# Asynchronous $L_{1}$ control of delayed switched positive systems with mode-dependent average dwell time 

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#### Abstract

This paper investigates the stability and asynchronous $L_{1}$ control problems for a class of switched positive linear systems (SPLSs) with time-varying delays by using the mode-dependent average dwell time (MDADT) approach. By allowing the co-positive type LyapunovKrasovskii functional to increase during the running time of active subsystems, a new stability criterion for the underlying system with MDADT is first derived. Then, the obtained results are extended to study the issue of asynchronous $L_{1}$ control, where "asynchronous" means that the switching of the controllers has a lag with respect to that of system modes. Sufficient conditions are provided to guarantee that the resulting closed-loop system is exponentially stable and has an $L_{1}$-gain performance. Finally, two numerical examples are given to show the effectiveness of the developed results.


## 1. Introduction

Positive systems are dynamic systems whose state variables are constrained to be positive (at least nonnegative) at all times. Such systems abound in various fields, e.g., biomedicine [4], ecology [5], and TCP-like Internet congestion control [15]. During the past decades, switched systems have been investigated by many researchers due to the theoretical devel-opment as well as practical applications [21]. Several methods have been developed to study the stability of switched systems, such as the common Lyapunov function approach, the average dwell time (ADT) scheme, and the multiple Lyapunov functions method (see [1,10,25-27]). Recently, the mode-dependent average dwell time (MDADT) approach [37] has been proposed for the stability analysis and control synthesis of switched systems. It has been shown that the results obtained by the MDADT approach are more general than those derived by other methods.

Recently, switched positive linear systems (SPLSs) which consist of a family of positive linear subsystems and a switching signal governing the switching among them have received considerable attention due to their broad applications in congestion control [3] and communication systems [14]. Many useful results on stability and stabilization of such systems have appeared (see $[2,8,9,12,13,31,32,36,38,39]$ ). Because time-delay phenomena exist widely in engineering and social systems and often cause instability or bad system performance in control systems, time-delay systems have been extensively studied (see [11,12,16-20,28-30,39]). Some results on SPLSs with time-delays have been obtained [12,39].

On the other hand, the disturbance rejection problem has been a hot topic [7]. Some results on $L_{1}$-gain analysis for positive systems have been reported in [6,24]. The reason for this study is that the $L_{1}$-gain can provide a more useful
description for positive systems because 1-norm gives the sum of the values of the components, which is more appropriate, for instance, if the values represent the amount of material or the number of animal in a species [6].

In almost all the aforementioned works on switched positive systems, a very common assumption in the state-feedback stabilization context is that the controllers are switched synchronously with the switching of system modes, which is quite unpractical. As pointed out in [27], there inevitably exists asynchronous switching in actual operation, i.e. the switching instants of the controllers exceed or lag behind those of the subsystems. Thus, it is necessary to consider asynchronous switching for realistic control. Some results on switched systems under asynchronous switching have been proposed in [22,23,25,26,33-35]. However, to the best of our knowledge, the asynchronous $L_{1}$ control problem for SPLSs, which constitutes the main motivation of the present study, has not been investigated yet.

The main contribution of this paper is threefold: (1) by constructing an appropriate co-positive type Lyapunov-Krasovskii functional, a new stability criterion is derived by using the MDADT method; (2) by allowing the Lyapunov-Krasovskii functional to increase during the running time of active subsystems, improved stability and $L_{1}$-gain analysis results are obtained; (3) the obtained results are extended to study the issue of asynchronous $L_{1}$ control.

The remainder of this paper is organized as follows. In Section 2, problem statements and necessary lemmas are given. In Section 3, based on the MDADT approach, stability and asynchronous $L_{1}$ control problems for SPLSs with time-varying delays are addressed, and sufficient conditions are also provided to guarantee the exponential stability of the closed-loop system. Two numerical examples are provided to show the effectiveness of the proposed approach in Section 4. Concluding remarks are given in Section 5.

Notation: In this paper, $A \succ 0(A \succeq 0)$ means that all the elements of $A$ are positive (nonnegative). $A \succ B(A \succeq B)$ means that $A-B \succ 0(A-B \succeq 0) . R^{n}$ is the $n$-dimensional real vector space and $R_{+}^{n}$ is the set of $n$-dimensional vectors with nonnegative elements; $R^{n \times s}$ is the set of all real matrices of $(n \times s)$-dimension. $A^{T}$ denotes the transpose of $A$. The vector 1 -norm of $x \in R^{n}$ is denoted by $\|x\|=\sum_{l=1}^{n}\left|x_{l}\right|$, where $x_{l}$ is the lth element of $x .1_{q}$ denotes the column vector with $q$ rows containing only 1 entries; $1_{f}$ denotes the column vector with $f$ rows containing only 1 entries; For scalars $y_{a}, y_{a+1}, \ldots, y_{b}, y_{a} y_{a+1} \ldots y_{b}$ is denoted by $\Pi_{i=a}^{b} y_{i} ; \exp \{\cdot\}$ is the exponential operate; $L_{1}\left[t_{0}, \infty\right)$ is the space of absolute integrable vector-valued functions on $\left[t_{0}, \infty\right)$, i.e., we say $z:\left[t_{0}, \infty\right) \rightarrow R^{q}$ is in $L_{1}\left[t_{0}, \infty\right)$ if $\int_{t_{0}}^{\infty}\|z(t)\| d t<\infty$.

## 2. Problem statements and preliminaries

Consider the following switched linear system with time-varying delay:

$$
\left\{\begin{array}{l}
\dot{x}(t)=A_{\sigma(t)} x(t)+G_{\sigma(t)} x(t-d(t))+B_{\sigma(t)} u(t)+E_{\sigma(t)} w(t),  \tag{1}\\
x\left(t_{0}+\theta\right)=\varphi(\theta), \theta \in[-\tau, 0] \\
z(t)=C_{\sigma(t)} x(t)+D_{\sigma(t)} w(t),
\end{array}\right.
$$

where $x(t) \in R^{n}, u(t) \in R^{p}$ and $z(t) \in R^{q}$ denote the system state, the control input, and the controlled output, respectively; $w(t) \in R^{f}$ is the disturbance input which belongs to $L_{1}\left[t_{0}, \infty\right) ; \sigma(t):\left[t_{0}, \infty\right) \rightarrow \underline{N}=\{1,2, \ldots, N\}$ is a piecewise constant function of time, called a switching signal; $N$ is the number of subsystems; $t_{0}$ is the initial instant; $A_{i}, G_{i}, B_{i}, C_{i}, D_{i}$ and $E_{i}, \forall i \in \underline{N}$, are known constant matrices with appropriate dimensions; $d(t)$ denotes the time-varying delay satisfying $0 \leqslant d(t) \leqslant \tau$ and $\dot{d}(t) \leqslant d<1$ for known positive constants $\tau$ and $d ; \varphi(\theta)$ is a continuous vector-valued initial function defined on interval $\tau, 0]$.
Next, we will give the definition of switched positive system (1).
Definition 1 [24]. System (1) is said to be an SPLS if for any switching signals $\sigma(t)$ and initial conditions $\varphi(\theta) \succeq 0, \theta \in[-\tau, 0]$, it satisfies $x(t) \succeq 0$ and $z(t) \succeq 0, \forall t \geqslant t_{0}$.

