New Result on Delay-dependent Stability for Markovian Jump Time-delay Systems With Partial Information on Transition Probabilities

Yan Zhang, Ke Lou, and Yuan Ge

Abstract—This paper focuses on the delay-dependent stability for a kind of Markovian jump time-delay systems (MJTDSs), whose transition rates are incompletely known. In order to reduce the computational complexity and achieve better performance, auxiliary function-based double integral inequality is combined with extended Wirtinger's inequality and Jensen inequality to deal with the double integral and the triple integral in augmented Lyapunov-Krasovskii function (ALKF) and their weak infinitesimal generator respectively, the more accurate approximation bounds with a fewer variables are derived. As a result, less conservative stability criteria are proposed in this paper. Finally, numerical examples are given to show the effectiveness and the merits of the proposed method.

Index Terms—Auxiliary function-based double integral inequality, delay-dependent stability, Markovian jump time-delay systems (MJTDSs), unknown transition rates.

I. INTRODUCTION

MARKOVIAN jump systems (MJSs) play an important role in hybrid systems because they are able to more accurately reflect the dynamic behavior in various stages of development than a single system, such as different elasticity of demand in power system, the flight-attitude-adjustment in flight control system, maneuvering target tracking [1], random variation in network environment and so on.

Taking into account that the transition probabilities to form the Markov chain are not exactly known in practice, Zhang *et al.* proposed the concept of partly unknown transition probabilities for MJSs in [2], thereafter MJSs with partly unknown transition probabilities have attracted much attention. For example, references [2]−[4] investigated how to separate the unknown transition probabilities or transition rates, and fixed-connection weighting matrix method and free-connection weighting matrix method were proposed in [2] and [3], respectively. Considering that time delay is a widespread phenomenon, it often causes the bad behavior of MJSs, even

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worsens the dynamic properties. The stability analysis and controller synthesis problems for Markovian jump time delay systems (MJTDSs) with partly unknown transition probabilities were studied in [5]−[8], and stochastic synchronization for Markovian coupled neural networks with partial information on transition probabilities was discussed in [9] and so on.

In the field of stability analysis based on Lyapunov-Krasovskii stability theorem, innovation of reducing the conservatism includes two aspects: one is how to obtain tight bounds of integral terms in the derivative of the Lyapunov-Krasovskii function (LKF), which is very important, and various methods were proposed in the past years, for instance, Jensen inequality, free-weighting matrix method, reciprocally convex approach, Wirtinger's inequality, Bessel-Legendre (B-L) inequality and so on, the corresponding results were reported in [10]−[14] respectively. The other is how to construct a novel LKF, for example, triple integral terms were introduced in the LKF to obtain the less conservative result, such as in [15]−[17] etc. But this method would increase the number of decision variables. e.g., [15] introduced more than $130n^2$ scalar variables, which means that the computational complexity is truly very large. Hence, investigating the problem of the tradeoff between the conservatism and the computational complexity becomes important. For example, the delay-dependent stability problem of continuous neural networks with a time-varying delay was investigated in [18] considering both conservativeness and computational complexity. Some new integral inequalities were proposed in [19], [20] to decrease the computational complexity. Especially, auxiliary function-based double integral inequality includes Jensen inequality and Wirtinger-Based integral inequality as special cases, and no additional decision variable is introduced when it is used to estimate the upper bounds of double integral terms.

Inspired by auxiliary function-based double integral inequality, an augmented Lyapunov-Krasovskii function (ALKF) including augmented term and triple integral term is constructed in this paper to investigate the stability of MJTDSs. Auxiliary function-based double integral inequality is combined with extended Wirtinger's inequality and Jensen inequality to deal with the double integral and the triple integral of the ALKF and their weak infinitesimal generator, the more accurate approximation bounds with a fewer variables are derived. Consequently, the improved stability criteria are proposed. Three examples are provided to show the effectiveness and

the merits of our method.

Notation: [7], [10], [21], [22] In this paper, $\mathbb{R}^n$ and $\mathbb{R}^{n\times m}$ denote the n-dimensional Euclidean space and the set of all $n\times m$ real matrices, respectively. The notation $P > 0$ ($P \geq 0$) means that P is real symmetric and positive definite (semi-positive-definite) matrix. The notation ε represents the mathematical expectation, i.e., the expected value of a random variable, which is a terminology in probability theory. $|\cdot|$ denotes the Euclidean norm of a vector. $\mathcal{C}\left([a,b],\mathbb{R}^n\right)$ is a family of continuous functions φ from $[a,b]$ to $\mathbb{R}^n$ with a norm $\|x(t)\|=\sup_{a\leq s\leq b}|x(s)|$. $\mathcal{C}^2_{\mathbb{F}_t}([-h,\ 0],\mathbb{R}^n)$ is a family of $\mathbb{F}_t$-measurable bounded $\mathcal{C}\left([a,b],\mathbb{R}^n\right)$-valued random variables. $\widehat{e}_i \in \mathbb{R}^{n\times 3n}, e_j \in \mathbb{R}^{n\times 7n}, i = 1,2,3, j = 1,2,\cdots,7$, are elementary matrices, i.e., $\widehat{e}_i = \left[\underbrace{0\cdots 0}_{i-1} I \underbrace{0\cdots 0}_{3-i}\right], e_j = \left[\underbrace{0\cdots 0}_{j-1} I \underbrace{0\cdots 0}_{7-j}\right], i = 1,2,3, j = 1,2,\cdots,7$. col$=\{x,y\}$ denotes the column vector of vectors x and y, i.e., col$\{x,y\} = \begin{bmatrix} x \\ y \end{bmatrix}$. For matrix X, Sym$\{X\} = X + X^T$. The symbol $*$ within a matrix represents the symmetric terms of the matrix. Matrices without their dimensions mentioned are assumed to satisfy the requirements of algebraic operations.

