

非线性广义半马尔可夫跳跃系统的 H_∞ 控制

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摘要

探讨一类非线性广义半Markov跳跃系统的随机稳定性和 H_∞ 控制问题。设计新型的Lyapunov-Krasovkii泛函(LKF), 用于减少冗余决策变量。同时, 引入参数依赖的互凸矩阵不等式(PDRCMI)来降低保守性, 保证了非线性广义Markov跳跃系统渐进稳定并满足性能, 最后, 通过数值算例验证了所得方法的有效性。

关键词

广义半马尔可夫跳跃系统, 非线性, 参数相关互凸矩阵不等式, H_∞ 控制

H_∞ Control of Nonlinear Singular Semi-Markov Jump Systems

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Abstract

The stochastic stability of a class of nonlinear singular semi-Markov jumping systems and H_∞ control problems are discussed. A new type of Lyapunov-Krasovkii functional (LKF) is designed to reduce redundant decision variables. At the same time, the parameter-dependent Convex Matrix Inequality (PDRCMI) is introduced to reduce the conservatism, which ensures the asymptotic stability and satisfies the performance of the nonlinear singular Markov jumping system, and finally, the effectiveness of the proposed method is verified by numerical examples.

Keywords

Singular Semi-Markov Jump System, Nonlinearity, Parameter-Dependent Reciprocally Convex

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1. 引言

广义系统因其广泛应用于电力的系统、航空航天工程、社会经济系统、化学过程、生物系统、网络分析、飞机控制系统、电力网络、太阳能热中央接收器、机器人机械手系统等[1]-[3]领域而引起了广泛的兴趣。几十年来,稳定性问题得到了广泛的研究[1]-[3]。然而,在这些实际系统的参数和结构中,一些现象,如子系统互连的修改、突变环境干扰、组件故障或维修等原因,可能导致系统结构,参数发生随机变化。而幸运的是,这些变化可以用马尔可夫跳跃系统(MJSs)适当地描述,并且每种操作模式都与一个动态系统关联,其中模型转换在一个马尔可夫过程的控制下。因此,学术界和工业界的大量关注都集中在广义马尔可夫跳跃系统(SMJSs)的研究上。

MJSs作为一种特殊的随机系统,由于其理论意义已经在网络系统控制、能源、制造和经济系统[1]-[6]中得到了广泛的研究。许多问题已经解决,取得良好的结果,例如, H_∞ 控制滤波的不确定时滞 MJSs [2], 具有部分未知转移速率的 MJSs 的稳定性分析[3], 不定二次最优控制问题[4], 基于分散二维马尔可夫跳跃系统的故障检测[7]。

2. 新判据

2.1. 模型介绍与假设

本文采用以下记号本文考虑下述不确定广义半马尔可夫跳跃系统:

$$\begin{cases} E\dot{x}(t) = (A(r_t) + \Delta A(r_t))x(t) + (A_d(r_t) + \Delta A_d(r_t))x(t-d(t)) + B_w(r_t)\varpi(t) \\ \quad + B(r_t)(u(t) + f(t, x(t), r(t))) + B_d(r_t)(u(t-d(t)) + f(t, x(t-d(t)), r(t))) \\ z(t) = C(r_t)x(t) + C_d(r_t)x(t-d(t)) + D(r_t)w(t) \end{cases} \quad (2.1)$$

其中 $x(t) \in \mathbb{R}^n$ 为系统状态向量, $u(t) \in \mathbb{R}^m$ 为控制输入, $z(t) \in \mathbb{R}^p$ 为控制输出, $w(t) \in \mathbb{R}^q$ 为外部扰动, 矩阵 $E \in \mathbb{R}^{n \times n}$ 可能是奇异矩阵, $\text{Rank}(E) = r \leq n$, $f(t, x(t))$ 为非线性函数。 $A(r_t)$, $A_d(r_t)$, $B(r_t)$, $B_d(r_t)$, $B_w(r_t)$, $C(r_t)$, $C_d(r_t)$ 和 $D(r_t)$ 是适当维度的已知常数矩阵。 $\Delta A(r_t)$ 和 $\Delta A_d(r_t)$ 是未知时变矩阵。时间延迟 $d(t)$ 满足 $0 \leq d_1 \leq d(t) \leq d_2$, $h_1 \leq \dot{d}(t) \leq h_2$, $\forall t \geq 0$, d_1, d_2, h_1, h_2 是给定常数边界, 同时定义 $0 < d_{12} = d_2 - d_1$ 。模态 $\{r_t, t \geq 0\}$ 是连续时间半马尔可夫过程, 该过程在有限集中取值 $S = \{1, 2, \dots, s\}$ [1], 转移速率矩阵 $\Pi = \{\pi_{ij}\}$ 如下定义:

$$P\{r(t+\Delta) = j | r(t) = i\} = \begin{cases} \pi_{ij}\Delta + o(\Delta), & i \neq j, \\ 1 + \pi_{ii}\Delta + o(\Delta), & i = j, \end{cases}$$

这里 $\Delta > 0$, $\lim_{\Delta \rightarrow 0} \frac{o(\Delta)}{\Delta} = 0$ 及 $\pi_{ij}(h) \geq 0 (i, j \in S, i \neq j)$, 指将 i 从 t 的模态转变到 $t+\Delta$ 的 j 模态, $\forall i \in S$ 都有

$\pi_{ij} = -\sum_{j=1, j \neq i}^s \pi_{ij}$ 。为了便于表达定义:

$$A(r_t) = A_i, A_d(r_t) = A_{di}, B(r_t) = B_i, B_d(r_t) = B_{di}, B_w(r_t) = B_{wi}, C(r_t) = C_i, C_d(r_t) = C_{di}, D(r_t) = D_i$$

$$\Delta A(r_t) = \Delta A_i, \Delta A_d(r_t) = \Delta A_{di}, f(t, x(t), r(t)) = f(t), f(t, x(t-d(t)), r(t)) = f(t-d(t)).$$

为了简化系统的形式, 我们可以将系统(2.1)重写如下:

$$\begin{cases} E\dot{x}(t) = (A_i + \Delta A_i)x(t) + (A_{di} + \Delta A_{di})x(t-d(t)) + B_i(u(t) + f(t)) \\ \quad + B_{di}(u(t-d(t)) + f(t-d(t))) + B_{wi}\varpi(t) \\ Z(t) = C_i x(t) + C_{di}x(t-d(t)) + D_i w(t) \end{cases} \quad (2.2)$$

假设转移速率矩阵 $\Pi \triangleq (\pi_{ij})$ 通常不确定, 我们为系统(2.1)计算转移速率矩阵, 如下所示:

$$\begin{bmatrix} \hat{\pi}_{11} + \Delta\pi_{11} & ? & \hat{\pi}_{13} + \Delta\pi_{13} & \cdots & ? \\ ? & ? & \hat{\pi}_{23} + \Delta\pi_{23} & \cdots & \hat{\pi}_{2s} + \Delta\pi_{2s} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ ? & \hat{\pi}_{s2} + \Delta\pi_{s2} & ? & \cdots & \hat{\pi}_{ss} + \Delta\pi_{ss} \end{bmatrix} \quad (2.3)$$

其中 $\hat{\pi}_{ij}$ 和 $\Delta\pi_{ij} \in [-\delta_{ij}, \delta_{ij}] (\delta_{ij} \geq 0)$ 分别代表估计值和不确定转移速率 π_{ij} 的估计误差。 $\hat{\pi}_{ij}$ 和 δ_{ij} 为已知, “?” 是完全未知的。对任意 $i \in S$, 集合 U^i 表示 $U^i = U_k^i \cup U_{uk}^i$ 。其中 $U_k^i \triangleq \{j: \pi_{ij} \text{ 为已知}, j \in S\}$, $U_{uk}^i \triangleq \{j: \pi_{ij} \text{ 为未知}, j \in S\}$ 。此外, 若 $U_k^i \neq \emptyset$, 它可以被描述为 $U_k^i = \{k_1^i, k_2^i, \dots, k_{m_i}^i\}$, 其中 $k_{m_i}^i \in S^+$ 表示矩阵 Π 的第 i 排的第 m 个已知边界。

备注 2.1 无论是有限不确定的 TR(BUTR)或部分未知 TR(PUTR)模型均不如上述不确定转移速率的描述那么宽泛。重写了以下两个不确定的模型:

$$\begin{bmatrix} \hat{\pi}_{11} + \Delta_{11} & \hat{\pi}_{12} + \Delta_{12} & \cdots & \hat{\pi}_{1s} + \Delta_{1s} \\ \hat{\pi}_{21} + \Delta_{21} & \hat{\pi}_{22} + \Delta_{22} & \cdots & \hat{\pi}_{2s} + \Delta_{2s} \\ \vdots & \vdots & \ddots & \vdots \\ \hat{\pi}_{s1} + \Delta_{s1} & \hat{\pi}_{s2} + \Delta_{s2} & \cdots & \hat{\pi}_{ss} + \Delta_{ss} \end{bmatrix} \quad (2.4)$$