Definition 2 [8]. A is called a Metzler matrix, if its off-diagonal entries are non-negative.
Due to the asynchronous switching, the switching instants of the controllers do not coincide with those of the subsystems. Without loss of generality, the "asynchronous" means that the switching of the controllers has a lag with respect to that of system modes. Then, the real control input will become

$$
\begin{equation*}
u(t)=K_{\sigma\left(t-\Delta_{\kappa}\right)} x(t), \quad \forall t \in\left[t_{\kappa}, t_{\kappa+1}\right), \quad \kappa=0,1, \ldots \tag{2}
\end{equation*}
$$

where $\Delta_{0}=0$, and $\Delta_{\kappa}<t_{\kappa+1}-t_{\kappa}$ represents the delayed period.
Remark 1. The period $\Delta_{\kappa}$ guarantees that switching instants of the controllers lag behind the switches of system modes, and also there exists a period during which the system mode and the controller operate synchronously.

Let the $i$ th subsystem be activated at the switching instant $t_{\kappa}$, and the $j$ th subsystem be activated at the switching instant $t_{\kappa+1}$. Then the corresponding controllers are activated at the switching instants $t_{\kappa}+\Delta_{\kappa}$ and $t_{\kappa+1}+\Delta_{\kappa+1}$, respectively. Upon applying the controller (2) to system (1), the resulting closed-loop system is given by

$$
\begin{cases}\dot{x}(t)=\bar{A}_{i} x(t)+G_{i} x(t-d(t))+E_{i} w(t), &  \tag{3}\\ z(t)=C_{i} x(t)+D_{i} w(t), & \forall t \in\left[t_{\kappa}+\Delta_{\kappa}, t_{\kappa+1}\right) \\ \dot{x}(t)=\bar{A}_{i, j} x(t)+G_{j} x(t-d(t))+E_{j} w(t), & \\ z(t)=C_{j} x(t)+D_{j} w(t), & \forall t \in\left[t_{\kappa+1}, t_{\kappa+1}+\Delta_{\kappa+1}\right)\end{cases}
$$

where $\bar{A}_{i}=A_{i}+B_{i} K_{i}$ and $\bar{A}_{i, j}=A_{j}+B_{j} K_{i}$.
Lemma 1 [24]. System (3) is positive if and only if $\bar{A}_{i}$ and $\bar{A}_{i, j}$ are Metzler matrices, and $G_{i} \succeq 0, E_{i} \succeq 0, C_{i} \succeq 0, D_{i} \succeq 0, \forall(i, j) \in \underline{N} \times \underline{N}$, $i \neq j$.

Definition 3 [24]. System (1) is said to be exponentially stable under the switching signal $\sigma(t)$, if there exist constants $\zeta>0$ and $\rho>0$ such that the solution of the system satisfies $\|x(t)\| \leqslant \zeta\left\|x\left(t_{0}\right)\right\|_{c} e^{-\rho\left(t-t_{0}\right)}, \forall t \geqslant t_{0}$, where $\left\|x\left(t_{0}\right)\right\|_{c}=\sup _{-\tau \leqslant \theta \leqslant 0}\|\varphi(\theta)\|$.

Definition 4. [37]For a switching signal $\sigma(t)$ and any $T_{2} \geqslant T_{1} \geqslant 0$, let $N_{\sigma i}\left(T_{1}, T_{2}\right)$ be the switching number that the $i$ th subsystem is active over the interval $\left[T_{1}, T_{2}\right)$ and $T_{i}\left(T_{1}, T_{2}\right)$ be the total running time of the $i$ th subsystem over the interval $\left[T_{1}, T_{2}\right), i$ $\in \underline{N}$. We say that $\sigma(t)$ has a mode-dependent average dwell time $T_{a i}$ if there exist positive numbers $N_{0 i}$ (we call $N_{0 i}$ the mode dependent chatter bounds) and $T_{a i}$ such that $N_{\sigma i}\left(T_{1}, T_{2}\right) \leqslant N_{0 i}+T_{i}\left(T_{1}, T_{2}\right) / T_{a i}$ holds.

Definition 5. System (1) is said to have an $L_{1}$-gain performance level $\gamma$ under the switching signal $\sigma(t)$, if the following conditions are satisfied:
(i) system (1) is exponentially stable when $w(t) \equiv 0$;
(ii) under zero initial conditions, i.e., $\varphi(\theta)=0, \theta \in[-\tau, 0]$, the following inequality holds for all nonzero $w(t) \in L_{1}\left[t_{0}, \infty\right)$ :

$$
\begin{equation*}
\int_{t_{0}}^{\infty} e^{-\sum_{i=1}^{N}\left\{\alpha_{i} T_{i}\left(t_{0}, t\right)\right.}\|z(t)\| d t \leqslant \gamma \int_{t_{0}}^{\infty}\|w(t)\| d t \tag{4}
\end{equation*}
$$

where $\alpha_{i}$ is a given positive constant and $T_{i}\left(t_{0}, t\right)$ is the total running time of the $i$ th subsystem over the interval $\left[t_{0}, t\right)$.

Remark 2. When $\alpha_{i}=\alpha, \forall i \in \underline{N}$, (4) will degenerate into (2) in [24]. Thus Definition 5 given here can be viewed as an extension of the one proposed in [24].

The aim of this paper is to design a state-feedback controller and a set of admissible switching signals with MDADT such that the resulting closed-loop system (3) is positive and exponentially stable with $L_{1}$-gain performance.

## 3. Main results

### 3.1. Stability and $L_{1}$-gain analysis

Before proceeding further, we present here the following results on the exponential stability for the SPLS (1) with $u(t) \equiv 0$ and $w(t) \equiv 0$ for later use.

Theorem 1. Consider the SPLS (1) with $u(t) \equiv 0$ and $w(t) \equiv 0$. Let $\alpha_{i}>0$ be given constants. If there exist vectors $v_{i} \succ 0, v_{i} \succ 0$ and $\vartheta_{i} \succ 0$ of appropriate dimensions, such that, $\forall i \in \underline{N}$,

$$
\begin{align*}
& A_{i}^{T} v_{i}+\alpha_{i} v_{i}+v_{i}+\tau \vartheta_{i} \preceq 0,  \tag{5}\\
& G_{i}^{T} v_{i}-(1-d) e^{-\alpha_{i} \tau} v_{i} \preceq 0 \tag{6}
\end{align*}
$$

then the system is exponentially stable for any switching signal $\sigma(t)$ with the following MDADT

$$
\begin{equation*}
T_{a i}>T_{a i}^{*}=\ln \mu_{i} / \alpha_{i}, \tag{7}
\end{equation*}
$$

where $\mu_{i} \geqslant 1$ satisfy

$$
\begin{equation*}
v_{i} \preceq \mu_{i} v_{j}, v_{i} \preceq \mu_{i} v_{j}, v_{i} \preceq \mu_{i} \vartheta_{j}, \quad \forall(i, j) \in \underline{N} \times \underline{N} \tag{8}
\end{equation*}
$$

Proof. For any $T>0$, let $t_{0}=0$ and denote by $t_{1}, t_{2}, \ldots, t_{\kappa-1}, t_{\kappa}, \ldots, t_{N_{\sigma}(0, T)}$ the switching instants on the interval $\left[t_{0}, T\right)$, where $N_{\sigma}(0, T)=\sum_{i=1}^{N} N_{\sigma i}(0, T)$. Let $T_{i}(0, T)$ be the total running time of the $i$ th subsystem over the interval $[0, T)$.