II. Problem Formulation

Consider MJSs with time delay described as follows [24]:

$$\dot{x}(t) = A(r_t)x(t) + A_d(r_t)x(t - d(t)) \tag{1}$$

where $x(t) \in \mathbb{R}^n$ is the state vector. $d(t)$ denotes the time-varying delay which satisfies the following constraint conditions

$$0 \leq d(t) \leq h, \quad \mu_1 \leq \dot{d}(t) \leq \mu_2, \quad \forall t \geq 0. \tag{2}$$

When $t \in [-h,\ 0]$, as the initial condition of $x(t)$, $\phi(t)$ belongs to $\mathcal{C}^2_{\mathbb{F}_t}([-h,\ 0],\mathbb{R}^n)$ and r_t belongs to the finite set $\mathcal{S} = \{1,2,\ldots,\mathcal{N}\}$. For simplicity, the solution $x(t,x_0,r_0)$ of system (1) under the initial condition $x(0) = x_0$ and $r_0 \in \mathcal{S}$ is denoted by $x(t)$. $A(r_t)$ and $A_d(r_t)$ are known matrix functions, which depend on a Markov process taking values in the finite set $\mathcal{S}$. The description of transition rate matrix $\Pi = (\pi_{ij}) \in \mathbb{R}^{\mathcal{N}\times\mathcal{N}}$ is given as follows

$$Pr\{r_{t+\delta} = j \mid r_t = i\} = \begin{cases} \pi_{ij}\delta + o(\delta), & i \neq j \\ 1 + \pi_{ij}\delta + o(\delta), & i = j \end{cases}$$

where $\delta > 0, \lim_{\delta \to 0} \frac{o(\delta)}{\delta} = 0$, and $\pi_{ij} \geq 0$ for $i \neq j$ is the transition rate from mode i at time t to mode j at time $t+\delta$, and $\sum_{\delta=1}^{\mathcal{N}} \pi_{ij} = 0$, which means $\pi_{ii} = -\sum_{j=1,i\neq j}^{\mathcal{N}} \pi_{ij}$.

The corresponding transition rate matrix for continuous-time MJSs is defined as

$$\Pi = \begin{bmatrix} \pi_{11} & \pi_{12} & \cdots & ? \\ ? & ? & \cdots & \pi_{2\mathcal{N}} \\ \vdots & \vdots & \ddots & \vdots \\ ? & \pi_{\mathcal{N}2} & \cdots & \pi_{\mathcal{N}\mathcal{N}} \end{bmatrix}. \tag{3}$$

In most cases, the elements of transition rate matrix are not completely known, the description in the form of (3) is much more universal. The symbol "?" in matrix represents the unknown transition rate. For notation clarity, $\forall i \in \mathcal{S}$, the set $U^i = U^i_k \bigcup U^i_{uk}$ with $U^i_k \triangleq \{j : \pi_{ij}$ is known for $j \in \mathcal{S}\}$ and $U^i_{uk} \triangleq \{j : \pi_{ij}$ is unknown for $j \in \mathcal{S}\}$.

Since the solution $\{x(t),t\}$ of (1) is a Markov process, the weak infinitesimal generator operator acting on function V in [23] will be used in next. The definition is

$$\mathcal{L}V(x(t),t,i)$$
$$= \lim_{\Delta \to 0^+} \frac{\mathcal{E}\{V(x(t+\Delta),t+\Delta,r_{t+\Delta})|x(t),r_t=i\} - V(x(t),t,i)}{\Delta}. \tag{4}$$

In addition, the following definition and lemmas are also used.

Definition 1: The system (1) is said to be globally asymptotically stable in the mean square sense if the following holds

$$\mathcal{E}\left\{\int_0^\infty \|x(t)\|^2 dt \,|x_0,r_0\right\} < \infty$$

for every initial condition $x_0 \in \mathbb{R}^n$ and $r_0 \in \mathcal{S}$.

Lemma 1: [10] (Jensen inequality) For any matrix $R > 0$, and an integrable function $\{x(t)|t \in [a,b]\}$, the following inequalities hold:

$$\int_a^b x^T(s)Rx(s)ds \geq \frac{1}{b-a}\left(\int_a^b x(s)\right)^T R \int_a^b x(s)ds \tag{5}$$

$$\int_b^a \int_\theta^b x^T(s)Rx(s)dsd\theta$$
$$\geq \frac{2}{(b-a)^2}\left(\int_b^a \int_\theta^b x(s)dsd\theta\right)^T R \int_b^a \int_\theta^b x(s)dsd\theta. \tag{6}$$

Lemma 2: [13] (Extended Wirtinger's inequality) For any matrix $R>0$, and all continuous differentiable functions $x(t) : [b,a] \to \mathbb{R}^n$, the following inequality holds.