其中 $\hat{\pi}_{ij} - \delta_{ij} \geq 0 (\forall j \in S, j \neq i)$, $\hat{\pi}_{ii} = -\sum_{j=1, j \neq i}^s \hat{\pi}_{ij}$ 和 $\delta_{ii} = -\sum_{j=1, j \neq i}^s \delta_{ij}$ 。

$$\begin{bmatrix} \pi_{11} & ? & \pi_{13} & \cdots & ? \\ ? & ? & \pi_{23} & \cdots & \pi_{2s} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ ? & \pi_{s2} & ? & \cdots & \pi_{ss} \end{bmatrix} \quad (2.5)$$

显然, 当 $U_k^i \neq \emptyset, \forall i \in S$ 时, GUTR 模型(2.3)简化为 BUTR 模型(2.5), 若 $\delta_{ij} = 0, \forall i \in S, \forall j \in U_k^i$ 则简化为 PUTR 模型(2.5)。显然, BUTR 或 PUTR 模型不如 GUTR 模型(2.3)通用, 这意味着它更实用。

假设已知 TR 的估计值如下定义:

假设 2.1 若 $U_k^i = S$, 则 $\hat{\pi}_{ij} - \delta_{ij} \geq 0, (\forall j \in S, j \neq i)$, $\hat{\pi}_{ii} = -\sum_{j=1, j \neq i}^s \hat{\pi}_{ij} \leq 0, \delta_{ii} = \sum_{j=1, j \neq i}^s \delta_{ij} > 0$;

假设 2.2 若 $U_k^i \neq S$ 且 $i \in U_k^i$, 则 $\hat{\pi}_{ij} - \delta_{ij} \geq 0, (\forall j \in U_k^i, j \neq i)$, $\hat{\pi}_{ii} + \delta_{ii} \leq 0, \sum_{j \in U_k^i} \hat{\pi}_{ij} \leq 0$;

假设 2.3 若 $U_k^i \neq S$ 且 $i \in U_k^i$, 则 $\hat{\pi}_{ij} - \delta_{ij} \geq 0, (\forall j \in U_k^i)$ 。

这三个假设是基于 TR 特征而得出的, 由此可以推断它们具有合理性。

$$\left(e.g. \pi_{ij} \geq 0 (\forall j \in S, j \neq i) \text{ and } \pi_{ii} = -\sum_{j=1, j \neq i}^s \pi_{ij} \right).$$

假设 2.4 在本文中所涉及的不确定性属于有界范数的, 假设如下:

$$[\Delta A_i \ \Delta A_{di}] = M_i F_i(t) [N_i \ N_{di}], \forall i \in S$$

已知的常数矩阵 M_i, N_i, N_{di} 具有适当的维度, 使 $F_i(t)$ 得满足 $F_i^T(t)F_i(t) \leq I, i \in S$ 。

假设 2.5 非线性函数 $f(t, \xi(t))$ 满足

$$\begin{cases} f^T(t)f(t) \leq x^T(t)T^2x(t) \\ f^T(t-d(t))f(t-d(t)) \leq x^T(t-d(t))T^2x(t-d(t)) \end{cases} \text{其中 } T = \text{diag}\{\sigma_1, \sigma_2, \dots, \sigma_s\}.$$

定义 2.1 [7] 若 $\det(s\hat{E} - \hat{A}_i)$ 对任意 $r_i = i \in S$ 均不为 0, 则系统(2.2)正则。

1) 若对任意 $r_i = i \in S$ 有 $\text{deg}(\det(s\hat{E} - \hat{A}_i)) = \text{rank}(\hat{E})$, 则系统(2.2)被称为无脉冲。

2) 当 $w(t) = 0, u(t) = 0$ 时且存在标量 $M(r_0, \varphi(t)) > 0$ 使得

$$\lim_{t \rightarrow \infty} \varepsilon \left\{ \int_0^\infty \|x(t)\|^2 dt \mid r_0, x(t) = \varphi(t), t \in [-h, 0] \right\} \leq M(r_0, \varphi(t)), \text{ 则(2.2)随机稳定的。}$$

3) 当 $w(t) = 0, u(t) = 0$ 时, 若系统(2.2)是正则的、无脉冲的, 随机稳定的, 则称它是随机容许的。

定义 2.2 对于给定的标量 $\gamma > 0$, 系统(2.2)是随机容许的并满足 H_∞ 性能 γ , 如果它在 $w(t) = 0$ 及零初始条件下是随机容许的, 非零的 $w(t) \in L_2[0, \infty)$, 满足以下条件:

$$\varepsilon \left\{ \int_0^\infty (z^T(t)z(t) - \lambda^2 w^T(t)w(t)) dt \right\} < 0.$$

引理 2.1 设 H 和 G 是具有适当维数的实数矩阵, 并且 $F_i^T(t)F_i(t) \leq I$ 。以下不等式适用于任何标量 $\lambda > 0$:

$$1) HF_i(t)G^T + GF_i^T(t)H^T \leq \lambda GG^T + \lambda^{-1}HH^T, \quad 2) \pm 2H^TG \leq H^TH + G^TG.$$

引理 2.2 设 $C \in R^{n \times n}, B \in R^{m \times m}$ 为正定矩阵, 当且仅当 $A > B^TC^{-1}B$, 则对 $\forall A \in R^{m \times m}$ 有

$$F = \begin{pmatrix} A & B^T \\ B & C \end{pmatrix} > 0$$

2.2. 主要结果

对于给定的标量 $0 \leq d_1 \leq d_2, h_1, h_2, \varepsilon_1 \geq 0, \varepsilon_2 \geq 0$ 和 $\gamma > 0$, 如果存在矩阵 $P_i > 0, R_{1i} > 0, R_{2i} > 0, R_{3i} > 0, Q_1 > 0, Q_2 > 0, S_1 > 0, S_2 > 0, Z_1 > 0, Z_2 > 0, Y_1, Y_2$ 使得以下不等式对任意 $i \in I$ 成立, 则系统渐进稳定的并具有相应的 H_∞ 衰减指数 γ

$$E^T P_i = P_i^T E > 0,$$

$$\Omega^{[m]} = \begin{bmatrix} E^{[m]} & E_{12} \\ & E_{22} \end{bmatrix} < 0, m = 1, 2, 3, 4,$$

$$\sum_{j=1}^N \pi_{ij} R_{1j} - Q_1 \leq 0, \sum_{j=1}^N \pi_{ij} R_{3j} - Q_2 \leq 0,$$

$$\sum_{j=1}^N \pi_{ij} R_{1j} - Q_1 \leq 0, \sum_{j=1}^N \pi_{ij} R_{3j} - Q_2 \leq 0, R_{2i} - R_{3i} \leq 0, \Phi_1 > 0, \Phi_2 > 0,$$

$$E(d(t), \dot{d}(t)) = E_{1i\mu|d(t), \dot{d}(t)} + E_2 + E_3,$$

$$E_{1i\mu|d(t), \dot{d}(t)} = \Xi + \Lambda_1^T \left(\sum_{j=1}^N \pi_{ij} P_j \right) \Lambda_1 + x^T(t) E^T \sum_{j=1}^N \pi_{ij} P_j x(t),$$

