geq 0, \quad R > 0.
\]

Two above equations are subtracted, by a series of computations, and we have
\[
A_0^T D_0 + D_0 A_0 + C^T D_0 C + D_0 B R^{-1} B^T D_0 = 0. \quad (11)
\]
The proof of Lemma 11 is complete.

**Lemma 12.** Suppose system (8) (or \( (A, B, C) \)) is mean square stabilizable, and the weighting real matrices \( Q \geq 0, R > 0 \). Then GARE (9) has the maximal solution which is only a feedback stabilizing solution of GARE (8).

**Proof.** The proof is easy, and we omit it.

**Theorem 13.** Assume that system \( (A, B, C) \) is mean square stabilizable and \( \bar{P} \) denotes the maximal solution of GARE (9). There does not exist the other strong solution of GARE (9) when \( |\operatorname{Re}(\lambda_i)| > |\operatorname{Re}(\mu_i)| \) for \( \lambda_i \in \sigma(A - BR^{-1}B^T \bar{P}) \) and \( \mu_i, \mu_j \in \sigma(C) \); that is, there is only a strong solution of GARE (9).

**Proof.** By Theorems 10 and 12 in [3], if system \( (A, B, C) \) is mean square stabilizable, then GARE (9) has the maximal solution \( \bar{P} \). By Theorem 8, it is also a strong solution. By contradiction, now assume that \( \tilde{P} \) is the other strong solution of GARE (9). Then
\[
\tilde{P} A + A^T \tilde{P} + C^T \tilde{P} C + Q - \tilde{P} B R^{-1} B^T \tilde{P} = 0, \quad Q \geq 0, \quad R > 0,
\]
\[
\bar{P} A + A^T \bar{P} + C^T \bar{P} C + Q - \bar{P} B R^{-1} B^T \bar{P} = 0, \quad Q \geq 0, \quad R > 0.
\]

In what follows, we will prove that \( \bar{P} = \tilde{P} \). Let \( H = \bar{P} - \tilde{P}(\geq 0) \). If \( H > 0 \), from Lemma 11, subtracting (13) from (12), we have
\[
A_0^T H + H A_0 + C^T H C + H B R^{-1} B^T H = 0; \quad (14)
\]
here \( A_0 = A - BR^{-1}B^T \tilde{P} \). From (14), we have
\[
A_0^T + BR^{-1} B^T H = -H^{-1} A_0^T H - H^{-1} C^T H C. \quad (15)
\]
Since \( |\operatorname{Re}(\lambda_i)| > |\operatorname{Re}(\mu_i)| \) for \( \lambda_i \in \sigma(A_0) \) and \( \mu_i, \mu_j \in \sigma(C) \), there exists a \( \theta \in \sigma(-H^{-1} A_0^T H - H^{-1} C^T H C) \) such that
\[
\operatorname{Re}(\theta) > 0. \quad 
\sigma(-H^{-1} A_0^T H - H^{-1} C^T H C) \subset \sigma(A_0 + BR^{-1} B^T H) = \sigma(A_0 + BR^{-1} B^T (\bar{P} - \tilde{P})) = \sigma(A - BR^{-1} B^T \bar{P} + BR^{-1} B^T (\bar{P} - \tilde{P})) = \sigma(A - BR^{-1} B^T \bar{P}) \subset C^{-0}, \text{ which is a contradiction.}
\]