$$\int_b^a x^T(s)Rx(s)ds \geq \frac{1}{(a-b)}\left(\int_b^a x(s)ds\right)^T R \left(\int_b^a x(s)ds\right)$$
$$+ \frac{3}{(a-b)}\Omega_1^T R\Omega_1 \tag{7}$$

$$\int_b^a \dot{x}^T(s)R\dot{x}(s)ds \geq \frac{1}{(a-b)}\left(x(a)-x(b)\right)^T R \left(x(a)-x(b)\right)$$
$$+ \frac{3}{(a-b)}\Omega_2^T R\Omega_2 \tag{8}$$

where

$$\Omega_1 = \int_b^a x(s)ds - \frac{2}{a-b}\int_b^a \int_\theta^a x(s)dsd\theta$$

$$\Omega_2 = x(a)+x(b) - \frac{2}{a-b}\int_b^a x(s)ds.$$

Lemma 3: [20] (Auxiliary function-based double integral inequality) For any matrix $R>0$, and a differentiable functions $\{x(t)|t \in [b,a]\}$, the following inequality holds:

$$\int_b^a \int_\theta^a \dot{x}^T(s)R\dot{x}(s)dsd\theta \geq 2\Omega_3^T R\Omega_3 + 4\Omega_4^T R\Omega_4 \tag{9}$$

where

$$\Omega_3 = x(a) - \frac{1}{a-b}\int_b^a x(s)ds$$

$$\Omega_4 = x(a) + \frac{2}{a-b}\int_b^a x(s)ds - \frac{6}{(a-b)^2}\int_b^a\int_\theta^a x(s)dsd\theta.$$

III. DELAY-DEPENDENT STABILITY ANALYSIS

In this section, the delay-dependent stability problem for MJTDSs with partially unknown transition rates is investigated. And some stability theorems are proposed.

Theorem 1: Given scalar $h > 0$, the system (1) is globally asymptotically stable in the mean square sense if there exist matrices with appropriate dimensions $P_i = P_i^T, W_i = W_i^T$, $Q_1 > 0, Q_2 > 0, R_1 > 0, R_2 > 0$ and $Z > 0$ such that the following LMIs hold for $\dot{d}(t) \in [\mu_1, \mu_2], i \in \mathcal{S}$

$$\begin{bmatrix} \widehat{e}_1 \\ h\widehat{e}_2 \end{bmatrix}^T P_i \begin{bmatrix} \widehat{e}_1 \\ h\widehat{e}_2 \end{bmatrix} + \frac{h^3}{4}\widehat{e}_1 Z\widehat{e}_1 + h\widehat{e}_2 Q_2\widehat{e}_2 + 2h^2\widehat{e}_3^T R_1\widehat{e}_3$$
$$+3h\left(\widehat{e}_2-2\widehat{e}_3\right)^T Q_2\left(\widehat{e}_2-2\widehat{e}_3\right)+2\left(\widehat{e}_1-\widehat{e}_2\right)^T R_2\left(\widehat{e}_1-\widehat{e}_2\right)$$
$$+4\left(\widehat{e}_1+2\widehat{e}_2-6\widehat{e}_3\right)^T R_2\left(\widehat{e}_1+2\widehat{e}_2-6\widehat{e}_3\right)+h^3\widehat{e}_3 Z\widehat{e}_3$$
$$-\frac{h^3}{2}Sym\{\widehat{e}_1 Z\widehat{e}_3\} > 0 \qquad (10)$$

$$\mathrm{Sym}\left\{\begin{bmatrix} e_1 \\ he_6 \end{bmatrix}^T P_i \begin{bmatrix} A_i e_1 + A_{di} e_2 \\ e_1 - e_3 \end{bmatrix}\right\}$$
$$+\sum_{j\in U_k^i} \pi_{ij}\begin{bmatrix} e_1 \\ he_6 \end{bmatrix}^T (P_j - W_i)\begin{bmatrix} e_1 \\ he_6 \end{bmatrix}$$
$$+ e_1^T\left(Q_1 + Q_2 + hR_1\right)e_1 - (1-\dot{d}(t))e_2^T Q_1 e_2$$
$$- e_3^T Q_2 e_3 - he_6^T R_1 e_6 - 3h\left(e_6-e_7\right)^T R_1\left(e_6-e_7\right)$$
$$+ \left(A_i e_1 + A_{di}e_2\right)^T\left(hR_2 + \frac{h^4}{4}Z\right)\left(A_i e_1 + A_{di}e_2\right)$$
$$- h^2\left(e_1-e_6\right)^T Z\left(e_1-e_6\right)$$
$$- 2h^2\left(e_1 + 2e_6 - 6e_7\right)^T Z\left(e_1 + 2e_6 - 6e_7\right)$$
$$- \frac{1}{h}\left[\left(e_1-e_2\right)^T R_2\left(e_1-e_2\right) + \left(e_2-e_3\right)^T R_2\left(e_2-e_3\right)\right]$$
$$- \frac{3}{h}\left(e_1+e_2-2e_4\right)^T R_2\left(e_1+e_2-2e_4\right)$$
$$- \frac{3}{h}\left(e_2+e_3-2e_5\right)^T R_2\left(e_2+e_3-2e_5\right) < 0 \qquad (11)$$

$$P_j - W_i \geq 0, \quad j \in U_{uk}^i; j \neq i; \qquad (12)$$
$$P_j - W_i \geq 0, \quad j \in U_{uk}^i; j = i. \qquad (13)$$