$$\begin{aligned}
 E_{i_i|d(t),d(t)} &= \text{sym}(\Lambda_1^T P_i \Lambda_2) + \Lambda_1^T \left(\sum_{j=1}^N \pi_{ij} P_j \right) \Lambda_1 + e_1^T (R_{li} + R_{3i} + d_1 Q_1 + d_2 Q_2 + \Theta_1) e_1 \\
 &\quad + e_2^T (\lambda_{i6}^{-1} T^2 - R_{li}) e_2 - e_3^T R_{2i} e_3 + e_5^T \left(d_2^2 S_1 + d_{12}^2 S_2 + \frac{1}{2} d_1^2 (Z_1 + Z_2) \right) e_5 \\
 &\quad - \Lambda_6^T \bar{Z}_1 \Lambda_6 - \Lambda_7^T \bar{Z}_2 \Lambda_7 + x^T(t) E^T \sum_{j=1}^N \pi_{ij} P_j x(t) + mte_4^T (R_{2i} - R_{3i}) e_4 + e_1^T P_i \Theta_2 e_2, \\
 \Xi &= \text{sym}(\Lambda_1^T P_i \Lambda_2) + e_1^T (R_{li} + R_{3i} + d_1 Q_1 + d_2 Q_2 + \Theta_1) e_1 + e_2^T (\lambda_{i6}^{-1} T^2 - R_{li}) e_2 - e_3^T R_{2i} e_3 \\
 &\quad + mte_4^T (R_{2i} - R_{3i}) e_4 + e_1^T P_i \Theta_2 e_2 + e_5^T \left(d_2^2 S_1 + d_{12}^2 S_2 + \frac{1}{2} d_1^2 (Z_1 + Z_2) \right) e_5 - \Lambda_6^T \bar{Z}_1 \Lambda_6 - \Lambda_7^T \bar{Z}_2 \Lambda_7, \\
 E_2 &= -\Gamma_1^T \Phi_1 \Gamma_1 - \Gamma_2^T \Phi_2 \Gamma_2, E_3 = e_{3i\mu}^T e_{3i\mu} - e_{14}^T \gamma^2 I e_{14}, \\
 E_{12} &= \left[\sqrt{1-\alpha_1} \Lambda_3^T Y_1 \sqrt{\alpha_1} \Lambda_4^T Y_1^T \sqrt{1-\alpha_2} \Lambda_5^T Y_2 \sqrt{\alpha_2} \Lambda_4^T Y_2^T \right], \\
 \Phi_1 &= \begin{bmatrix} (2-\alpha_1) \bar{S}_1 + (1-\alpha_1) \varepsilon_1 I & Y_1 \\ & (1+\alpha_1) \bar{S}_1 + \alpha_1 \varepsilon_1 I \end{bmatrix}, \\
 \Phi_2 &= \begin{bmatrix} (2-\alpha_2) \bar{S}_2 + (1-\alpha_2) \varepsilon_2 I & Y_2 \\ & (1+\alpha_2) \bar{S}_2 + \alpha_2 \varepsilon_2 I \end{bmatrix}, \\
 \Theta_1 &= \sum_{j=1}^N \pi_{ij} P_j + P_i \left(2A_i + 2B_i K + \lambda_{i1}^{-1} N_i^T N_i + \lambda_{i3}^{-1} N_{ki}^T N_{ki} + \lambda_{i1} M_i M_i^T P_i^T \right. \\
 &\quad \left. + \lambda_{i3} B_i M_i M_i^T P_i^T B_i^T + \lambda_{i6} B_{di} B_{di}^T P_i^T \right) + \lambda_{i5}^{-1} T^2 \\
 \Theta_2 &= 2A_{di} + 2B_{di} K_{di} + \lambda_{i2}^{-1} N_{di}^T N_{di} + \lambda_{i4}^{-1} N_{kdi}^T N_{kdi} + \lambda_{i2} M_i M_i^T P_i^T + \lambda_{i4} B_i M_i M_i^T P_i^T B_i^T, \\
 E_{22} &= \text{diag} \{ -\bar{S}_1, -\bar{S}_1, -\bar{S}_2, -\bar{S}_2 \}, \\
 \Lambda_1 &= \left[(d(t) - d_1) e_6^T \quad (d_2 - d(t)) e_7^T \quad d(t) e_8^T \quad (d(t) - d_1)^2 e_{10}^T \quad (d_2 - d(t))^2 e_{11}^T \quad d(t)^2 e_{12}^T \right]^T, \\
 \Gamma_1 &= \left[\Lambda_3^T \quad \Lambda_4^T \right]^T, \Gamma_2 = \left[\Lambda_5^T \quad \Lambda_4^T \right]^T, \\
 \Lambda_2 &= \left[E^T e_2^T - mtE^T e_4^T \quad mtE^T e_4^T - E^T e_3^T e_1^T - mtE^T e_4^T \quad (d(t) - d_1) E^T e_1^T - (d(t) - d_1) e_6^T \right. \\
 &\quad \left. (d_2 - d(t)) E^T e_1^T - (d_2 - d(t)) e_7^T \quad d(t) E^T e_1^T - d(t) e_8^T \right]^T, \\
 \Lambda_3 &= \left[r_1^T \quad r_2^T \quad r_3^T \right]^T, \Lambda_4 = \left[r_4^T \quad r_5^T \quad r_6^T \right]^T, \Lambda_5 = \left[r_7^T \quad r_8^T \quad r_9^T \right]^T, \Lambda_6 = \left[r_{10}, r_{11} \right]^T, \Lambda_7 = \left[r_{12}, r_{13} \right]^T, \\
 r_1 &= e_1 - e_4, r_2 = e_1 + e_4 - 2e_8, r_3 = e_1 - e_4 - 6e_8 + 12e_{12}, \\
 r_4 &= e_4 - e_3, r_5 = e_3 + e_4 - 2e_7, r_6 = e_3 - e_4 - 6e_7 + 12e_{11}, \\
 r_7 &= e_2 - e_4, r_8 = e_2 + e_4 - 2e_6, r_9 = e_2 - e_4 - 6e_6 + 12e_{10}, \\
 r_{10} &= e_1 - e_9, r_{11} = e_1 + 2e_9 - 6e_{13}, r_{12} = e_2 - e_9, r_{13} = e_2 - 4e_9 + 6e_{13}, \\
 \mu_1 &= (1-\alpha_1) \Lambda_3^T Y_1 \bar{S}_1^{-1} Y_1 \Lambda_3 + \alpha_1 \Lambda_4^T Y_1 \bar{S}_1^{-1} Y_1 \Lambda_4, \mu_2 = (1-\alpha_2) \Lambda_5^T Y_2 \bar{S}_2^{-1} Y_2 \Lambda_5 + \alpha_2 \Lambda_6^T Y_2 \bar{S}_2^{-1} Y_2 \Lambda_6, \\
 \alpha_1 &= \frac{d(t)}{d_2}, \alpha_2 = \frac{d(t) - d_1}{d_{12}}, mt = 1 - \dot{d}(t), \bar{S}_1 = \text{diag} \{ S_1, 3S_1, 5S_1 \}, \bar{S}_{2i} = \text{diag} \{ S_2, 3S_2, 5S_2 \}, \\
 \bar{Z}_1 &= \text{diag} \{ 2Z_1, 4Z_1 \}, \bar{Z}_2 = \text{diag} \{ 2Z_2, 4Z_2 \}, \\
 e_i &= \left[0_{(2n+k) \times (i-1)(2n+k)} \quad I_{(2n+k)} \quad 0_{(2n+k) \times (13-i)(2n+k)} \quad 0_{(2n+k) \times (m+q+l)} \right], i = 1, \dots, 14.
 \end{aligned}$$

证明: 首先证明系统是正则的、无脉冲的, 存在 M, N 满足

$$MEN = \begin{bmatrix} I_r & 0 \\ 0 & 0 \end{bmatrix}, MA_i N = \begin{bmatrix} A_{i1} & A_{i2} \\ A_{21} & A_{22} \end{bmatrix}, M^{-T} P_i N = \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix},$$

且 $E^T P_i = P_i^T E$, 可知 $P_{12} = 0$. 将左右乘以 N 和 N^T 有 $\Omega_{i11} < 0$, 则 $A_{22}^T P_{22} + P_{22}^T A_{22} < 0$, 故 A_{22} 是非奇异的. 因此 $\det(A_{22}) \neq 0$ 和 $\det(\det(sE - A_i)) = r = \text{rank}(E)$, 则 (E, A_i) 是正则的、无脉冲的. 同样地, 进行左右分别乘以 $[I \ \dots \ I]_{7 \times 1}^T$ 和 $[I \ \dots \ I]_{7 \times 1}$ 有 $(A_i + A_{di})^T P_i + P_i^T (A_i + A_{di}) + \sum_{j=1}^s \pi_{ij} E^T P_j E < 0$ 可知 $(E, A_i + A_{di})$ 是正则的, 无脉冲的. 根据[2]中定义 1 系统是正则, 无脉冲的. 接下来, 证明系统是渐进稳定, 构造 LKF 函数如下.

$$V(x_t) = \sum_{k=1}^5 V_k,$$

$$V_1(x_t) = x^T(t) E^T P_t x(t) + \eta^T(t) P_t \eta(t),$$

$$V_2(x_t) = \int_{t-d_1}^t x^T(s) R_{1i} x(s) ds + \int_{-d_1}^0 \int_{t-d_1}^t x^T(s) Q_1 x(s) ds d\theta,$$

$$V_3(x_t) = \int_{t-d_2}^{t-d(t)} x^T(s) R_{2i} x(s) ds + \int_{t-d(t)}^t x^T(s) R_{3i} x(s) ds + \int_{-d_2}^0 \int_{t+\theta}^t x^T(s) Q_2 x(s) ds d\theta,$$

$$V_4(x_t) = d_2 \int_{-d_2}^0 \int_{t+\theta}^t \dot{x}^T(s) E^T S_1 E \dot{x}(s) ds d\theta + d_{12} \int_{-d_2}^{-d_1} \int_{t+\theta}^t \dot{x}^T(s) E^T S_2 E \dot{x}(s) ds d\theta,$$

$$V_5(x_t) = \int_{t-d_1}^t \int_u^t \dot{x}^T(s) E^T Z_1 E \dot{x}(s) ds d\theta du + \int_{t-d_1}^t \int_{t-d_1}^u \int_\theta^t \dot{x}^T(s) E^T Z_2 E \dot{x}(s) ds d\theta du,$$

$$\eta(t) = \left[\int_{t-d(t)}^{t-d_1} x^T(t) E^T ds \quad \int_{t-d_2}^{t-d(t)} x^T(t) E^T ds \quad \int_{t-d(t)}^t x^T(t) E^T ds \quad \int_{t-d(t)}^{t-d_1} \int_\theta^t x^T(t) E^T ds d\theta \right. \\ \left. \int_{t-d_2}^{t-d(t)} \int_\theta^t x^T(t) E^T ds d\theta \quad \int_{t-d(t)}^t \int_\theta^t x^T(t) E^T ds d\theta \right]^T.$$