So \( H > 0 \) does not hold. Hence \( H \geq 0. \) If \( H \geq 0 \) and \( H \neq 0 \), there exists \( x \in R^n \) such that \( H x = 0 \). Premultiplying by \( x^T \) and postmultiplying by \( x \) in (14) yield
\[
x^T A_0^T H x + x^T H A_0 x + x^T C^T H C x + x^T H B R^{-1} B^T H x = 0; \quad (16)
\]
we see that \( H C x = 0 \). Postmultiplying by \( x \) in (14) yields
\[
A_0^T H x + H A_0 x + C^T H C x + H B R^{-1} B^T H x = 0, \quad (17)
\]
we see that \( H A_0 x = 0 \). Hence \( \operatorname{Ker}(H) \) is an invariant subspace with respect to both \( A_0 \) and \( C \). By matrix theory, there exists an orthogonal matrix \( S \) such that
\[
S^T A_0 S = \begin{pmatrix} A_1 & * \\ 0 & A_2 \end{pmatrix}, \quad S^T C S = \begin{pmatrix} C_1 & * \\ 0 & C_2 \end{pmatrix}, \quad (18)
\]
\[
S^T H S = \begin{pmatrix} 0 & 0 \\ 0 & W \end{pmatrix}, \quad S^T B S = \begin{pmatrix} B_1 \\ B_2 \end{pmatrix}.
\]
And premultiplying by \( S^T \) and postmultiplying by \( S \) in (14) yield
\[
S^T A_0^T S S^T H S + S^T H S^T A_0 S + S^T C^T H S^T C S
+ S^T H S^T B S S^T R^{-1} S S^T B^T S^T H S = 0.
\]
So
\[
A_2^T W + WA_2 + C_2^T W C_2 + M = 0, \quad M \geq 0. \quad (20)
\]
Here,
\[
S^T H B R^{-1} B^T H S = \begin{pmatrix} 0 & * \\ 0 & M \end{pmatrix}. \quad (21)
\]
In turn, \( W > 0 \) in (20) satisfies
\[
A_1^s W + WA_2 + C_1^t W C_2 \leq 0. \quad (22)
\]
Hence \( \sigma(\mathcal{L}_{A-C}) \subset C^0 \) follows from Lyapunov theory and whenever \( \text{Re}(\lambda_i) = 0, \lambda_i \in \sigma(\mathcal{L}_{A-C}) \) and all associated Jordan blocks of \( \lambda_i \) are \( 1 \times 1 \). In fact, by Lemma 12, we know that \( \sigma(\mathcal{L}_{A-C}) \subset C^- \). By Theorem 5, \( \sigma(A_0) \subset C^- \), so \( \sigma(A_2) \subset C^- \). Finally,
\[
S'(A_0 + BR^{-1}B'H) S = \begin{pmatrix} A_1 & \ast \\ 0 & -W^{-1}A_1^s W - W^{-1}C_1^t W C_2 \end{pmatrix}.
\quad (23)
\]
Since \( |\text{Re}(\lambda_i)| > |\text{Re}(\mu_j)| \) for \( \lambda_i \in \sigma(A_2) \) and \( \mu_j \in \sigma(C_2) \), there exists a \( \theta_i \in \sigma(-W^{-1}A_1^s W - W^{-1}C_1^t W C_2) \) such that \( \text{Re}(\theta_i) > 0 \). If \( \sigma(-W^{-1}A_1^s W - W^{-1}C_1^t W C_2) \subset \sigma(A_0 + BR^{-1}B'H) = \sigma(A_0 + BR^{-1}B' (\tilde{P} - \bar{P})) = \sigma(A - BR^{-1}B' \tilde{P} + BR^{-1}B' (\tilde{P} - \bar{P})) = \sigma(A - BR^{-1}B' \tilde{P}) \). It directly provides that \( \sigma(\mathcal{L}_K) \) of system (8) must have unstable modes by Theorem 4 where \( K = -R^{-1}B' \tilde{P} \); namely, there is a \( \lambda \in \sigma(\mathcal{L}_K) \) such that \( \lambda > 0 \). It contradicts with \( \sigma(\mathcal{L}_K) \) of system (8) which is contained in \( C^- \). Hence \( H = 0 \). That is, \( \tilde{P} = \bar{P} \). The proof of Theorem 13 is complete.

Remark 14. In fact, the condition \( |\text{Re}(\lambda_i)| > |\text{Re}(\mu_j)| \) for \( \lambda_i \in \sigma(A - BR^{-1}B' \tilde{P}) \) and \( \mu_j \in \sigma(C) \) is a little stronger. From the proof of Theorem 13, if there exists a \( \theta_i \in \sigma(-W^{-1}A_1^s W - W^{-1}C_1^t W C_2) \) such that \( \text{Re}(\theta_i) > 0 \), the result still holds.