Consider the following Lyapunov-Krasovskii functional for the $i$ th subsystem:

$$
\begin{equation*}
V_{i}(t, x(t))=x^{T}(t) v_{i}+\int_{t-d(t)}^{t} e^{\alpha_{i}(-t+s)} \boldsymbol{x}^{T}(s) v_{i} d s+\int_{-\tau}^{0} \int_{t+\theta}^{t} e^{\alpha_{i}(-t+s)} \boldsymbol{x}^{T}(s) \vartheta_{i} d s d \theta \tag{9}
\end{equation*}
$$

where $v_{i} \succ 0, v_{i} \succ 0$ and $\vartheta_{i} \succ 0$ are vectors to be determined.
For the sake of simplicity, $V_{i}(t, x(t))$ is written as $V_{i}(t)$ in this paper. Taking the derivation of the Lyapunov-Krasovskii functional along the trajectory of the $i$ th subsystem yields:

$$
\begin{aligned}
\dot{V}_{i}(t) & =\dot{x}^{T}(t) v_{i}-\alpha_{i} \int_{t-d(t)}^{t} e^{\alpha_{i}(-t+s)} x^{T}(s) v_{i} d s+x^{T}(t) v_{i}-(1-\dot{d}(t)) e^{-\alpha_{i} \tau} x^{T}(t-d(t)) v_{i} \\
& -\alpha_{i} \int_{-\tau}^{0} \int_{t+\theta}^{t} e^{\alpha_{i}(-t+s)} x^{T}(s) \vartheta_{i} d s d \theta+\tau x^{T}(t) \vartheta_{i}-\int_{-\tau}^{0} e^{\alpha_{i} \theta} x^{T}(t+\theta) \vartheta_{i} d \theta \\
& =-\alpha_{i} V_{i}(t)+\alpha_{i} x^{T}(t) v_{i}+\dot{x}^{T}(t) v_{i}+x^{T}(t) v_{i}-(1-\dot{d}(t)) e^{-\alpha_{i} \tau} x^{T}(t-d(t)) v_{i} \\
& +\tau x^{T}(t) \vartheta_{i}-\int_{t-\tau}^{t} e^{\alpha_{i}(s-t)} x^{T}(s) \vartheta_{i} d s \\
& \leqslant-\alpha_{i} V_{i}(t)+x^{T}(t)\left(\alpha_{i} v_{i}+A_{i}^{T} v_{i}+v_{i}+\tau \vartheta_{i}\right) \\
& +x^{T}(t-d(t))\left(G_{i}^{T} v_{i}-(1-d) e^{-\alpha_{i} \tau} v_{i}\right)-\int_{t-d(t)}^{t} e^{-\alpha_{i} \tau} x^{T}(s) \vartheta_{i} d s \\
& \leqslant-\alpha_{i} V_{i}(t)+x^{T}(t)\left(\alpha_{i} v_{i}+A_{i}^{T} v_{i}+v_{i}+\tau \vartheta_{i}\right) \\
& +x^{T}(t-d(t))\left(G_{i}^{T} v_{i}-(1-d) e^{-\alpha_{i} \tau} v_{i}\right)
\end{aligned}
$$

It can be obtained from (5) and (6) that

$$
\begin{equation*}
\dot{V}_{i}(t) \leqslant-\alpha_{i} V_{i}(t) \tag{10}
\end{equation*}
$$

It follows that

$$
\begin{equation*}
V_{\sigma(t)}(t) \leqslant e^{-\alpha_{\sigma(t)}\left(t-t_{\kappa}\right)} V_{\sigma(t)}\left(t_{\kappa}\right), \quad t \in\left[t_{\kappa}, t_{\kappa+1}\right) \tag{11}
\end{equation*}
$$

From (7) and (8), one has

$$
\begin{align*}
& V_{\sigma(T)}(T) \leqslant e^{-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma}(0, T)}\right)} V_{\sigma\left(t_{N_{\sigma}(0, T)}\right)}\left(t_{N_{\sigma}(0, T)}\right) \leqslant \mu_{\sigma(T)} e^{-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma}(0, T)}\right)} V_{\sigma\left(t_{N_{\sigma(0, T)}}\right)}\left(t_{N_{\sigma}(0, T)}^{-}\right) \\
& \leqslant \mu_{\sigma(T)} e^{-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma}(0, T)}\right)} e^{-\alpha_{\sigma\left(t_{N_{\sigma}(0, T)-1}\right)}\left(t_{N_{\sigma}(0, T)}-t_{\left.N_{\sigma}(0, T)-1\right)}\right.} V_{\sigma\left(t_{\left.N_{\sigma}(0, T)-1\right)}\right)}\left(t_{N_{\sigma(0, T)-1}}\right) \leqslant \cdots \\
& \left.\leqslant V_{\sigma(0)}(0) \prod_{s=0}^{N_{\sigma}(0, T)-1} \mu_{\sigma\left(t_{s+1}\right)} \exp \left\{\sum_{s=0}^{N_{\sigma}(0, T)-1}\left[-\alpha_{\sigma\left(t_{s}\right)}\left(t_{s+1}-t_{s}\right)\right]-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma}(0, T)}\right)\right\}\right) \\
& \left.\leqslant V_{\sigma(0)}(0) \prod_{i=1}^{N} \mu_{i}^{N_{\sigma i}(0, T)} \exp \left\{\sum_{i=1}^{N}\left[-\alpha_{i} \sum_{s \in \phi(i)}\left(t_{s+1}-t_{s}\right)\right]-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma(0, T)}}\right)\right\}\right) \\
& \leqslant \exp \left\{\sum_{i=1}^{N} N_{0 i} \ln \mu_{i}\right\} \exp \left\{\sum_{i=1}^{N} T_{i}(0, T) \ln \mu_{i} / T_{a i}-\sum_{i=1}^{N} \alpha_{i} T_{i}(0, T)\right\} V_{\sigma(0)}(0) \\
& =\exp \left\{\sum_{i=1}^{N} N_{0 i} \ln \mu_{i}\right\} \exp \left\{\sum_{i=1}^{N}\left(\ln \mu_{i} / T_{a i}-\alpha_{i}\right) T_{i}(0, T)\right\} V_{\sigma(0)}(0), \tag{12}
\end{align*}
$$

where $\phi(i)$ denotes the set of $s$ satisfying $\sigma\left(t_{s}\right)=i, t_{s} \in\left\{t_{0}, t_{1}, t_{2}, \ldots, t_{\kappa-1}, t_{\kappa}, t_{\kappa+1}, \ldots, t_{N_{\sigma(0, T)}}\right\}, \mu_{\sigma\left(t_{s+1}\right)}, \mu_{\sigma(T)} \in\left\{\mu_{1}, \mu_{2}, \ldots, \mu_{N}\right\}$, $\alpha_{\sigma(T)} \in\left\{\alpha_{1}, \alpha_{2}, \ldots, \alpha_{N}\right\}$.