Proof: Construct an ALKF candidate as

$$V(t) = \eta^T(t)P(r_t)\eta(t) + \int_{t-d(t)}^t x^T(s)Q_1 x(s)ds$$
$$+ \int_{t-h}^t x^T(s)Q_2 x(s)ds + \int_{t-h}^t\int_\theta^t x^T(s)R_1 x(s)dsd\theta$$
$$+ \int_{t-h}^t\int_\theta^t \dot{x}^T(s)R_2\dot{x}(s)dsd\theta$$
$$+ \frac{h^2}{2}\int_{t-h}^t\int_\theta^t\int_s^t \dot{x}^T(v)Z\dot{x}(v)dvdsd\theta \qquad (14)$$

where $\eta(t) = \mathrm{col}\left\{x(t), \int_{t-h}^t x(s)ds\right\}$, $P(r_t) \in \mathbb{R}^{2n\times 2n}$ are real symmetric matrices, $r_t \in \mathcal{S}$. Q_1, Q_2, R_1, R_2 and Z are symmetric-positive-definite matrices, which belong to $\mathbb{R}^{n\times n}$ and to be determined.

When $r_t = i \in \mathcal{S}$, matrix $P(r_t)$ is re-expressed as P_i, it is noted that the matrix P_i is only symmetric matrix. Compared with P_i satisfying the condition $P_i > 0$, the constraint in this paper is weakened. For this case, LKF (14) is not a positive function. So the course of the proof is divided into the following two steps based on Lyapunov-Krasovskii stability theorem [21].

Step 1: Find conditions to guarantee the positiveness of the ALKF (14).

Applying Lemmas 1, 2 and 3 to value the lower bound of the integral terms in the ALKF respectively, the following inequalities hold:

$$\int_{t-h}^t x^T(s)Q_2 x(s)ds$$
$$\geq \frac{1}{h}\left(\int_{t-h}^t x(s)ds\right)^T Q_2\int_{t-h}^t x(s)ds$$
$$+ \frac{3}{h}\left(\int_{t-h}^t x(s)ds - \frac{2}{h}\int_{t-h}^t\int_\theta^t x(s)dsd\theta\right)^T Q_2$$
$$\times \int_{t-h}^t x(s)ds - \frac{2}{h}\int_{t-h}^t\int_\theta^t x(s)dsd\theta \qquad (15)$$

$$\int_{t-h}^t\int_\theta^t x^T(s)R_1 x(s)dsd\theta$$
$$\geq \frac{2}{h^2}\left(\int_{t-h}^t\int_\theta^t x(s)dsd\theta\right)^T R_1\int_{t-h}^t\int_\theta^t x(s)dsd\theta \qquad (16)$$

$$\frac{h^2}{2}\int_{t-h}^t\int_\theta^t\int_s^t \dot{x}^T(v)Z\dot{x}(v)dvdsd\theta$$
$$\geq \frac{h^2}{2}\int_{t-h}^t\int_\theta^t \frac{1}{t-s}\left[x(t)-x(s)\right]^T Z\left[x(t)-x(s)\right]dsd\theta$$
$$\geq \frac{h}{2}\int_{t-h}^t\int_\theta^t x^T(t)Zx(t)dsd\theta - hx^T(t)Z\int_{t-h}^t\int_\theta^t x(t)dsd\theta$$
$$+ \frac{h}{2}\int_{t-h}^t\int_\theta^t x^T(s)Zx(s)dsd\theta.$$
$$\geq \frac{h^3}{4}x^T(t)Zx(t) - hx^T(t)Z\int_{t-h}^t\int_\theta^t x(t)dsd\theta$$
$$+ \frac{1}{h}\left(\int_{t-h}^t\int_\theta^t x(t)dsd\theta\right)^T Z\int_{t-h}^t\int_\theta^t x(t)dsd\theta \qquad (17)$$

$$\int_{t-h}^t\int_\theta^t \dot{x}^T(s)R_2\dot{x}(s)dsd\theta \geq 2\Upsilon_1^T R_2\Upsilon_1 + 4\Upsilon_2^T R_2\Upsilon_2 \quad (18)$$

where

$$\Upsilon_1 = x(t) - \frac{1}{h}\int_{t-h}^t x(s)ds$$
$$\Upsilon_2 = x(t) + \frac{2}{h}\int_{t-h}^t x(s)ds - \frac{6}{h^2}\int_{t-h}^t\int_\theta^t x(s)dsd\theta.$$

Setting $\zeta(t) = \mathrm{col}\left\{x(t), \frac{1}{h}\int_{t-h}^t x(s)ds, \frac{1}{h^2}\int_{t-h}^t\int_\theta^t x(s)dsd\theta\right\}$ then

$$\eta(t) = \begin{bmatrix} \widehat{e}_1 \\ h\widehat{e}_2 \end{bmatrix}\zeta(t).$$

Combining (15)−(18) with (14), the lower bound of $V(t)$ can be derived as follows:

$$V(t) \geq \zeta^T(t)\left[\begin{bmatrix}\widehat{e}_1 \\ h\widehat{e}_2\end{bmatrix}^T P_i \begin{bmatrix}\widehat{e}_1 \\ h\widehat{e}_2\end{bmatrix} + h\widehat{e}_2 Q_2 \widehat{e}_2\right.$$
$$+3h\left(\widehat{e}_2 - 2\widehat{e}_3\right)^T Q_2 \left(\widehat{e}_2 - 2\widehat{e}_3\right)$$
$$+2h^2\widehat{e}_3^T R_1\widehat{e}_3 + 2\left(\widehat{e}_1 - \widehat{e}_2\right)^T R_2 \left(\widehat{e}_1 - \widehat{e}_2\right)$$
$$+4\left(\widehat{e}_1 + 2\widehat{e}_2 - 6\widehat{e}_3\right)^T R_2 \left(\widehat{e}_1 + 2\widehat{e}_2 - 6\widehat{e}_3\right)$$
$$\left.+\frac{h^3}{4}\widehat{e}_1 Z\widehat{e}_1 - \frac{h^3}{2}Sym\{\widehat{e}_1 Z\widehat{e}_3\} + h^3\widehat{e}_3 Z\widehat{e}_3\right]\zeta(t)$$
$$+\int_{t-d(t)}^t x^T(s)Q_1 x(s)ds. \tag{19}$$

Note that $Q_1 > 0$, so LMI (10) ensures that $V(t) > 0$.