Remark 15. In fact, Theorem 13 gives a positive proof of Conjecture 10 when \( D = 0 \) in a large extent. That is, the result of Conjecture 10 holds for a big class of stochastic systems.

**Corollary 16.** Assume system \( (A, B, C) \) is mean square stabilizable; if GARE (9) admits a strong solution \( \tilde{P} \) such that \( \sigma(\mathcal{L}_{A-1^tB'P}) \subset C^- \), then it is also the maximal solution.

**Proof.** If GARE (9) admits a strong solution \( \tilde{P} \) such that \( \sigma(\mathcal{L}_{A-1^tB'P}) \subset C^- \); that is, \( \lambda_i \in \sigma(\mathcal{L}_{A-1^tB'P}) \neq 0 \), then \( \tilde{P} \) is a feedback stabilizing solution. Hence it is also the maximal solution by the uniqueness of stabilizing solution.

Remark 17. Under the conditions of Theorem 13, either \( \sigma(\mathcal{L}_{A-1^tB'P}) \subset C^- \) or \( \sigma(\mathcal{L}_{A-1^tB'P}) \subset C^- \) holds, where \( P \) is the solution of GARE (9).

Now we consider the positive semi-definiteness of the strong solution \( P \) of GARE (9); first of all, we give the following condition.

The condition \( H_1 \); for each \( \varepsilon > 0 \), there exists a \( \delta > 0 \), such that \( E\|x(t, x_0)\| < \varepsilon \) whenever \( t \geq 0 \) and \( \|x_0\| < \delta \).

**Theorem 18.** Assume system \( (A, B, C) \) is mean square stabilizable. If the strong solution \( P \) of GARE (9) has the property that any eigenvalue \( \lambda \in \sigma(\mathcal{L}_{A-1^tB'P}) \) satisfying \( \text{Re}(\lambda) = 0 \) is a simple characteristic root, then \( P \geq 0 \).

**Proof.** Consider the following three optimal performance values with the constraint of system (8):
\[
V_1(x_0) = \inf_{\mu \in U_{\infty}} \left\{ E \int_0^{\infty} (x' Q x + u' Ru) dt, \right. \quad \lim_{t \to +\infty} E \|x\| = 0 \left. \right\},
\]
\[
V_2(x_0) = \inf_{\mu \in U_{\infty}} \left\{ E \int_0^{\infty} (x' Q x + u' Ru) dt, \quad x(0) = x_0 \right\}.
\]
In fact, Let \( X = Ex \), using Itô formula and the knowledge of Kronecker multiplication in theory of matrices, the stochastic system (4) becomes a deterministic systems in [10]. By Theorem 5.2.3 in [15], the condition \( H_1 \) is equivalent to the eigenvalue, \( \lambda \in \sigma(\mathcal{L}_K) \) satisfying \( \text{Re}(\lambda) = 0 \), is a simple characteristic root.

From Lemma 4.1 in [9] and Theorem 5 in [10], we can, respectively, get
\[
V_1(x_0) = x_0 P_{\text{max}} x_0, \quad V_3(x_0) = x_0 P_{\text{min}} x_0.
\]
where \( P_{\text{min}} \) and \( P_{\text{max}} \) are the minimal and the maximal positive semidefinite solution of GARE (9). Similarly to proof of Theorem 5 in [10], we get \( V_3(x_0) = x_0 P_{\text{max}} x_0 \).

Hence,
\[
P_{\text{min}} \leq P \leq P_{\text{max}}, \quad P \geq 0.
\quad (27)
\]

**Corollary 19.** Assume system \( (A, B, C, D) \) is mean square stabilizable. If the strong solution \( P \) of GARE (5) has the property that any eigenvalue \( \lambda \in \sigma(\mathcal{L}_{A-1^tB'P}) \) satisfying \( \text{Re}(\lambda) = 0 \) is a simple characteristic root, then \( P \geq 0 \).