L 为弱无穷小算子, 有:

$$LV_1(x_t) = 2\Psi^T(t) \Lambda_1^T P_t \Lambda_2 \Psi(t) + 2x^T(t) E P_t \dot{x}(t) + \Psi^T(t) \Lambda_1^T \sum_{j=1}^N \pi_{ij} P_j \Lambda_1 \Psi(t) + x^T(t) E^T \sum_{j=1}^N \pi_{ij} P_j x(t) \\ = \Psi^T(t) \left[\text{sym}(\Lambda_1^T P_t \Lambda_2) + \Lambda_1^T \left(\sum_{j=1}^N \pi_{ij} P_j \right) \Lambda_1 \right] \Psi(t) + x^T(t) \sum_{j=1}^N \pi_{ij} P_j x(t) + \text{sym}[x^T(t) E^T P_t \dot{x}(t)] \\ = \Psi^T(t) \left[\text{sym}(\Lambda_1^T P_t \Lambda_2) + \Lambda_1^T \left(\sum_{j=1}^N \pi_{ij} P_j \right) \Lambda_1 \right] \Psi(t) + x^T(t) \sum_{j=1}^N \pi_{ij} P_j x(t) + 2x^T(t) P_t A_i x(t) \\ + 2x^T(t) P_t \Delta A_i x(t) + 2x^T(t) P_t B_i \Delta K_i x(t) + 2x^T(t) P_t A_{di} x(t-d(t)) + 2x^T(t) P_t \Delta A_{di} x(t-d(t)) \\ + 2x^T(t) P_t B_i K_i x(t) + 2x^T(t) P_t B_{di} K_{di} x(t-d(t)) + 2x^T(t) P_t B_{di} \Delta K_{di} x(t-d(t)) \\ + 2x^T(t) P_t B_i f(t) + 2x^T(t) P_t B_{di} f(t-d(t)),$$

$$LV_2(x_t) = \Psi^T(t) (e_1^T (R_{1i} + d_1 Q_1) e_1 - e_2^T R_{1i} e_2) \Psi(t) + \int_{t-d_1}^t x^T(s) \left(\sum_{j=1}^N \pi_{ij} R_{1j} - Q_1 \right) x(s) ds,$$

$$LV_3(x_t) = \Psi^T(t) (e_1^T (R_{3i} + d_2 Q_2) e_1 + mte_4^T (R_{2i} - R_{3i}) e_4 - e_3^T R_{2i} e_3) \Psi(t) \\ + \int_{t-d_2}^{t-d(t)} x^T(s) \left(\sum_{j=1}^N \pi_{ij} (R_{2j} - R_{3j}) \right) x^T(s) ds + \int_{t-d_2}^t x^T(s) \left(\sum_{j=1}^N \pi_{ij} R_{3j} - Q_2 \right) x(s) ds,$$

$$\begin{aligned}
 LV_4(x_t) &= \Psi^T(t) \left(e_5^T (d_2^2 S_1 + d_{12}^2 S_2) e_5 \right) \Psi(t) - d_2 \int_{t-d_2}^t E^T \dot{x}^T(s) S_1 E \dot{x}(s) ds \\
 &\quad - d_{12} \int_{t-d_2}^{t-d_1} E^T \dot{x}^T(s) S_2 E \dot{x}(s) ds, \\
 LV_5(x_t) &= \Psi^T(t) \left(\frac{d_1^2}{2} e_5^T (Z_1 + Z_2) e_5 \right) \Psi(t) - \int_{t-d_1}^t \int_{\theta}^t E^T \dot{x}^T(s) Z_1 E \dot{x}(s) ds d\theta \\
 &\quad - \int_{t-d_1}^t \int_{t-d_1}^{\theta} E^T \dot{x}^T(s) Z_2 E \dot{x}(s) ds d\theta, \\
 \Psi(t) &= [\psi_1^T(t) \ \psi_2^T(t) \ \psi_3^T(t) \ w^T(t)]^T, \\
 \psi_1(t) &= [x^T(t) \ x^T(t-d_1) \ x^T(t-d_2) \ x^T(t-d(t)) \ \dot{x}^T(t) E^T]^T, \\
 \psi_2(t) &= \left[\frac{1}{d(t)-d_1} \int_{t-d_1}^{t-d(t)} x^T(s) E^T ds \ \frac{1}{d_2-d(t)} \int_{t-d_2}^{t-d(t)} x^T(s) E^T ds \ \frac{1}{d(t)} \int_{t-d(t)}^t x^T(s) E^T ds \ \frac{1}{d_1} \int_{t-d_1}^t x^T(s) E^T ds \right]^T, \\
 \psi_3(t) &= \left[\frac{1}{(d(t)-d_1)^2} \int_{t-d(t)}^{t-d_1} \int_{\theta}^t x^T(s) E^T ds d\theta \ \frac{1}{(d_2-d(t))^2} \int_{t-d_2}^{t-d(t)} \int_{\theta}^t x^T(s) E^T ds d\theta \right. \\
 &\quad \left. \frac{1}{(d(t))^2} \int_{t-d(t)}^t \int_{\theta}^t x^T(s) E^T ds d\theta \ \frac{1}{d_1^2} \int_{t-d_1}^t \int_{\theta}^t x^T(s) E^T ds \right]^T.
 \end{aligned}$$

由[8]中引理 2, 4 和假设 1, 2

$$\begin{aligned}
 2x^T(t) P_i \Delta A_i x(t) &= 2x^T(t) P_i M_i F_i(t) N_i x(t) \leq \lambda_{i1} x^T(t) P_i M_i M_i^T P_i^T x(t) + \lambda_{i1}^{-1} x^T(t) N_i^T N_i x(t), \\
 2x^T(t) P_i \Delta A_{di} x(t-d(t)) &= 2x^T(t) P M_i F_i(t) N_{di} x(t-d(t)) \\
 &\leq \lambda_{i2} x^T(t) P_i M_i M_i^T P_i^T x(t-d(t)) + \lambda_{i2}^{-1} x^T(t) N_{di}^T N_{di} x(t-d(t)), \\
 2x^T(t) P_i B_i \Delta K_i x(t) &= 2x^T(t) P_i B_i M_i F_i(t) N_{ki} x(t) \leq \lambda_{i3} x^T(t) P_i B_i M_i M_i^T P_i^T B_i^T x(t) + \lambda_{i3}^{-1} x^T(t) N_{ki}^T N_{ki} x(t), \\
 2x^T(t) P_i B_i \Delta K_{kdi} x(t-d(t)) &\leq \lambda_{i4} x^T(t) P_i B_i M_i M_i^T P_i^T B_i^T x(t-d(t)) + \lambda_{i4}^{-1} x^T(t) N_{kdi}^T N_{kdi} x(t-d(t)), \\
 2x^T(t) P_i B_i f(t) &\leq \lambda_{i5} x^T(t) P_i B_i B_i^T P_i^T x(t) + \lambda_{i5}^{-1} f^T(t) f(t) \leq \lambda_{i5} x^T(t) P_i B_i B_i^T P_i^T x(t) + \lambda_{i5}^{-1} x^T(t) T^2 x(t), \\
 2x^T(t) P_i B_{di} f(t-d(t)) &\leq \lambda_{i6} x^T(t) P_i B_{di} B_{di}^T P_i^T x(t) + \lambda_{i6}^{-1} f^T(t-d(t)) f(t-d(t)) \\
 &\leq \lambda_{i6} x^T(t) P_i B_{di} B_{di}^T P_i^T x(t) + \lambda_{i6}^{-1} x^T(t-d(t)) T^2 x(t-d(t)). \\
 LV_1(x_t) &\leq \Psi^T(t) \left[\text{sym}(\Lambda_1^T P_i \Lambda_2) + \Lambda_1^T \left(\sum_{j=1}^N \pi_{ij} P_j \right) \Lambda_1 \right] \Psi(t) + x^T(t) \left[\sum_{j=1}^N \pi_{ij} P_j + P_i (2A_i + 2B_i K + \lambda_{i1}^{-1} N_i^T N_i \right. \\
 &\quad \left. + \lambda_{i3}^{-1} N_{ki}^T N_{ki} + \lambda_{i1} M_i M_i^T P_i^T + \lambda_{i3} B_i M_i M_i^T P_i^T B_i^T + \lambda_{i6} B_{di} B_{di}^T P_i^T) + \lambda_{i5}^{-1} T^2 \right] x(t) \\
 &\quad + x^T(t) P_i (2A_{di} + 2B_{di} K_{di} + \lambda_{i2}^{-1} N_{di}^T N_{di} + \lambda_{i4}^{-1} N_{kdi}^T N_{kdi} + \lambda_{i2} M_i M_i^T P_i^T \\
 &\quad + \lambda_{i4} B_i M_i M_i^T P_i^T B_i^T) x(t-d(t)) + \lambda_{i6}^{-1} x^T(t-d(t)) T^2 x(t-d(t)) \\
 &= \Psi^T(t) \left[\text{sym}(\Lambda_1^T P_i \Lambda_2) + \Lambda_1^T \left(\sum_{j=1}^N \pi_{ij} P_j \right) \Lambda_1 + e_1^T \Theta_1 e_1 + e_1^T P_i \Theta_2 e_2 + e_2^T \lambda_{i6}^{-1} T^2 e_2 \right] \Psi(t),
 \end{aligned}$$