Step 2: Derive conditions to ensure the negativeness of weak infinitesimal generator of the ALKF (14).

Since the solution of $\{x(t), t\}$ of MJTDSs is a Markov process, the weak infinitesimal generator acting on function $V(t)$ is

$$\mathcal{L}V(t) = 2\eta^T(t)P_i\dot{\eta}(t) + \sum_{j=1}^{\mathcal{N}}\pi_{ij}\eta^T(t)P_j\eta(t)$$
$$+x^T(t)Q_1 x(t) - (1-\dot{d}(t))x^T(t-d(t))Q_1 x(t-d(t))$$
$$+x^T(t)Q_2 x(t) - x^T(t-h)Q_2 x(t-h)$$
$$+hx^T(t)R_1 x(t) - \int_{t-h}^t x^T(s)R_1 x(s)ds$$
$$+h\dot{x}^T(t)R_2\dot{x}(t) - \int_{t-h}^t \dot{x}^T(s)R_2\dot{x}(s)ds$$
$$+\frac{h^4}{4}\dot{x}^T(t)Z\dot{x}(t) - \frac{h^2}{2}\int_{t-h}^t\int_\theta^t \dot{x}^T(s)Z\dot{x}(s)dsd\theta. \tag{20}$$

Using Lemmas (2) and (3), the following inequalities hold:

$$-\int_{t-h}^t x^T(s)R_1 x(s)ds$$
$$\leq -\frac{1}{h}\left(\int_{t-h}^t x(s)ds\right)^T R_1\int_{t-h}^t x(s)ds - \frac{3}{h}\Upsilon_3^T R_1\Upsilon_3 \tag{21}$$

$$-\int_{t-d(t)}^t \dot{x}^T(s)R_2\dot{x}(s)ds$$
$$\leq -\frac{1}{d(t)}\left[x(t)-x(t-d(t))\right]^T R_2\left[x(t)-x(t-d(t))\right]$$
$$-\frac{3}{d(t)}\Upsilon_4^T R_2\Upsilon_4 \tag{22}$$

$$-\int_{t-h}^{t-d(t)} \dot{x}^T(s)R_2\dot{x}(s)ds$$
$$\leq -\frac{1}{h-d(t)}\left[x(t-d(t))-x(t-h)\right]^T R_2$$
$$\times\left[x(t-d(t))-x(t-h)\right] - \frac{3}{h-d(t)}\Upsilon_5^T R_2\Upsilon_5 \tag{23}$$

$$-\frac{h^2}{2}\int_{t-h}^t\int_\theta^t \dot{x}^T(s)Z\dot{x}(s)dsd\theta$$
$$\leq -h^2\left[x(t)-\frac{1}{h}\int_{t-h}^t x(s)ds\right]^T Z\left[x(t)-\frac{1}{h}\int_{t-h}^t x(s)ds\right]$$
$$-2h^2\Upsilon_6^T Z\Upsilon_6 \tag{24}$$

where

$$\Upsilon_3 = \int_{t-h}^t x(s)ds - \frac{2}{h}\int_{t-h}^t\int_\theta^t x(s)dsd\theta$$

$$\Upsilon_4 = x(t) + x(t-d(t)) - \frac{2}{d(t)}\int_{t-d(t)}^t x(s)ds$$

$$\Upsilon_5 = x(t-d(t)) + x(t-h) - \frac{2}{h-d(t)}\int_{t-h}^{t-d(t)} x(s)ds$$

$$\Upsilon_6 = x(t) + \frac{2}{h}\int_{t-h}^t x(s)ds - \frac{6}{h^2}\int_{t-h}^t\int_\theta^t x(s)dsd\theta.$$

On the other hand, the transition rates cannot be completely known, the following equations hold via introducing the free-connection weighting matrices $W_i = W_i^T \in \mathbb{R}^{2n\times 2n}, i \in \mathcal{S}$:

$$0 = -\eta^T(t)\sum_{j=1}^{\mathcal{N}}\pi_{ij}W_i\eta(t). \tag{25}$$

Setting $\xi(t) = col\left\{x(t), x(t-d(t)), x(t-h), \frac{1}{d(t)}\int_{t-d(t)}^t x(s)ds,\right.$ $\frac{1}{h-d(t)}\int_{t-h}^{t-d(t)}x(s)ds, \frac{1}{h}\int_{t-h}^t x(s)ds, \frac{1}{h^2}\int_{t-h}^t\int_\theta^t x(s)dsd\theta\left.\right\}$ and combining (21)−(25) with (20), it yields