Set $\varepsilon_{1}=\min _{(r, i) \in \underline{n} \times \underline{N}}\left\{v_{i r}\right\}, \varepsilon_{2}=\max _{(r, i) \in \underline{n} \times \underline{N}}\left\{v_{i r}\right\}, \varepsilon_{3}=\max _{(r, i) \in \underline{n} \times \underline{N}}\left\{v_{i r}\right\}$ and $\varepsilon_{4}=\max _{(r, i) \in \underline{n} \times \underline{N}}\left\{\vartheta_{i r}\right\}$, where $v_{i r}, v_{i r}$ and $\vartheta_{i r}$ represent the $r$ th elements of $v_{i}, v_{i}$ and $\vartheta_{i}$, respectively, $\underline{n}=\{1,2, \ldots, n\}$. Then, one obtains

$$
\begin{aligned}
& V_{\sigma(T)}(T) \geqslant \varepsilon_{1}\|x(T)\| \\
& V_{\sigma\left(t_{0}\right)}\left(t_{0}\right) \leqslant \varepsilon_{2}\left\|x\left(t_{0}\right)\right\|+\left(\varepsilon_{3} e^{-\tau \alpha_{\sigma\left(t_{0}\right)}}+\varepsilon_{4} \tau e^{\left.-\tau \alpha_{\sigma\left(t_{0}\right)}\right)} \int_{t_{0}-\tau}^{t_{0}}\|x(s)\| d s\right.
\end{aligned}
$$

It follows that

$$
\begin{aligned}
\|x(T)\| & \leqslant \frac{1}{\varepsilon_{1}} \exp \left\{\sum_{i=1}^{N} N_{0 i} \ln \mu_{i}\right\} \exp \left\{\sum_{i=1}^{N}\left(\ln \mu_{i} / T_{a i}-\alpha_{i}\right) T_{i}(0, T)\right\}\left(\varepsilon_{2}\left\|x\left(t_{0}\right)\right\|+\left(\varepsilon_{3} e^{-\alpha_{\sigma\left(t_{0}\right)} \tau}+\varepsilon_{4} \tau e^{-\alpha_{\sigma\left(t_{0}\right)} \tau}\right) \int_{t_{0}-\tau}^{t_{0}}\|x(s)\| d s\right) \\
& \leqslant \frac{1}{\varepsilon_{1}} \exp \left\{\sum_{i=1}^{N} N_{0 i} \ln \mu_{i}\right\}\left(\varepsilon_{2}+\varepsilon_{3} \tau e^{-\alpha_{\sigma\left(t_{0}\right)} \tau}+\varepsilon_{4} \tau^{2} e^{-\alpha_{\sigma\left(t_{0}\right)} \tau}\right) \exp \left\{\max _{i \in \underline{N}}\left(\ln \mu_{i} / T_{a i}-\alpha_{i}\right)\left(T-t_{0}\right)\right\} \sup _{-\tau \leqslant \theta \leqslant 0}\|\varphi(\theta)\|
\end{aligned}
$$

Set

$$
\begin{aligned}
& \zeta=\frac{1}{\varepsilon_{1}} \exp \left\{\sum_{i=1}^{N} N_{0 i} \ln \mu_{i}\right\}\left(\varepsilon_{2}+\varepsilon_{3} \tau e^{-\alpha_{\sigma\left(t_{0}\right)} \tau}+\varepsilon_{4} \tau^{2} e^{-\alpha_{\sigma\left(t_{0}\right)} \tau}\right), \\
& \rho=\min _{i \in \underline{N}}\left(\alpha_{i}-\ln \mu_{i} / T_{a i}\right) .
\end{aligned}
$$

It can be obtained from (7) that

$$
\begin{equation*}
\|x(T)\| \leqslant \zeta e^{-\rho\left(T-t_{0}\right)}\left\|x\left(t_{0}\right)\right\|_{c}, \forall T \geqslant t_{0} \tag{13}
\end{equation*}
$$

This completes the proof.

Remark 3. In Theorem 1, we get a delay-dependent stability criterion by utilizing the MDADT method instead of the ADT method [24]. In [24], the parameters $\lambda$ and $\mu$ are same for all subsystems, i.e., mode-independent. However, the parameters $\alpha_{i}$ and $\mu_{i}$ in this paper are mode-dependent, which would reduce the conservativeness existed in [24].

When $d(t) \equiv 0$, the SPLS (1) will generate to a delay-free system and the result in Theorem 1 reduces to the one proposed in Theorem 1 of [36].

Based on Theorem 1, we will present a stability result for the SPLS (3) with $w(t) \equiv 0$ by considering a class of LyapunovKrasovskii functionals allowed to increase with bounded increase rate during some intervals.

Theorem 2. Consider the SPLS (3) with $w(t) \equiv 0$. Let $\alpha_{i}>0$ and $\beta_{i}>0$ be given constants. If there exist vectors $v_{i} \succ 0, v_{i} \succ 0, v_{i} \succ 0$, $v_{i, j} \succ 0, v_{i, j} \succ 0$ and $\vartheta_{i, j} \succ 0$ of appropriate dimensions, such that, $\forall(i, j) \in \underline{N} \times \underline{N}, i \neq j$,

$$
\begin{align*}
& \bar{A}_{i}^{T} v_{i}+\alpha_{i} v_{i}+v_{i}+\tau \vartheta_{i} \preceq 0,  \tag{14}\\
& G_{i}^{T} v_{i}-(1-d) e^{-\alpha_{i} \tau} v_{i} \preceq 0,  \tag{15}\\
& \bar{A}_{i, j}^{T} v_{i, j}-\beta_{i} v_{i, j}+v_{i, j}+\tau \vartheta_{i, j} \preceq 0,  \tag{16}\\
& G_{i}^{T} v_{i, j}-(1-d) v_{i, j} \preceq 0, \tag{17}
\end{align*}
$$

then the system is exponentially stable for any switching signal $\sigma(t)$ with the following MDADT scheme

$$
\begin{equation*}
T_{a j}>T_{a j}^{*}=\left(\Delta_{m j}\left(\alpha_{j}+\beta_{j}\right)+\ln \left(\mu_{0 j} \mu_{1 j} \mu_{2 j}\right)\right) / \alpha_{j} \tag{18}
\end{equation*}
$$

where $\Delta_{m j}$ denotes the maximal delay period that the switching of the controller of the $j$ th subsystem lags behind that of the subsystem, $\mu_{0 j}=\mu_{j}=e^{\tau\left(\alpha_{j}+\beta_{j}\right)}$ and $\mu_{1 j} \mu_{2 j} \geqslant 1$ satisfy

$$
\begin{equation*}
v_{j} \preceq \mu_{1 j} v_{i, j}, v_{j} \preceq \mu_{1 j} v_{i, j}, \vartheta_{j} \preceq \mu_{1 j} \vartheta_{i, j}, \quad v_{i, j} \preceq \mu_{2 j} \mu_{0 j} v_{i}, v_{i, j} \preceq \mu_{2 j} v_{i}, \vartheta_{i, j} \preceq \mu_{2 j} \vartheta_{i} \tag{19}
\end{equation*}
$$

Proof. For any $T>0$, let $t_{0}=0$ and denote by $t_{1}, t_{2}, \ldots, t_{\kappa-1}, t_{\kappa}, \ldots, t_{N_{\sigma}(0, T)}$ the switching instants in the interval $\left[t_{0}, T\right)$, where $N_{\sigma}(0, T)=\sum_{i=1}^{N} N_{\sigma i}(0, T)$. Let $T_{i}(0, T)$ be the total running time of the $i$ th subsystem over the interval $[0, T)$.