$$\Theta_1 = \sum_{j=1}^N \pi_{ij} P_j + P_i \left(2A_i + 2B_i K + \lambda_{i1}^{-1} N_i^T N_i + \lambda_{i3}^{-1} N_{ki}^T N_{ki} + \lambda_{i1} M_i M_i^T P_i^T + \lambda_{i3} B_i M_i M_i^T P_i^T B_i^T + \lambda_{i6} B_{di} B_{di}^T P_i^T \right) + \lambda_{i5}^{-1} T^2,$$

$$\Theta_2 = 2A_{di} + 2B_{di} K_{di} + \lambda_{i2}^{-1} N_{di}^T N_{di} + \lambda_{i4}^{-1} N_{kdi}^T N_{kdi} + \lambda_{i2} M_i M_i^T P_i^T + \lambda_{i4} B_i M_i M_i^T P_i^T B_i^T,$$

[9]结论 1 和[10]引理 3, 有

$$\begin{aligned} & -d_2 \int_{t-d_2}^t E^T \dot{x}^T(s) S_1 E \dot{x}(s) ds = -d_2 \int_{t-d(t)}^t E^T \dot{x}^T(s) S_1 E \dot{x}(s) ds - d_2 \int_{t-d_2}^{t-d(t)} E^T \dot{x}^T(s) S_2 E \dot{x}(s) ds \\ & \leq \Psi^T(t) \left[-\frac{1}{\alpha_1} (r_1^T S_1 r_1 + 3r_2^T S_1 r_2 + 5r_3^T S_1 r_3) - \frac{1}{1-\alpha_1} (r_4^T S_1 r_4 + 3r_5^T S_1 r_5 + 5r_6^T S_1 r_6) \right] \Psi(t) \\ & \leq \Psi^T(t) \Gamma_1^T \left[\begin{array}{c} \bar{S}_1 + (1-\alpha_1)(\bar{S}_1 + \varepsilon_1 I - Y_1 \bar{S}_1^{-1} Y_1^T) \\ Y_1 \\ \bar{S}_1 + \alpha_1 (\bar{S}_1 + \varepsilon_1 I - Y_1 \bar{S}_1^{-1} Y_1) \end{array} \right] \Gamma_1 \Psi(t) \\ & = \Psi^T(t) (\mu_1 - \Gamma_1^T \Phi_1 \Gamma_1) \Psi^T(t) \end{aligned}$$

$$-d_{12} \int_{t-d_2}^{t-d_1} E \dot{x}^T(s) S_2 E \dot{x}(s) ds \leq \Psi^T(t) (\mu_2 - \Gamma_2^T \Phi_2 \Gamma_2) \Psi^T(t)$$

$$\mu_1 = (1-\alpha_1) \Lambda_3^T Y_1 \bar{S}_1^{-1} Y_1 \Lambda_3 + \alpha_1 \Lambda_4^T Y_1 \bar{S}_1^{-1} Y_1 \Lambda_4,$$

$$\mu_2 = (1-\alpha_2) \Lambda_5^T Y_2 \bar{S}_2^{-1} Y_2 \Lambda_5 + \alpha_2 \Lambda_6^T Y_2 \bar{S}_2^{-1} Y_2 \Lambda_6, \bar{S}_n = \text{diag} \{S_n, 3S_n, 5S_n\},$$

根据[11]中(9), (10), 我们得到

$$\begin{aligned} & -\int_{t-d_1}^t \int_{\theta}^t E^T \dot{x}^T(s) Z_1 E \dot{x}(s) ds d\theta \leq -\Psi^T(t) [2r_{10}^T Z_1 r_{10} + 4r_{11}^T Z_1 r_{11}] \Psi(t) = \Psi^T(t) [-(\Lambda_6^T \bar{Z}_1 \Lambda_6)] \Psi(t) \\ & -\int_{t-d_1}^t \int_{t-d_1}^{\theta} E^T \dot{x}^T(s) Z_2 E \dot{x}(s) ds d\theta \leq -\Psi^T(t) [2r_{12}^T Z_2 r_{12} + 4r_{13}^T Z_2 r_{13}] \Psi(t) = \Psi^T(t) [-(\Lambda_7^T \bar{Z}_2 \Lambda_7)] \Psi(t). \end{aligned}$$

通过上述, 我们可以得到

$$LV(x_t) \leq \xi^T(t) \Omega_0 \xi(t),$$

其中 $\Omega_0 = E_{i\mu} |_{d(t), \dot{d}(t)} + E_2 + \eta_1 + \eta_2$, 由 Schur 补, 我们有,

$$\Omega_0 = E_{i\mu} |_{d(t), \dot{d}(t)} + E_2 + \eta_1 + \eta_2 < 0,$$

$$LV(x_t) < 0,$$

这意味着 $LV(x_t) \leq -\zeta x^T(t) x(t)$, 使用 Dynkin 公式, 所有 $i \in I, T > 0$, 遵循 $i \in I, T > 0$,

$$E \{V(x(T), r(T))\} - V(x(0), r(0)) \leq -\zeta E \left\{ \int_0^T x^T(s) x(s) | (x(0), r(0)) \right\},$$

此外, 让 $T \rightarrow \infty, E \left\{ \int_0^T x^T(s) x(s) | (x(0), r(0)) \right\} \leq \frac{1}{\zeta} V(x(0), r(0)) < \infty$.

根据定义 2.1, 系统(2.2)是随机稳定的。接下来我们考虑 H_∞ 性能函数 J ,

$$\begin{aligned} J & = \int_0^\infty z^T(t) z(t) - \gamma^2 w^T(t) w(t) dt \leq \int_0^\infty LV(x_t) + z^T(t) z(t) - \gamma^2 w^T(t) w(t) dt \\ & = \int_0^\infty \Psi^T(t) \Omega \Psi(t) dt, \end{aligned}$$

我们有 $J < 0$. H_∞ 性能已验证。根据 GUTR 矩阵的定义, 我们要探讨以下三种情况下的上述不等式。

情况 I $i \notin U_k^i, U_k^i = \{k_1^i, \dots, k_{m_i}^i\}$, 存在正定矩阵, $V_{ij} \in R^{n \times n} (i \notin U_k^i, j \in U_k^i)$ 及 $l \in U_{ik}^i$ 有

$$\begin{bmatrix} E^{[m]} & E_{12} & e_{j\mu}^T & \Pi_1^T E(P_{k_l^i} - F_i) & \cdots & \Pi_1^T E(P_{ik_{mi}^i} - F_i) \\ * & E_{22} & 0 & 0 & \cdots & 0 \\ * & * & -I & 0 & \cdots & 0 \\ * & * & * & -V_{ik_l^i} & \cdots & 0 \\ * & * & * & * & \ddots & \vdots \\ * & * & * & * & * & -V_{ik_{mi}^i} \end{bmatrix} < 0,$$

证明 $i \notin U_k^i$, 应该注意的是, 在这种情况下 $\sum_{j \in U_k^i, j \neq i} \pi_{ij} = -\pi_{ii} - \sum_{j \in U_k^i} \pi_{ij}$, $\pi_{ij} \geq 0$, 且必须有 $l \in U_k^i$, $l \neq j$,