$$\mathcal{L}V(t) \leq \xi^T(t)\left\{Sym\left\{\begin{bmatrix}e_1 \\ he_6\end{bmatrix}^T P_i \begin{bmatrix}A_i e_1 + A_{di}e_2 \\ e_1 - e_3\end{bmatrix}\right\}\right.$$
$$+\sum_{j\in U_k^i}\pi_{ij}\begin{bmatrix}e_1 \\ he_6\end{bmatrix}^T (P_j - W_i)\begin{bmatrix}e_1 \\ he_6\end{bmatrix}$$
$$+e_1^T (Q_1 + Q_2 + hR_1)e_1 - (1-\dot{d}(t))e_2^T Q_1 e_2$$
$$-e_3^T Q_2 e_3 - he_6^T R_1 e_6 - 3h(e_6 - e_7)^T R_1(e_6 - e_7)$$
$$+(A_i e_1 + A_{di}e_2)^T\left(hR_2 + \frac{h^4}{4}Z\right)(A_i e_1 + A_{di}e_2)$$
$$-h^2(e_1 - e_6)^T Z(e_1 - e_6)$$
$$-2h^2(e_1 + 2e_6 - 6e_7)^T Z(e_1 + 2e_6 - 6e_7)$$
$$-\frac{1}{h}\left[(e_1 - e_2)^T R_2(e_1 - e_2) + (e_2 - e_3)^T R_2(e_2 - e_3)\right]$$
$$-\frac{3}{h}(e_1 + e_2 - 2e_4)^T R_2(e_1 + e_2 - 2e_4)$$
$$\left.-\frac{3}{h}(e_2 + e_3 - 2e_5)^T R_2(e_2 + e_3 - 2e_5)\right\}\xi(t)$$
$$+\sum_{j\in U_{uk}^i}\pi_{ij}\begin{bmatrix}e_1 \\ he_6\end{bmatrix}^T (P_j - W_i)\begin{bmatrix}e_1 \\ he_6\end{bmatrix}. \tag{26}$$

Due to $\pi_{ij} \geq 0, j \neq i$ and $\pi_{ij} \leq 0, j = i$, hence (11)−(13) can guarantee the negativeness of weak infinitesimal generator of the ALKF (14). Therefore, the trivial solution of the MJTDSs (1) is asymptotically stable in the mean square sense. The proof is completed. ∎

Remark 1: To reduce the computational complexity and achieve better performance, auxiliary function-based double integral inequality is combined with extended Wirtinger's inequality and Jensen inequality to estimate the lower bound of the ALKF, the constraint conditions of matrices $P_i, i \in \mathcal{S}$ are weakened. At the same time, the above three inequalities are used to estimate the upper bound of weak infinitesimal generator of the ALKF, as a result, the more accurate approximation bounds with a fewer variables are derived.

Remark 2: Compared with [24], the positive-definite matrices $Q_1(r_t)$ and $Q_2(r_t)$ are replaced with Q_1 and Q_2, respectively. Although this treatment could result in conservative criterion, it could greatly reduce the computational complexity from the introduction of the variables. For the same reason, the time delay, $d(t)$, and the difference between h and $d(t)$, $h - d(t)$, are enlarged as h (23) and (24), respectively. As is known, reciprocally convex approach in [12] and the extended vector inequality in [25] can deal with this case, but the variables will be introduced in large quantities, which also increase the computational complexity.

If the information of transition probabilities is known completely, the following Corollary can be derived based on Theorem 1.

Corollary 1: Given scalar $h > 0$, the system (1) is globally asymptotically stable in the mean square sense if there exist matrices with appropriate dimensions $P_i = P_i^T, Q_1 > 0, Q_2 > 0, R_1 > 0, R_2 > 0$ and $Z > 0$ such that the following LMIs hold for $\dot{d}(t) \in [\mu_1, \mu_2], i \in \mathcal{S}$

$$\begin{bmatrix} \widehat{e}_1 \\ h\widehat{e}_2 \end{bmatrix}^T P_i \begin{bmatrix} \widehat{e}_1 \\ h\widehat{e}_2 \end{bmatrix} + \frac{h^3}{4} \widehat{e}_1 Z \widehat{e}_1 + h\widehat{e}_2 Q_2 \widehat{e}_2 + 2h^2 \widehat{e}_3^T R_1 \widehat{e}_3$$
$$+ 3h (\widehat{e}_2 - 2\widehat{e}_3)^T Q_2 (\widehat{e}_2 - 2\widehat{e}_3) + 2 (\widehat{e}_1 - \widehat{e}_2)^T R_2 (\widehat{e}_1 - \widehat{e}_2)$$
$$+ 4 (\widehat{e}_1 + 2\widehat{e}_2 - 6\widehat{e}_3)^T R_2 (\widehat{e}_1 + 2\widehat{e}_2 - 6\widehat{e}_3) + h^3 \widehat{e}_3 Z \widehat{e}_3$$
$$- \frac{h^3}{2} \text{Sym} \{\widehat{e}_1 Z \widehat{e}_3\} > 0 \qquad (27)$$