Let the $i$ th subsystem be activated at $t_{\kappa-1}$ and the $j$ th subsystem be activated at $t_{\kappa},(i, j) \in \underline{N} \times \underline{N}, i \neq j$. Construct the following Lyapunov-Krasovskii functional for the SPLS (3):

$$
V(t)= \begin{cases}x^{T}(t) v_{i}+\int_{t-d(t)}^{t} e^{\alpha_{i}(-t+s)} x^{T}(s) v_{i} d s+\int_{-\tau}^{0} \int_{t+\theta}^{t} e^{\alpha_{i}(-t+s)} x^{T}(s) \vartheta_{i} d s d \theta, & \forall t \in\left[t_{\kappa-1}+\Delta_{\kappa-1}, t_{\kappa}\right)  \tag{20}\\ x^{T}(t) v_{i, j}+\int_{t-d(t)}^{t} e^{\beta_{j}(t-s)} x^{T}(s) v_{i, j} d s+\int_{-\tau}^{0} \int_{t+\theta}^{t} e^{\beta_{j}(t-s)} \boldsymbol{x}^{T}(s) \vartheta_{i, j} d s d \theta, & \forall t \in\left[t_{\kappa}, t_{\kappa}+\Delta_{\kappa}\right)\end{cases}
$$

When $w(t) \equiv 0$, by Theorem 1, we obtain from (14)-(17) that

$$
\dot{V}(t) \leqslant\left\{\begin{array}{l}
-\alpha_{i} V(t), \quad \forall t \in\left[t_{\kappa-1}+\Delta_{\kappa-1}, t_{\kappa}\right)  \tag{21}\\
\beta_{i} V(t), \quad \forall t \in\left[t_{\kappa-1}, t_{\kappa-1}+\Delta_{\kappa-1}\right)
\end{array}\right.
$$

From (19) and (20), at the instants $t_{\kappa}$ and $t_{\kappa}+\Delta_{\kappa}$, we have

$$
V\left(t_{\kappa}+\Delta_{\kappa}\right) \leqslant \mu_{1 j} V\left(\left(t_{\kappa}+\Delta_{\kappa}\right)^{-}\right), \quad V\left(t_{\kappa}\right) \leqslant \mu_{2 j} \mu_{j} V\left(t_{\kappa}^{-}\right)
$$

For $T \geqslant t_{N_{\sigma}(0, T)}+\Delta_{N_{\sigma}(0, T)}$, we obtain by induction that

$$
\begin{aligned}
V(T) \leqslant & e^{-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma}(0, T)}-\Delta_{N_{\sigma}(0, T)}\right) V\left(t_{N_{\sigma(0, T)}}+\Delta_{N_{\sigma}(0, T)}\right)} \\
\leqslant & \mu_{\sigma(T) 1} e^{-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma}(0, T)}-\Delta_{\left.N_{\sigma(0, T)}\right)} V\left(\left(t_{N_{\sigma}(0, T)}+\Delta_{N_{\sigma}(0, T)}\right)^{-}\right)\right.} \\
& \leqslant \mu_{\sigma(T) 1} e^{-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma}(0, T)-}-\Delta_{\left.N_{\sigma(0, T)}\right)} e^{\beta_{\sigma(T)} \Delta_{N_{\sigma}(0, T)}} V\left(t_{N_{\sigma}(0, T)}\right)\right.} \\
\leqslant & \mu_{\sigma(T)} \mu_{\sigma(T) 1} \mu_{\sigma(T) 2} e^{-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma}(0, T)}-\Delta_{\left.N_{\sigma(0, T)}\right)} e^{\beta_{\sigma(T)} \Delta_{N_{\sigma}(0, T)}} V\left(t_{N_{\sigma}(0, T)}^{-}\right)\right.} \\
\leqslant & \mu_{\sigma(T)} \mu_{\sigma(T) 1} \mu_{\sigma(T) 2} e^{-\alpha_{\sigma(T)}\left(T-t_{\left.N_{\sigma(0, T)}-\Delta_{m \sigma(T)}\right)} e^{\beta_{\sigma(T)} \Delta_{m \sigma(T)}} V\left(t_{N_{\sigma(0, T)}}^{-}\right)\right.} \\
\leqslant & \cdots \\
\leqslant & \left.\exp \left\{\sum_{i=1}^{N}-\alpha_{i} \sum_{s \in \phi(i)}\left(t_{s+1}-t_{s}-\Delta_{m i}\right)+\beta_{i} \sum_{s \in \phi(i)} \Delta_{m i}\right)\right\} \\
& V\left(t_{0}\right) \prod_{i=1}^{N}\left(\mu_{i} \mu_{i 1} \mu_{i 2}\right)^{N_{\sigma i}(0, T)} \\
& \exp \left\{\sum_{i=1}^{N} N_{0 i} \ln \left(\mu_{i} \mu_{i 1} \mu_{i 2}\right)\right\} \\
+ & \left.\sum_{i=1}^{N}\left(-\alpha_{i}\left(T_{i}(0, T)-\Delta_{m i} N_{\sigma i}(0, T)\right)+\beta_{i} \Delta_{m i} N_{\sigma i}(0, T)\right)\right\} V\left(t_{0}\right) \\
\leqslant & \exp \left\{\sum_{i=1}^{N} \mu_{i 1} \mu_{i 2}\right) T_{i}(0, T) / T_{a i} \\
& \exp \left\{\sum_{i=1}^{N}\left(\ln \left(\mu_{i} \mu_{i 1} \mu_{i 2}\right)+\left(\alpha_{i}+\beta_{i}\right) \Delta_{m i}\right)\right\}
\end{aligned}
$$

where $\phi(i)$ denotes the set of $s$ satisfying $\sigma\left(t_{s}\right)=i, t_{s} \in\left\{t_{0}, t_{1}, t_{2}, \ldots, t_{\kappa-1}, t_{\kappa}, t_{\kappa+1}, \ldots, t_{\left.N_{\sigma(0, T}\right\}}\right\}, \Delta_{m \sigma(T)} \in\left\{\Delta_{m 1}, \Delta_{m 2}, \ldots, \Delta_{m N}\right\}$, $\mu_{\sigma(T)} \in\left\{\mu_{1}, \mu_{2}, \ldots, \mu_{N}\right\}, \quad \mu_{\sigma(T) 1} \in\left\{\mu_{11}, \mu_{21}, \ldots, \mu_{N 1}\right\}, \quad \mu_{\sigma(T) 2} \in\left\{\mu_{12}, \mu_{22}, \ldots, \mu_{N 2}\right\}, \quad \alpha_{\sigma(T)} \in\left\{\alpha_{1}, \alpha_{2}, \ldots, \alpha_{N}\right\} \quad$ and $\quad \beta_{\sigma(T)} \in\left\{\beta_{1}, \beta_{2}\right.$, $\left.\ldots, \beta_{N}\right\}$.

It follows from (18) that $V(T)$ converges to zero as $T \rightarrow \infty$. Then the exponential stability of the SPLS (3) with $w(t) \equiv 0$ can be deduced by following the proof line of Theorem 1.

The proof is completed.

Remark 4. The proof of Theorem 2 is similar to the one of Theorem 1. Note that the Lyapunov-Krasovskii functional considered in Theorem 2 can be increasing both at switching instants and during the interval $\left[t_{\kappa-1}, t_{\kappa-1}+\Delta_{\kappa-1}\right.$ ). However, the possible increment will be compensated by the more specific decrement (by limiting the lower bound of MDADT), therefore, the system exponential stability is still guaranteed.

Now, we are in a position to consider the $L_{1}$-gain analysis for the SPLS (3).
Theorem 3. Consider the SPLS (3). Let $\alpha_{i}>0, \beta_{i}>0$ and $\gamma>0$ be given constants. If there exist vectors $v_{i} \succ 0, v_{i} \succ 0, \vartheta_{i} \succ 0, v_{i, j} \succ 0$, $v_{i, j} \succ 0$ and $\vartheta_{i, j} \succ 0$ of appropriate dimensions, $\forall(i, j) \in \underline{N} \times \underline{N}, i \neq j$, such that

$$
\begin{align*}
& \bar{A}_{i}^{T} v_{i}+\alpha_{i} v_{i}+v_{i}+\tau \vartheta_{i}+C_{i}^{T} 1_{q} \preceq 0,  \tag{22}\\
& \bar{A}_{i, j}^{T} v_{i, j}-\beta_{i} v_{i, j}+v_{i, j}+\tau \vartheta_{i, j}+C_{j}^{T} 1_{q} \preceq 0,  \tag{23}\\
& G_{i}^{T} v_{i}-(1-d) e^{-\alpha_{i} \tau} v_{i} \preceq 0,  \tag{24}\\
& G_{j}^{T} v_{i, j}-(1-d) v_{i, j} \preceq 0,  \tag{25}\\
& E_{i}^{T} v_{i}+D_{i}^{T} 1_{q}-\gamma 1_{f} \preceq 0,  \tag{26}\\
& E_{j}^{T} v_{i, j}+D_{j}^{T} 1_{q}-\gamma 1_{f} \preceq 0, \tag{27}
\end{align*}
$$

then the system is exponentially stable and has a prescribed $L_{1}$-gain performance level $\gamma$ for any switching signal with the MDADT scheme (18), where $\mu_{0 j}, \mu_{1 j}$ and $\mu_{2 j}$ satisfy (19).