In particular, we consider the case of one dimension.

**Theorem 20.** Assume system \( (A, B, C) \) is mean square stabilizable. If GARE (9) admits a strong solution \( P \), then \( P \) is also the maximal solution.

**Proof.** The solutions of GARE (9) are \( P = (2A + C^2 \pm \sqrt{(2A + C^2)^2 + 4QB^2R^{-1}})/2B^2R^{-1} \). When \( 2A + C^2 \neq 0 \),
in [9] holds for one dimension.


4. An Example

In this section, we will exhibit the effectiveness of Theorem 13 by an example.

Example 1. Consider the following stochastic Itô differential system:

\[ dx = (Ax + Bu) \, dt + Cx \, dw, \tag{29} \]

Choose

\[ A = \begin{pmatrix} 4 & 2 & 3 \\ 0 & 5 & 2 \\ 0 & 0 & 6 \end{pmatrix}, \quad B = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \quad C = \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{pmatrix}, \tag{30} \]

\[ Q = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \quad R = \begin{pmatrix} 3 & 2 & 1 \\ 2 & 3 & 0 \\ 1 & 0 & 1 \end{pmatrix}. \]

Using Matlab, solving the maximal solution of GARE (9), that is, solving the optimal solution of the following SDP problem:

\[
\max \quad \mathcal{T}(P)
\]

subject to

\[
\begin{pmatrix} PA + A'P + C'PC + Q & PB \\ B'P & R \end{pmatrix} \succeq 0, \quad P \geq 0,
\]

we can get the following optimal solution \( P_0 \); by Theorem 10 in [3], we know that the given \( P_0 \) is the maximal solution of GARE (9).

One has

\[ P_0 = \begin{pmatrix} 7.7145 & 2.7249 & 1.8685 \\ 2.7249 & 12.5383 & 5.6915 \\ 1.8685 & 5.6915 & 21.4621 \end{pmatrix}. \tag{32} \]

So

\[
A_0 = A - BR^{-1}B'P_0
\]

\[ = \begin{pmatrix} -2.0441 & 19.9882 & 37.0819 \\ 3.1211 & -10.5049 & -23.2851 \\ 4.1756 & -22.6797 & -50.5440 \end{pmatrix}. \tag{33} \]

Hence,

\[
\sigma(A_0) = \{-64.4336, 2.5783, -1.2377\}, \quad \sigma(C) = \{1, 1, 1\}. \tag{34}
\]

Obviously, \( |\text{Re}(\lambda_i)| > |\text{Re}(\mu_j)| \) for \( \lambda_i \in \sigma(A_0) \) and \( \mu_j, \mu_j \in \sigma(C) \). Hence, by Theorem 13, the maximal solution of GARE (9) is only a strong solution.

5. Conclusion

In this paper, with the aid of the operator spectrum and generalized Lyapunov equation approach, we prove that the strong solution of GARE is also the maximal solution under certain condition, and it positively proves Conjecture 10 in [9]. From the proof of Theorem 13, we know that the uniqueness of strong solution of GARE (9) has a lot to do with \( \sigma(C) \) and the maximal solution \( P \). The condition \( |\text{Re}(\lambda_i)| > |\text{Re}(\mu_j)| \) for \( \lambda_i \in \sigma(A_0) \) and \( \mu_j, \mu_j \in \sigma(C) \) ensures that there exists a \( \theta_i \in \sigma(-W^{-1}A_2W - W^{-1}C_1W C_2) \), \( \text{Re}(\theta_i) > 0 \). However, whether there exists a weaker condition is still a challenge. Although we don’t completely prove Conjecture 10 in [9], we find that the result holds for a big class of stochastic systems; that is, GARE (9) has only a solution \( P \) which stabilized system (8) and mean square stabilized system (8); others cannot stabilize system (8). We will look deeper into the problem, which is perhaps related with the time invariant version of the stochastic system and of the generalized Riccati equations in [16].

Conflict of Interests

The authors declare that there is no conflict of interests regarding the publication of this paper.

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