$$\text{Sym} \left\{ \begin{bmatrix} e_1 \\ he_6 \end{bmatrix}^T P_i \begin{bmatrix} A_i e_1 + A_{di} e_2 \\ e_1 - e_3 \end{bmatrix} \right\} + \sum_{j=1}^{\mathcal{N}} \pi_{ij} \begin{bmatrix} e_1 \\ he_6 \end{bmatrix}^T P_j \begin{bmatrix} e_1 \\ he_6 \end{bmatrix}$$
$$+ e_1^T (Q_1 + Q_2 + hR_1) e_1 - (1 - \dot{d}(t)) e_2^T Q_1 e_2$$
$$- e_3^T Q_2 e_3 - he_6^T R_1 e_6 - 3h (e_6 - e_7)^T R_1 (e_6 - e_7)$$
$$+ (A_i e_1 + A_{di} e_2)^T \left(hR_2 + \frac{h^4}{4} Z \right) (A_i e_1 + A_{di} e_2)$$
$$- h^2 (e_1 - e_6)^T Z (e_1 - e_6)$$
$$- 2h^2 (e_1 + 2e_6 - 6e_7)^T Z (e_1 + 2e_6 - 6e_7)$$
$$- \frac{1}{h} \left[(e_1 - e_2)^T R_2 (e_1 - e_2) + (e_2 - e_3)^T R_2 (e_2 - e_3) \right]$$
$$- \frac{3}{h} (e_1 + e_2 - 2e_4)^T R_2 (e_1 + e_2 - 2e_4)$$
$$- \frac{3}{h} (e_2 + e_3 - 2e_5)^T R_2 (e_2 + e_3 - 2e_5) < 0. \qquad (28)$$

IV. NUMERICAL EXAMPLES

In this section, two numerical examples are given to prove the effectiveness and the merits of the criterion presented in this paper.

Example 1: [24] Consider the MJSs (1) with four modes, whose state matrices are

$$A_1 = \begin{bmatrix} -1.7460 & -1.4410 \\ -1.5937 & -2.4289 \end{bmatrix}, \quad A_2 = \begin{bmatrix} -1.3999 & 0.8156 \\ -0.6900 & -0.2881 \end{bmatrix}$$

$$A_3 = \begin{bmatrix} -0.2523 & 0.7500 \\ 1.5630 & -1.8540 \end{bmatrix}, \quad A_4 = \begin{bmatrix} -1.2840 & 0.3640 \\ 1.3670 & -1.0640 \end{bmatrix}$$

$$A_{d1} = \begin{bmatrix} -0.5 & -0.5 \\ 0.05 & 0.01 \end{bmatrix}, \qquad A_{d2} = \begin{bmatrix} 0.01 & 0 \\ 0.05 & -0.01 \end{bmatrix}$$

$$A_{d3} = \begin{bmatrix} -0.01 & -0.02 \\ 0.05 & 0.01 \end{bmatrix}, \quad A_{d4} = \begin{bmatrix} 0.03 & 0.01 \\ -0.2 & -0.1 \end{bmatrix}.$$

The transition rate matrix Π is

$$\Pi = \begin{bmatrix} -1.3 & 0.2 & ? & ? \\ ? & ? & 0.3 & 0.3 \\ ? & ? & -1.5 & ? \\ ? & ? & ? & -1.2 \end{bmatrix}.$$

The corresponding upper bounds of time-varying delay for various μ are listed in Table I, which shows that Theorem 1 indeed ameliorate the allowable upper bound h. Table II provides numbers of scalar variables for different methods to further illustrate the merit of our method.

TABLE I
ALLOWABLE UPPER BOUNDS h FOR VARIOUS μ

Methods	$\mu = 0$	$\mu = 0.2$	$\mu = 0.5$	$\mu \geq 1$
Theorem 1 in [24]	1.4931	1.4926	1.4926	1.4926
Theorem 1 in this paper	1.8436	1.6908	1.5272	1.4980

TABLE II
NUMBERS OF SCALAR VARIABLES FOR DIFFERENT METHODS

Methods	Numbers of scalar variables
Theorem 1 in [24]	$(6.5n^2 + 2.5n) \mathcal{N} + n^2 + n$
Theorem 1 in this paper	$(4n^2 + 2n) \mathcal{N} + 2.5n^2 + 2.5n$

Because $\mathcal{N}, n \in \mathbb{N}$, $(6.5n^2 + 2.5n) \mathcal{N} + n^2 + n$ is much larger than $(4n^2 + 2n) \mathcal{N} + 2.5n^2 + 2.5n$ unless $\mathcal{N} = n = 1$, so our method greatly reduces the computational complexity.
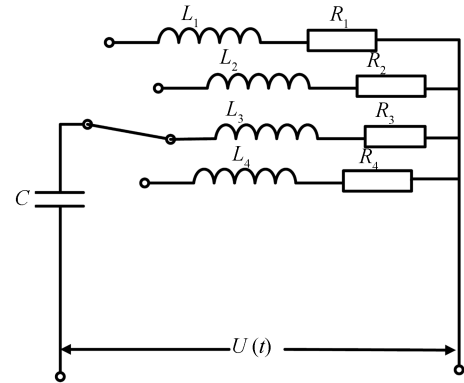

Fig. 1. RLC series circuit.