Proof. Choose the Lyapunov-Krasovskii functional (20) for the SPLS (3). By Theorem 2, the exponential stability of the SPLS (3) with $w(t) \equiv 0$ is ensured by (22)-(25).

We are now in a position to consider the $L_{1}$-gain performance.

Define $\Gamma(t)=\|z(t)\|-\gamma\|w(t)\|$. When $w(t) \neq 0$, it follows from (22)-(27) that

$$
\dot{V}(t) \leqslant\left\{\begin{array}{l}
-\alpha_{i} V(t)-\Gamma(t), \quad \forall t \in\left[t_{\kappa-1}+\Delta_{\kappa-1}, t_{\kappa}\right)  \tag{28}\\
\beta_{i} V(t)-\Gamma(t), \quad \forall t \in\left[t_{\kappa-1}, t_{\kappa-1}+\Delta_{\kappa-1}\right)
\end{array}\right.
$$

Then, integrating both sides of (28), we have

$$
V(t) \leqslant\left\{\begin{array}{l}
e^{-\alpha_{i}\left(t-t_{\kappa-1}-\Delta_{\kappa-1}\right)} V\left(t_{\kappa-1}+\Delta_{\kappa-1}\right)-\int_{t_{\kappa-1}+\Delta_{\kappa-1}}^{t} e^{-\alpha_{i}(t-s)} \Gamma(s) d s, \quad \forall t \in\left[t_{\kappa-1}+\Delta_{\kappa-1}, t_{\kappa}\right)  \tag{29}\\
e^{\beta_{i}\left(t-t_{\kappa-1}\right)} V\left(t_{\kappa-1}\right)-\int_{t_{\kappa-1}}^{t} e^{\beta_{i}(t-s)} \Gamma(s) d s, \quad \forall t \in\left[t_{\kappa-1}, t_{\kappa-1}+\Delta_{\kappa-1}\right)
\end{array}\right.
$$

where $\Gamma(s)=\|z(s)\|-\gamma\|w(s)\|$.
For $T \geqslant t_{N_{\sigma}(0, T)}+\Delta_{N_{\sigma}(0, T)}$, by Definition 4 , (19) and (29), we can obtain by induction that

$$
\begin{aligned}
& V(T) \leqslant e^{-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma}(0, T)}-\Delta_{N_{\sigma}(0, T)}\right)} V\left(t_{N_{\sigma}(0, T)}+\Delta_{N_{\sigma}(0, T)}\right) \\
& -\int_{t_{N_{\sigma}(0, T)}+\Delta_{N_{\sigma}(0, T)}}^{T} e^{-\alpha_{\sigma(T)}(T-s)} \Gamma(s) d s \\
& \leqslant \mu_{\sigma(T) 1} e^{-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma}(0, T)}-\Delta_{N_{\sigma}(0, T)}\right)} V\left(\left(t_{N_{\sigma}(0, T)}+\Delta_{N_{\sigma}(0, T)}\right)^{-}\right) \\
& -\int_{t_{N_{\sigma}(0, T)}+\Delta_{N_{\sigma}(0, T)}}^{T} e^{-\alpha_{\sigma(T)}(T-s)} \Gamma(s) d s \\
& \leqslant \mu_{\sigma(T) 1} e^{-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma}(0, T)}-\Delta_{N_{\sigma}(0, T)}\right)}\left\{e^{\beta_{\sigma(T)} \Delta_{N_{\sigma}(0, T)}} V\left(t_{N_{\sigma}(0, T)}\right)\right. \\
& \left.-\int_{t_{N_{\sigma}(0, T)}}^{t_{N_{\sigma}(0, T)}+\Delta_{N_{\sigma}(0, T)}} e^{\beta_{\sigma(T)}\left(t_{N_{\sigma}(0, T)}+\Delta_{N_{\sigma}(0, T)}-s\right)} \Gamma(s) d s\right\} \\
& -\int_{t_{N_{\sigma}(0, T)}+\Delta_{N_{\sigma}(0, T)}}^{T} e^{-\alpha_{\sigma(T)}(T-s)} \Gamma(s) d s \\
& \leqslant \mu_{\sigma(T)} \mu_{\sigma(T) 1} \mu_{\sigma(T) 2} e^{-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma}(0, T)}-\Delta_{N_{\sigma}(0, T)}\right.} e^{\beta_{\sigma(T)} \Delta_{N_{\sigma}(0, T)}} V\left(t_{N_{\sigma}(0, T)}^{-}\right) \\
& -\mu_{\sigma(T) 1} e^{-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma}(0, T)}-\Delta_{N_{\sigma}(0, T)}\right)} \int_{t_{N_{\sigma}(0, T)}}^{t_{N_{\sigma}(0, T)}+\Delta_{N_{\sigma}(0, T)}} e^{\beta_{\sigma(T)}\left(t_{N_{\sigma}(0, T)}+\Delta_{N_{\sigma}(0, T)}-s\right)} \Gamma(s) d s \\
& -\int_{t_{N_{\sigma}(0, T)}+\Delta_{N_{\sigma}(0, T)}}^{T} e^{-\alpha_{\sigma(T)}(T-s)} \Gamma(s) d s \\
& \leqslant \mu_{\sigma(T)} \mu_{\sigma(T) 1} \mu_{\sigma(T) 2} e^{-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma}(0, T)}-\Delta_{m \sigma(T)}\right.} e^{\beta_{\sigma(T)} \Delta_{m \sigma(T)}} V\left(t_{N_{\sigma}(0, T)}^{-}\right) \\
& -\int_{t_{N \sigma(0, T)}+\Delta_{N \sigma(0, T)}}^{T} e^{-\alpha_{\sigma(T)}(T-s)} \Gamma(s) d s \\
& \left.-\mu_{\sigma(T) 1} e^{-\alpha_{\sigma(T)}\left(T-t_{N_{\sigma}(0, T)}-\Delta_{N_{\sigma}(0, T)}\right)} \int_{t_{N_{\sigma}(0, T)}}^{t_{N_{\sigma}(0, T)}+\Delta_{N_{\sigma}(0, T)}} e^{\beta_{\sigma(T)}\left(t_{N_{\sigma}(0, T)}+\Delta_{N_{\sigma}(0, T)}-s\right)} \Gamma(s) d s\right\} \\
& \leqslant \cdots \\
& \leqslant \exp \left\{\sum_{i=1}^{N} N_{0 i} \ln \left(\mu_{i} \mu_{i 1} \mu_{i 2}\right)\right\} \\
& \exp \left\{\sum_{i=1}^{N} \ln \left(\mu_{i} \mu_{i 1} \mu_{\mathrm{i} 2}\right) T_{i}(0, T) / T_{a i}\right. \\
& \left.+\sum_{i=1}^{N}\left(-\alpha_{i}\left(T_{i}(0, T)-N_{\sigma i}(0, T) \Delta_{m i}\right)+\beta_{i} N_{\sigma i}(0, T) \Delta_{m i}\right)\right\} V\left(t_{0}\right) \\
& -\int_{t_{0}}^{T} e^{\left\{\sum_{i=1}^{N}-\alpha_{i}\left(T_{i}(s, T)-N_{\sigma i}(s, T) \Delta_{m i}\right)+\beta_{i} N_{\sigma i}(s, T) \Delta_{m i}\right\}} \Gamma(s) \prod_{i=1}^{N}\left(\mu_{i} \mu_{i 1} \mu_{i 2}\right)^{N_{\sigma i}(s, T)} d s
\end{aligned}
$$