使 $E^T \bar{P}_i E - E^T \bar{P}_j E \geq 0$, 定义

$$\begin{aligned} \Omega = & \Xi + E_2 + E_3 + \Lambda_1^T \left(\sum_{j \in U_k^i} \pi_{ij} P_j + \pi_{ii} P_i + \sum_{j \in U_k^i, j \neq i} \pi_{ij} P_j \right) \Lambda_1 \\ & + x^T(t) E^T \left(\sum_{j \in U_k^i} \pi_{ij} P_j + \pi_{ii} P_i + \sum_{j \in U_k^i, j \neq i} \pi_{ij} P_j \right) x(t), \end{aligned} \tag{2.6}$$

$$\begin{aligned} \Omega = & \Xi + E_2 + E_3 + \Lambda_1^T \sum_{j \in U_k^i} \pi_{ij} P_j \Lambda_1 + \Lambda_1^T \pi_{ii} P_i \Lambda_1 + \Lambda_1^T \sum_{j \in U_k^i, j \neq i} \pi_{ij} P_j \Lambda_1 + x^T(t) E^T \sum_{j \in U_k^i} \pi_{ij} P_j x(t) \\ & + x^T(t) E^T \pi_{ii} P_i x(t) + x^T(t) E^T \sum_{j \in U_k^i, j \neq i} \pi_{ij} P_j x(t) \\ \leq & \Xi + E_2 + E_3 + \Lambda_1^T \sum_{j \in U_k^i} \pi_{ij} P_j \Lambda_1 + \Lambda_1^T \pi_{ii} P_i \Lambda_1 + \Lambda_1^T \left(-\pi_{ii} - \sum_{j \in U_k^i} \pi_{ij} \right) P_i \Lambda_1 + x^T(t) E^T \sum_{j \in U_k^i} \pi_{ij} P_j x(t) \\ & + x^T(t) E^T \pi_{ii} P_i x(t) + x^T(t) E^T \left(-\pi_{ii} - \sum_{j \in U_k^i} \pi_{ij} \right) P_i x(t) \\ = & \Xi + E_2 + E_3 + \sum_{j \in U_k^i} \pi_{ij} \Lambda_1^T (P_j - P_i) \Lambda_1 + \sum_{j \in U_k^i} \pi_{ij} x^T(t) E^T (P_j - P_i) x(t) \\ = & \Xi + E_2 + E_3 + \sum_{j \in U_k^i} (\hat{\pi}_{ij} + \pi_{ij}) \Lambda_1^T (P_j - P_i) \Lambda_1 + \sum_{j \in U_k^i} (\hat{\pi}_{ij} + \pi_{ij}) x^T(t) E^T (P_j - P_i) x(t) \\ = & \Xi + E_2 + E_3 + \sum_{j \in U_k^i} \hat{\pi}_{ij} \Lambda_1^T (P_j - P_i) \Lambda_1 + \sum_{j \in U_k^i} \pi_{ij} \Lambda_1^T (P_j - P_i) \Lambda_1 + \sum_{j \in U_k^i} \hat{\pi}_{ij} x^T(t) E^T (P_j - P_i) x(t) \\ & + \sum_{j \in U_k^i} \pi_{ij} x^T(t) E^T (P_j - P_i) x(t). \end{aligned} \tag{2.7}$$

此外, 由于引理 2.1, 可以得到

$$\begin{aligned} & \sum_{j \in U_k^i} \pi_{ij} \Lambda_1^T (P_j - P_i) \Lambda_1 + \sum_{j \in U_k^i} \pi_{ij} E^T x^T(t) (P_j - P_i) x(t) \\ = & \sum_{j \in U_k^i} \left[\frac{1}{2} \pi_{ij} \Lambda_1^T (P_j - P_i) \Lambda_1 + \frac{1}{2} \pi_{ij} \Lambda_1^T (P_j - P_i) \Lambda_1 \right] \\ & + \sum_{j \in U_k^i} \left[\frac{1}{2} \pi_{ij} E^T x^T(t) (P_j - P_i) x(t) + \frac{1}{2} \pi_{ij} E^T x^T(t) (P_j - P_i) x(t) \right] \\ \leq & \sum_{j \in U_k^i} \left[\frac{\delta_{ij1}^2}{4} V_{ij1} + \Lambda_1^T (P_j - P_i) \Lambda_1 V_{ij1}^{-1} \Lambda_1^T (P_j - P_i) \Lambda_1 \right] \\ & + \sum_{j \in U_k^i} \left[\frac{\delta_{ij2}^2}{4} V_{ij2} + E^T x^T(t) (P_j - P_i) x(t) V_{ij2}^{-1} E^T x^T(t) (P_j - P_i) x(t) \right], \end{aligned} \tag{2.8}$$

从(2.6)~(2.8)中, 我们有

$$\begin{aligned} \Omega \leq & \Xi + E_2 + \sum_{j \in U_k^i} \hat{\pi}_{ij} \Lambda_1^T (P_j - P_l) \Lambda_1 + \sum_{j \in U_k^i} \frac{\delta_{ij1}^2}{4} V_{ij1} + \sum_{j \in U_k^i} \left[\Lambda_1^T (P_j - P_l) \Lambda_1 V_{ij1}^{-1} \Lambda_1^T (P_j - P_l) \Lambda_1 \right] \\ & + \sum_{j \in U_k^i} \frac{\delta_{ij2}^2}{4} V_{ij2} + \sum_{j \in U_k^i} \left[E^T x^T(t) (P_j - P_l) x(t) V_{ij2}^{-1} E^T x^T(t) (P_j - P_l) x(t) \right] \\ & + \sum_{j \in U_k^i} \hat{\pi}_{ij} E^T x^T(t) (P_j - P_l) x(t) \end{aligned} \quad (2.9)$$

如果,

$$\begin{aligned} \Xi + E_2 + \sum_{j \in U_k^i} \hat{\pi}_{ij} \Lambda_1^T (P_j - P_l) \Lambda_1 + \sum_{j \in U_k^i} \frac{\delta_{ij1}^2}{4} V_{ij1} + \sum_{j \in U_k^i} \left[\Lambda_1^T (P_j - P_l) \Lambda_1 V_{ij1}^{-1} \Lambda_1^T (P_j - P_l) \Lambda_1 \right] + \sum_{j \in U_k^i} \frac{\delta_{ij2}^2}{4} V_{ij2} \\ + \sum_{j \in U_k^i} \left[E^T x^T(t) (P_j - P_l) x(t) V_{ij2}^{-1} E^T x^T(t) (P_j - P_l) x(t) \right] + \sum_{j \in U_k^i} \hat{\pi}_{ij} E^T x^T(t) (P_j - P_l) x(t) < 0 \end{aligned} \quad (2.10)$$

则 $\Omega < 0$, 由引理 2.2 有(2.10)成立。可以看出

$$LV(x_t) \leq \gamma^2 w^T(t) w(t) - z^T(t) z(t).$$

由定义 2.2, 系统(2.2)在 H_∞ 扰动水平为零时, 是随机允许的 γ 。

情况 II $i \in U_k^i$, $U_k^i \neq \emptyset$, 存在正定矩阵, $G_{ij} \in R^{n \times n}$ ($i, j \in U_k^i, l \in U_{ik}^i$) 有

$$\begin{bmatrix} E^{[m]} & E_{12} & e_{21}^T & \Pi_1^T E(P_{k_1^i} - F_i) & \cdots & \Pi_1^T E(P_{ik_{mi}^i} - F_i) \\ * & E_{22} & 0 & 0 & \cdots & 0 \\ * & * & -I & 0 & \cdots & 0 \\ * & * & * & -G_{ik_1^i} & \cdots & 0 \\ * & * & * & * & \ddots & \vdots \\ * & * & * & * & * & -G_{ik_{mi}^i} \end{bmatrix} < 0,$$

证明 $i \in U_k^i$ 且 $U_{ik}^i \neq \emptyset$, 必须有 $l \in U_{ik}^i$, $E^T P(l) E - E^T P(j) E \geq 0$, $j \in U_{ik}^i$ 。因为它认为

$$\Omega = \Xi + E_2 + E_3 + \Lambda_1^T \left(\sum_{j \in U_k^i} \pi_{ij} P_j + \pi_{ii} P_i + \sum_{j \in U_{ik}^i} \pi_{ij} P_j \right) \Lambda_1 + x^T(t) E^T \left(\sum_{j \in U_k^i} \pi_{ij} P_j + \pi_{ii} P_i + \sum_{j \in U_{ik}^i} \pi_{ij} P_j \right) x(t),$$