Example 2: (A similar example can be found in reference [7]) Consider the RLC series circuit (Fig. 1), open-loop system can be established as the MJSs (1) taking into account the time-varying delay, where $x(t) = \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix}$ with $x_1(t) = i_L(t), x_2(t) = U_C(t)$, states parameters are obtained as follows based on Kirchhoff's law:

$$A_i = \begin{bmatrix} -\frac{R_i}{L_i} & -\frac{1}{L_i} \\ \frac{1}{C} & 0 \end{bmatrix}, \quad i = 1, 2, 3, 4$$

with $R_1 = 0.61\Omega, R_2 = 0.02\Omega, R_3 = 0.62\Omega, R_4 = 0.06\Omega, L_1 = 1mH, L_2 = 2mH, L_3 = 4.58mH, L_4 = 2.58mH, C = 1.3mF$. And other parameters are

$$A_{d1} = \begin{bmatrix} -0.05 & -0.05 \\ 0.05 & 0.01 \end{bmatrix}, \quad A_{d2} = \begin{bmatrix} 0.01 & 0 \\ 0.05 & -0.01 \end{bmatrix}$$

$$A_{d3} = \begin{bmatrix} -0.01 & -0.02 \\ 0.05 & 0.01 \end{bmatrix}, \quad A_{d4} = \begin{bmatrix} 0.03 & 0.01 \\ -0.02 & -0.1 \end{bmatrix}.$$

The transition rate matrix Π is described in Example 1. Applying Theorem 1, the corresponding upper bounds of time-varying delay for various μ are obtained and listed in Table III.

TABLE III
ALLOWABLE UPPER BOUNDS h FOR VARIOUS μ

$\mu = 0$	$\mu = 0.2$	$\mu = 0.5$	$\mu \geq 1$
0.8216	0.8148	0.8089	0.8089

Furthermore, simulation results under an initial condition $x_0 = [-0.8 \ 0.2]^T$ and time-varying delay $d(t) = 0.8\sin^2(t) + 0.0089$ are shown in Fig. 2. To make the example more persuasive, the state response with 1000 random samples (Fig. 3) is provided in this example, which demonstrates the stability of this system.
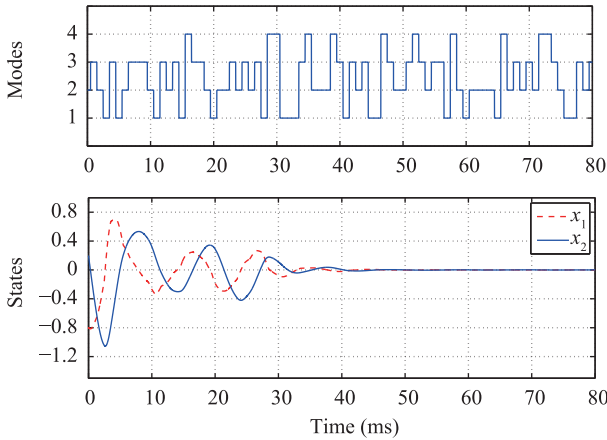
Fig. 2.   State responses with switching modes.
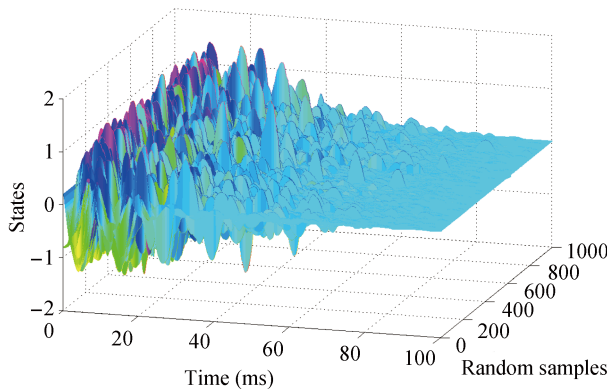
Fig. 3.   State response with 1000 random samples.

Example 3: Consider the MJSs (1) with two modes, whose state matrices are given in [26].

$$A_1 = \begin{bmatrix} -3.4888 & 0.8057 \\ -0.6451 & -3.2684 \end{bmatrix}, \quad A_{d1} = \begin{bmatrix} -0.8620 & -1.2919 \\ -0.6841 & -2.0729 \end{bmatrix}$$

$$A_2 = \begin{bmatrix} -2.4898 & 0.2895 \\ 1.3396 & -0.0211 \end{bmatrix}, \quad A_{d2} = \begin{bmatrix} -2.8306 & 0.4978 \\ -0.8436 & -1.0115 \end{bmatrix}.$$

The transition rate matrix Π is considered as

$$\Pi = \begin{bmatrix} -0.1 & 0.1 \\ 0.8 & -0.8 \end{bmatrix}.$$

The maximal allowable upper bounds h for different μ are obtained by applying the Theorem 1 in [26] and Corollary 1 in this paper, which are listed in Table IV. This table shows the merit of our method again.

TABLE IV
THE MAXIMAL ALLOWABLE UPPER BOUND h FOR DIFFERENT μ

Methods	$\mu=0.6$	$\mu=0.8$	$\mu=1.6$	Numbers of scalar variables
Theorem 1 in [26]	0.5159	0.4814	0.4789	$(2n^2+2n)\mathcal{N}+2.5n^2+2.5n$
Corollary 1 in this paper	0.5870	0.5200	0.4925	$(2n^2+n)\mathcal{N}+2.5n^2+2.5n$

V. CONCLUSION

By combining auxiliary function-based double integral with extended Wirtinger's inequality and Jensen inequality, the conditions to ensure the positiveness of ALKF are weakened and much tighter bound of double integral terms is estimated in the weak infinitesimal generator of the ALKF. At the same time, free-connection weighting matrix method is used to separate the unknown transition rates. Consequently, the new results on delay-dependent stability for MJTDSs are obtained in this paper. Compared with previous criteria, our results require fewer scalar variables and have less conservative. Numerical examples illustrate this point.

Notwithstanding our results have less conservative at present. As we say in Remark 2, $d(t)$ and $h - d(t)$ are enlarged as h, the results of such processing give rise to conservatism. Therefore, how to find a compromise formula between computational complexity and conservatism is our next work.

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