where $\phi(i)$ denotes the set of $s$ satisfying $\sigma\left(t_{s}\right)=i, t_{s} \in\left\{t_{0}, t_{1}, t_{2}, \ldots, t_{\kappa-1}, t_{\kappa}, t_{\kappa+1}, \ldots, t_{N_{\sigma(0, T)}}\right\}$,
$\Delta_{m \sigma(T)} \in\left\{\Delta_{m 1}, \Delta_{m 2}, \ldots, \Delta_{m N}\right\}, \quad \mu_{\sigma(T)} \in\left\{\mu_{1}, \mu_{2}, \ldots, \mu_{N}\right\}, \quad \mu_{\sigma(T) 1} \in\left\{\mu_{11}, \mu_{21}, \ldots, \mu_{N 1}\right\}, \quad \mu_{\sigma(T) 2} \in\left\{\mu_{12}, \mu_{22}, \ldots, \mu_{N 2}\right\}, \quad \alpha_{\sigma(T)} \in$ $\left\{\alpha_{1}, \alpha_{2}, \ldots, \alpha_{N}\right\}$ and $\beta_{\sigma(T)} \in\left\{\beta_{1}, \beta_{2}, \ldots, \beta_{N}\right\}$.

Under the zero initial condition, one has

$$
\begin{equation*}
0 \leqslant-\int_{t_{0}}^{T} e^{\left\{\sum_{i=1}^{N}-\alpha_{i}\left(T_{i}(s, T)-N_{\sigma i}(s, T) \Delta_{m i}\right)+\beta_{i} N_{\sigma i}(s, T) \Delta_{m i}\right\}} \Gamma(s) \prod_{i=1}^{N}\left(\mu_{i} \mu_{i 1} \mu_{i 2}\right)^{N_{\sigma i}(s, T)} \tag{30}
\end{equation*}
$$

That is

$$
\begin{align*}
& \int_{t_{0}}^{T} e^{\left\{\sum_{i=1}^{N}-\alpha_{i}\left(T_{i}(s, T)-N_{\sigma i}(s, T) \Delta_{m i}\right)+\beta_{i} N_{\sigma i}(s, T) \Delta_{m i}\right\}}\|z(s)\| \prod_{i=1}^{N}\left(\mu_{i} \mu_{i 1} \mu_{i 2}\right)^{N_{\sigma i}(s, T)} d s \\
& \quad \leqslant \gamma \int_{t_{0}}^{T} e^{\left\{\sum_{i=1}^{N}-\alpha_{i}\left(T_{i}(s, T)-N_{\sigma i}(s, T)\right)+\beta_{i} N_{\sigma i}(s, T)\right\}}\|w(s)\| \prod_{i=1}^{N}\left(\mu_{i} \mu_{i 1} \mu_{i 2}\right)^{N_{\sigma i}(s, T)} d s \tag{31}
\end{align*}
$$

Multiplying both sides of (31) by $e^{-\sum_{i=1}^{N}\left(\alpha_{i}+\beta_{i}\right) N_{\sigma i}\left(t_{0}, T\right) \Delta_{m i}} \prod_{i=1}^{N}\left(\mu_{i} \mu_{i 1} \mu_{i 2}\right)^{-N_{\sigma i}\left(t_{0}, T\right)}$ yields

$$
\begin{equation*}
\int_{t_{0}}^{T} e^{\sum_{i=1}^{N}\left\{-\alpha_{i} T_{i}(s, T)-\left(\alpha_{i}+\beta_{i}\right) N_{\sigma i}\left(t_{0}, s\right) \Delta_{m i}-N_{\sigma i}\left(t_{0}, s\right) \ln \left(\mu_{i} \mu_{i 1} \mu_{i 2}\right)\right\}}\|z(s)\| d s \leqslant \gamma \int_{t_{0}}^{T} e^{\sum_{i=1}^{N}\left\{-\alpha_{i} T_{i}(s, T)-\left(\alpha_{i}+\beta_{i}\right) N_{\sigma i}\left(t_{0}, s\right) \Delta_{m i}\right\}}\|w(s)\| d s \tag{32}
\end{equation*}
$$

It follows from (18) that

$$
\begin{equation*}
\int_{t_{0}}^{T} e^{\sum_{i=1}^{N}\left\{-\alpha_{i} T_{i}(s, T)-\left(\left(\alpha_{i}+\beta_{i}\right) \Delta_{m i}+\ln \left(\mu_{i} \mu_{i 1} \mu_{i 2}\right)\right) T_{i}\left(t_{0}, s\right) / T_{a i}\right\}}\|z(s)\| d s \leqslant \gamma \int_{t_{0}}^{T} e^{\sum_{i=1}^{N}-\alpha_{i} T_{i}(s, T)}\|w(s)\| d s \tag{33}
\end{equation*}
$$

Integrating both sides of (33) from $T=t_{0}$ to $\infty$ leads to

$$
\begin{equation*}
\int_{t_{0}}^{\infty} e^{-\sum_{i=1}^{N}\left\{\alpha_{i} T_{i}\left(t_{0}, t\right)\right\}}\|z(t)\| d t \leqslant \gamma \int_{t_{0}}^{\infty}\|w(t)\| d t \tag{34}
\end{equation*}
$$

This means that system (3) achieves a prescribed $L_{1}$-gain performance level $\gamma$.
This completes the proof.

### 3.2. Asynchronous $L_{1}$ control

Based on the obtained stability and $L_{1}$-gain analysis results, the following theorem presents sufficient conditions for the existence of a state-feedback controller for the SPLS (1) in the presence of asynchronous switching such that the corresponding closed-loop system (3) is positive and exponentially stable with an $L_{1}$-gain performance level $\gamma$.

Theorem 4. Consider the SPLS (1). Let $\alpha_{i}>0, \beta_{i}>0$ and $\gamma>0$ be given constants. If there exist vectors $v_{i} \succ 0, v_{i} \succ 0, \vartheta_{i} \succ 0, v_{i, j} \succ 0$, $v_{i, j} \succ 0, \vartheta_{i, j} \succ 0$, and any matrices $K_{i}$ of appropriate dimensions, $\forall(i, j) \in \underline{N} \times \underline{N}, i \neq j$, such that $\bar{A}_{i}=A_{i}+B_{i} K_{i}$ and $\bar{A}_{i, j}=A_{j}+B_{j} K_{i}$ are Metzler matrices, and (24)-(27) and the following inequalities hold

$$
\begin{align*}
& A_{i}^{T} v_{i}+h_{i}+\alpha_{i} v_{i}+v_{i}+\tau \vartheta_{i}+C_{i}^{T} 1_{q} \preceq 0,  \tag{35}\\
& A_{i, j}^{T} v_{i, j}+K_{i}^{T} B_{j}^{T} v_{i, j}-\beta_{i} v_{i, j}+v_{i, j}+\tau \vartheta_{i, j}+C_{j}^{T} 1_{q} \preceq 0, \tag{36}
\end{align*}
$$

where $h_{i} \succeq K_{i}^{T} B_{i}^{T} v_{i}$, then the resulting closed-loop system (3) is positive and exponentially stable with an $L_{1}$-gain performance level $\gamma$ for any switching signal with the MDADT scheme (18), where $\mu_{0 j}, \mu_{1 j}$ and $\mu_{2 j}$ satisfy (19).