然后我们有

$$\begin{aligned} \Omega \leq & \Xi + E_2 + E_3 + \Lambda_1^T \left(\sum_{j \in U_k^i} \pi_{ij} P_j + \sum_{j \in U_{ik}^i} \pi_{ij} P_l \right) \Lambda_1 + E^T x^T(t) \left(\sum_{j \in U_k^i} \pi_{ij} P_j + \sum_{j \in U_{ik}^i} \pi_{ij} P_l \right) x(t) \\ = & \Xi + E_2 + E_3 + \sum_{j \in U_k^i} \pi_{ij} \Lambda_1^T P_j \Lambda_1 - \sum_{j \in U_{ik}^i} \pi_{ij} \Lambda_1^T P_l \Lambda_1 + \sum_{j \in U_k^i} \pi_{ij} x^T(t) E^T P_j x(t) - \sum_{j \in U_{ik}^i} \pi_{ij} x^T(t) E^T P_l x(t) \\ = & \Xi + E_2 + E_3 + \sum_{j \in U_k^i} (\hat{\pi}_{ij} + \pi_{ij}) \Lambda_1^T (P_j - P_l) \Lambda_1 + \sum_{j \in U_k^i} (\hat{\pi}_{ij} + \pi_{ij}) x^T(t) E^T (P_j - P_l) x(t) \\ = & \Xi + E_2 + E_3 + \sum_{j \in U_k^i} \hat{\pi}_{ij} \Lambda_1^T (P_j - P_l) \Lambda_1 + \sum_{j \in U_k^i} \pi_{ij} \Lambda_1^T (P_j - P_l) \Lambda_1 + \sum_{j \in U_k^i} \hat{\pi}_{ij} x^T(t) E^T (P_j - P_l) x(t) \\ & + \sum_{j \in U_k^i} \pi_{ij} x^T(t) E^T (P_j - P_l) x(t), \end{aligned}$$

此外, 由于引理 2.2, 我们可以得到

$$\begin{aligned}
 & \sum_{j \in U_k^i} \pi_{ij} \Lambda_1^T (P_j - P_l) \Lambda_1 + \sum_{j \in U_k^i} \pi_{ij} E^T x^T(t) (P_j - P_l) x(t) \\
 &= \sum_{j \in U_k^i} \left[\frac{1}{2} \pi_{ij} \Lambda_1^T (P_j - P_l) \Lambda_1 + \frac{1}{2} \pi_{ij} \Lambda_1^T (P_j - P_l) \Lambda_1 \right] \\
 & \quad + \sum_{j \in U_k^i} \left[\frac{1}{2} \pi_{ij} E^T x^T(t) (P_j - P_l) x(t) + \frac{1}{2} \pi_{ij} E^T x^T(t) (P_j - P_l) x(t) \right] \\
 & \leq \sum_{j \in U_k^i} \left[\frac{\delta_{ij1}^2}{4} G_{ij1} + \Lambda_1^T (P_j - P_l) \Lambda_1 G_{ij1}^{-1} \Lambda_1^T (P_j - P_l) \Lambda_1 \right] \\
 & \quad + \sum_{j \in U_k^i} \left[\frac{\delta_{ij2}^2}{4} G_{ij2} + E^T x^T(t) (P_j - P_l) x(t) G_{ij2}^{-1} E^T x^T(t) (P_j - P_l) x(t) \right],
 \end{aligned} \tag{2.11}$$

另外, 我们有

$$\begin{aligned}
 \Omega \leq & \Xi + E_2 + E_3 + \sum_{j \in U_k^i} \hat{\pi}_{ij} \Lambda_1^T (P_j - P_l) \Lambda_1 + \sum_{j \in U_k^i} \frac{\delta_{ij1}^2}{4} G_{ij1} \\
 & + \sum_{j \in U_k^i} \left[\Lambda_1^T (P_j - P_l) \Lambda_1 G_{ij1}^{-1} \Lambda_1^T (P_j - P_l) \Lambda_1 \right] \\
 & + \sum_{j \in U_k^i} \hat{\pi}_{ij} E^T x^T(t) (P_j - P_l) x(t) + \sum_{j \in U_k^i} \frac{\delta_{ij2}^2}{4} G_{ij2} \\
 & + \sum_{j \in U_k^i} \left[E^T x^T(t) (P_j - P_l) x(t) G_{ij2}^{-1} E^T x^T(t) (P_j - P_l) x(t) \right],
 \end{aligned} \tag{2.12}$$

如果

$$\begin{aligned}
 & \Xi + E_2 + E_3 + \sum_{j \in U_k^i} \hat{\pi}_{ij} \Lambda_1^T (P_j - P_l) \Lambda_1 + \sum_{j \in U_k^i} \left[\Lambda_1^T (P_j - P_l) E^T \Lambda_1 G_{ij1}^{-1} \Lambda_1^T (P_j - P_l) \Lambda_1 \right] \\
 & + \sum_{j \in U_k^i} \frac{\delta_{ij1}^2}{4} G_{ij1} + \sum_{j \in U_k^i} \left[E^T x^T(t) (P_j - P_l) x(t) G_{ij2}^{-1} E^T x^T(t) (P_j - P_l) x(t) \right] \\
 & + \sum_{j \in U_k^i} \frac{\delta_{ij2}^2}{4} G_{ij2} + \sum_{j \in U_k^i} \hat{\pi}_{ij} E^T x^T(t) (P_j - P_l) x(t) < 0,
 \end{aligned} \tag{2.13}$$

因此 $\Omega < 0$, 由引理 2.2 有(2.13)成立。可以看出 $LV(x_t) \leq \gamma^2 w^T(t) w(t) - z^T(t) z(t)$ 。由定义 2.2, 系统(2.2)在 H_∞ 扰动水平为零时是随机允许的 γ 。

情况 III $i \in U_k^i, U_{uk}^i = \emptyset$, 存在正定矩阵, $L_{ij} \in R^{n \times n} (i, j \in U_k^i)$ 有

$$\begin{bmatrix}
 E^{[m]} & E_{12} & e_{\sigma\mu}^T & \Pi_1^T E \begin{pmatrix} P_{k_1^i} \\ -F_i \end{pmatrix} & \cdots & \Pi_1^T E \begin{pmatrix} P_{ik_{mi}^i} \\ -F_i \end{pmatrix} \\
 * & E_{22} & 0 & 0 & \cdots & 0 \\
 * & * & -I & 0 & \cdots & 0 \\
 * & * & * & -L_{ik_1^i} & \cdots & 0 \\
 * & * & * & * & \ddots & \vdots \\
 * & * & * & * & * & -L_{ik_{mi}^i}
 \end{bmatrix} < 0,$$

证明 $i \in U_k^i, U_{uk}^i = \emptyset$ 。同样, 它认为

$$\begin{aligned} \Omega &= \Xi + E_2 + E_3 + \Lambda_1^T \sum_{j=1, j \neq i}^s \pi_{ij} P_j \Lambda_1 + \Lambda_1^T \pi_{ii} P_i \Lambda_1 + x^T(t) E^T \sum_{j=1, j \neq i}^s \pi_{ij} P_j x(t) + x^T(t) E^T \pi_{ii} P_i x(t) \\ &= \Xi + E_2 + E_3 + \Lambda_1^T \sum_{j=1, j \neq i}^s \pi_{ij} (P_j - P_i) \Lambda_1 + x^T(t) E^T \sum_{j=1, j \neq i}^s \pi_{ij} (P_j - P_i) x(t) \\ &= \Xi + E_2 + E_3 + \Lambda_1^T \sum_{j=1, j \neq i}^s (\hat{\pi}_{ij} + \pi_{ij}) (P_j - P_i) \Lambda_1 + x^T(t) E^T \sum_{j=1, j \neq i}^s (\hat{\pi}_{ij} + \pi_{ij}) (P_j - P_i) x(t) \\ &\leq \Xi + E_2 + E_3 + \sum_{j \in U_k^i} \hat{\pi}_{ij} \Lambda_1^T (P_j - P_i) \Lambda_1 + \sum_{j \in U_k^i} \frac{\delta_{ij1}^2}{4} L_{ij1} + \sum_{j \in U_k^i} \left[\Lambda_1^T (P_j - P_i) \Lambda_1 L_{ij1}^{-1} \Lambda_1^T (P_j - P_i) \Lambda_1 \right] \\ &\quad + \sum_{j \in U_k^i} \left[E^T x^T(t) (P_j - P_i) x(t) L_{ij2}^{-1} E^T x^T(t) (P_j - P_i) x(t) \right] + \sum_{j \in U_k^i} \hat{\pi}_{ij} E^T x^T(t) (P_j - P_i) x(t) + \sum_{j \in U_k^i} \frac{\delta_{ij2}^2}{4} L_{ij2}, \end{aligned}$$

如果

$$\begin{aligned} &\Xi + E_2 + E_3 + \sum_{j \in U_k^i} \hat{\pi}_{ij} \Lambda_1^T (P_j - P_i) \Lambda_1 + \sum_{j \in U_k^i} \frac{\delta_{ij1}^2}{4} L_{ij1} + \sum_{j \in U_k^i} \hat{\pi}_{ij} E^T x^T(t) (P_j - P_i) x(t) + \sum_{j \in U_k^i} \frac{\delta_{ij2}^2}{4} L_{ij2} \\ &+ \sum_{j \in U_k^i} \left[\Lambda_1^T (P_j - P_i) \Lambda_1 L_{ij1}^{-1} \Lambda_1^T (P_j - P_i) \Lambda_1 \right] + \sum_{j \in U_k^i} \left[E^T x^T(t) (P_j - P_i) x(t) L_{ij2}^{-1} E^T x^T(t) (P_j - P_i) x(t) \right] < 0 \end{aligned} \tag{2.14}$$

则 $\Omega < 0$, 引理 2.2, (2.14)成立可以看出

$$LV(x_t) \leq \gamma^2 w^T(t) w(t) - z^T(t) z(t).$$

由定义 2.2 可知, 系统(2.2)在扰动水平为 H_∞ 时为零时是随机允许的 γ . 考虑具有恒定延迟 h_2 的 SMJs, 其中的参数与[8]中的参数相同, 数值示例说明本文比以前方法有效更通用见表 1:

Table 1. Upper bound contrast
表 1. 上界对比

比较结果	d
[8]	1.2362
本文 $\varepsilon = 0$	1.4731
本文 $\varepsilon = 2$	1.5836

$$\begin{aligned} E &= \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, A = \begin{bmatrix} 0.4972 & 0 \\ 0 & -0.9541 \end{bmatrix}, A_d = \begin{bmatrix} -1.4972 & 1.5415 \\ 0 & 0.5449 \end{bmatrix}, \\ M &= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, N = \begin{bmatrix} 2 & 1 \\ -1 & -1 \end{bmatrix}, MEN = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \end{aligned}$$

此外, 我们选择

$$\begin{aligned} \Sigma &= \begin{bmatrix} 0.1 & 0.4 \\ 0.2 & 0.5 \end{bmatrix}, B = \begin{bmatrix} -1 \\ 0 \end{bmatrix}, B_d = \begin{bmatrix} -2 \\ 1 \end{bmatrix}, B_w = \begin{bmatrix} 1 \\ 2.5 \end{bmatrix}, C = [1 \ 0], \\ C_d &= [0.3 \ -0.3], D_w = \begin{bmatrix} 1 & 1.5 \\ 3 & 2 \end{bmatrix}, H_R = 4, \gamma = 0.1. \end{aligned}$$

为了验证所提出的方法绘制了 $x(0) = [0 \ 0]^T$ 和 $w(t) = [1 \ 0]^T \sin e^{-0.2t}$ 可以看出 SNJs 状态响应收敛于零见图 1.

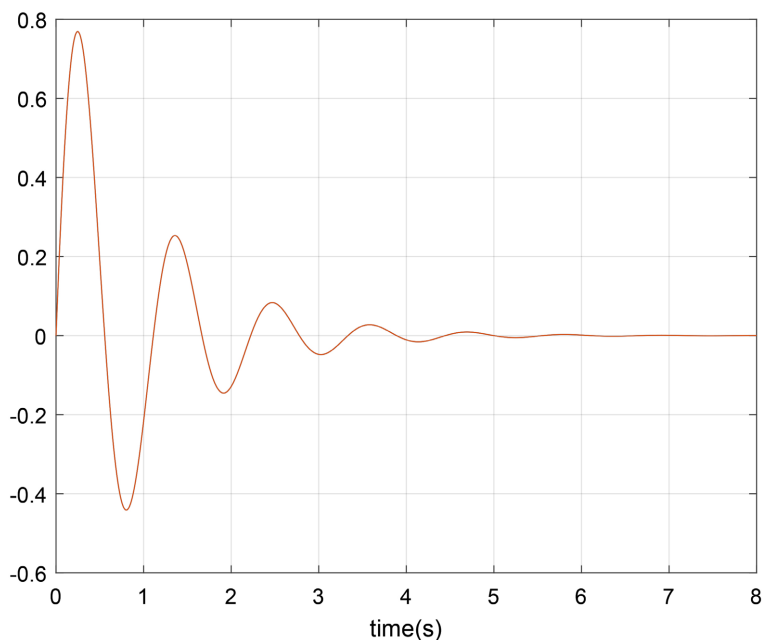


Figure 1. $x(t)$ status response with external input $w(t)$

图 1. 具有外部输入 $w(t)$ 的 $x(t)$ 状态响应

备注 2.2: 与一般的 LKF 相比, 乘以更多的积分项, 利用了更多的时变延迟信息, 在增强的 LKF 中减少保守性。(2.21)和(2.22)中的条件可以通过使用[10]引理 3 得到更紧的界限。因此, 本文定理的保守性比 Wang 等[12]、Fu 和 Ma [13]更低。

3. 结论

对于给定的标量本章探讨一类非线性广义 Markov 跳跃系统的随机稳定性和 H_∞ 控制问题。设计新型的 Lyapunov-Krasovkii 泛函(LKF), 构造适当的 Lyapunov-Krasovskii 泛函, 应用依赖参数相关的互复凸矩阵不等式(PDRCMI)、改进的 Wirtinger 不等式, Chen 提出的广义积分不等式, 通过 LMI, 得到保守性较低的准则。同时, 引入参数依赖的互凸矩阵不等式(PDRCMI)来降低保守性, 保证了非线性广义 Markov 跳跃系统渐进稳定并满足性能。此外, 将自适应方法应用于 MJSs 具有现实意义, 值得进一步探索。值得一提的是, 我们的理论结果也可以在未来更复杂的系统。

参考文献

- [1] Boukas, E.-K. (2005) Stochastic Switching Systems: Analysis and Design. Birkhauser.
- [2] Dai, L. (1989) Singular Control Systems. Springer Verlag. <https://doi.org/10.1007/BFb0002475>
- [3] Lu, R.Q., Su, H.Y., Xue, A.K. and Chu, J. (2008) Robust Control Theory of Singular Systems. Science Press.
- [4] Shi, P., Boukas, E.-K. and Agarwal, R.K. (2000) Robust H_∞ Control of Singular Continuous-Time Systems with Delays and Uncertainties. *Proceedings of the 39th Conference on Decision and Control*, Sidney, 12-15 December 2000, 1515-1520.
- [5] Xu, S. and Lam, J. (2006) Robust Control and Filtering of Singular Systems. Springer Verlag.
- [6] 郑成德, 肖岩, 贾贺贺. 中立型 Markov 脉冲神经网络的随机稳定性[J]. 大连交通大学学报, 2019, 40(4): 116-120.
- [7] Zong, G. and Ren, H. (2019) Guaranteed Cost Finite-Time Control for Semi-Markov Jump Systems with Event-Triggered Scheme and Quantization Input. *International Journal of Robust and Nonlinear Control*, **29**, 5251-5273. <https://doi.org/10.1002/rnc.4672>

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- [8] Ma, Y.C., Chen, M.H. and Zhang, Q.L. (2015) Memory Dissipative Control for Singular T-S Fuzzy Time-Varying Delay Systems under Actuator Saturation. *Journal of the Franklin Institute*, **352**, 3947-3970. <https://doi.org/10.1016/j.jfranklin.2015.05.030>
- [9] Kim, J.-H. (2016) Further Improvement of Jensen Inequality and Application to Stability of Time-Delayed Systems. *Automatica*, **64**, 121-125. <https://doi.org/10.1016/j.automatica.2015.08.025>
- [10] Tian, Y.F., Wang, Y.Z. and Ren, J.C. (2020) Stability Analysis and Control Design of Singular Markovian Jump Systems via a Parameter-Dependent Reciprocally Convex Matrix Inequality. *Applied Mathematics and Computation*, **386**, Article 125471. <https://doi.org/10.1016/j.amc.2020.125471>
- [11] Chen, J., Xu, S.Y., Chen, W.M., Zhang, B.Y., Ma, Q. and Zou, Y. (2016) Two General Integral Inequalities and Their Applications to Stability Analysis for Systems with Time-Varying Delay. *International Journal of Robust and Nonlinear Control*, **26**, 4088-4103. <https://doi.org/10.1002/rnc.3551>
- [12] Wang, J.R., Wang, H.J., Xue, A.K. and Lu, R.Q. (2013) Delay-Dependent H_∞ Control for Singular Markovian Jump Systems with Time Delay. *Nonlinear Analysis: Hybrid Systems*, **8**, 1-12. <https://doi.org/10.1016/j.nahs.2012.08.003>
- [13] Fu, L. and Ma, Y.C. (2018) Dissipative Filtering for Singular Markov Jump Systems with Generally Uncertain Transition Rates via New Integral Inequality Approach. *Journal of the Franklin Institute*, **355**, 7354-7383. <https://doi.org/10.1016/j.jfranklin.2018.07.023>