Proof. Upon introducing vectors $h_{i}$ satisfying $h_{i} \succeq K_{i}^{T} B_{i}^{T} v_{i}$, and substituting them into (22), the theorem can be directly obtained from Theorem 3.

This completes the proof.

Remark 5. Differently from the result in [35], we get sufficient conditions for the existence of an $L_{1}$-gain performance level. Also, the result proposed in Theorem 4 is derived via the MDADT approach, which is different from those adopted in [33-35]. The parameters $\alpha_{i}$ and $\beta_{i}$ are mode-dependent, which brings more flexibility to find feasible controllers. On the other hand, our result can cover the result of [24] as a special case, where the asynchronous switching is not considered.

Remark 6. It is noticed that (24), (25), (26), (27), (35) and (36) are mutually dependent. We can firstly solve (24), (26) and (35) to obtain the vectors $v_{i}, v_{i}, \vartheta_{i}, h_{i}$. Then we can get $K_{i}$ by $h_{i} \succeq K_{i}^{T} B_{i}^{T} v_{i}$. By substituting the obtained $K_{i}$ into (36), and solving (25), (27) and (36), we can obtain these vectors $v_{i j}, v_{i, j}, \vartheta_{i j}$. In addition, it can be seen that a smaller $\alpha_{i}$ will be favorable to the feasibility of (24), (26) and (35), and a larger $\beta_{i}$ will be favorable to the feasibility of (25), (27) and (36). In view of these, we put forward the following algorithm to obtain $K_{i}$.

## Algorithm 1.

Step (1) For each $i \in \underline{N}$, choose a $\alpha_{i}$ (For the first time, we can choose a larger $\alpha_{i}$ ), and solve (24), (26) and (35).
Step (2) If (24), (26) and (35) are unfeasible, then decrease $\alpha_{i}$ appropriately, and go to Step (1).

Step (3) If there exists a feasible solution, then get $v_{i}, v_{i}, \vartheta_{i}, h_{i}$. By $h_{i} \succeq K_{i}^{T} B_{i}^{T} v_{i}$, find a $K_{i}$ such that $\bar{A}_{i}=A_{i}+B_{i} K_{i}$ and $\bar{A}_{i, j}=A_{j}+B_{j} K_{i}$ are Metzler matrices, and then substitute it into (36).
Step (4) Choose a $\beta_{i}$ (For the first time, we can choose a smaller $\beta_{i}$ ), and solve (25), (27) and (36).
Step (5) If (25), (27) and (36) are unfeasible, then increase $\beta_{i}$ appropriately, and go to Step (4).
Step (6) If there exists a feasible solution, then get $v_{i j}, v_{i j}, \vartheta_{i j}$, and compute $T_{a j}^{*}$ by (18) and (19).

## 4. Numerical examples

In this section, two examples will be presented to demonstrate the potential and validity of our developed theoretical results.

Example 1. Consider the switched linear systems consisting of two positive subsystems described by:

$$
\begin{aligned}
& A_{1}=\left[\begin{array}{ccc}
-0.5302 & 0.0012 & 0.0873 \\
0.2185 & -0.7494 & 0.5411 \\
0.7370 & 0.1543 & -0.3606
\end{array}\right], \quad A_{2}=\left[\begin{array}{ccc}
-0.5136 & 0.4419 & 0.3689 \\
0.1840 & -0.3951 & 0.0080 \\
0.3163 & 0.6099 & -1.0056
\end{array}\right], \\
& G_{1}=\left[\begin{array}{ccc}
0.01 & 0.001 & 0 \\
0 & 0.01 & 0.1 \\
0.05 & 0 & 0.01
\end{array}\right], \quad G_{2}=\left[\begin{array}{ccc}
0.012 & 0 & 0 \\
0.014 & 0.01 & 0 \\
0 & 0 & 0.01
\end{array}\right] .
\end{aligned}
$$

It is obvious that the trajectories of such a switched system will remain positive if $x(0) \succeq 0$. Our purpose here is to find the admissible switching signals with MDADT such that the system is exponentially stable.

To illustrate the advantages of the proposed MDADT switching, we shall also present the design results of switching signals for the system with ADT switching for the sake of comparison. By different approaches and setting the relevant parameters appropriately, the computation results for the system with two different switching schemes are listed in Table 1.

It can be seen from Table 1 that the minimal MDADT are reduced to $T_{a 1}^{*}=6.8663, T_{a 2}^{*}=7.8472$ for given $\mu_{1}=\mu_{2}=3$, and one special case of MDADT switching is $T_{a 1}^{*}=T_{a 2}^{*}=7.8472$ by setting $\alpha_{1}=\alpha_{2}=0.14$, which is the ADT switching, i.e., the designed MDADT switching is more general.

Example 2. Consider system (1) with parameters as follows

$$
\begin{aligned}
& A_{1}=\left[\begin{array}{cc}
-1 & 7 \\
8.5 & -2.5
\end{array}\right], \quad G_{1}=\left[\begin{array}{ll}
0.1 & 0.2 \\
0.3 & 0.1
\end{array}\right], \quad B_{1}=\left[\begin{array}{l}
0.2 \\
0.4
\end{array}\right], \quad E_{1}=\left[\begin{array}{l}
0.5 \\
0.2
\end{array}\right], \\
& C_{1}=\left[\begin{array}{ll}
0.1 & 0.3
\end{array}\right], \quad D_{1}=0.3 \\
& A_{2}=\left[\begin{array}{cc}
-6.8 & 3.5 \\
9.3 & -6.6
\end{array}\right], \quad G_{2}=\left[\begin{array}{ll}
0.2 & 0.1 \\
0.1 & 0.2
\end{array}\right], \quad B_{2}=\left[\begin{array}{l}
0.1 \\
0.3
\end{array}\right], \quad E_{2}=\left[\begin{array}{l}
0.3 \\
0.4
\end{array}\right], \\
& C_{2}=\left[\begin{array}{ll}
0.2 & 0.4
\end{array}\right], \quad D_{2}=0.2
\end{aligned}
$$

By Lemma 1, the trajectories of such a switched system will obviously remain positive if $\varphi(\theta) \succeq 0, \theta \in[-\tau, 0]$. Our purpose here is to design a set of stabilizing controllers and find the admissible switching signals with MDADT such that the resulting closed-loop system is exponentially stable with an $L_{1}$ disturbance attenuation performance level in the presence of asynchronous switching.

Taking $\Delta_{m 1}=1.0, \Delta_{m 2}=0.5, \alpha_{1}=0.4, \alpha_{2}=0.3, \tau=0.1, d=0.1$ and $\gamma=1$, and solving (24), (26) and (35) in Theorem 4 give rise to

Table 1
Computation results for the system with two different switching schemes

|  | ADT switching [38] | MDADT switching |
| :---: | :---: | :---: |
| Feasible solutions | $\begin{aligned} & v_{1}=\left[\begin{array}{lll} 70.6769 & 26.0675 & 27.0391 \end{array}\right]^{T}, \\ & v_{2}=\left[\begin{array}{lll} 24.3130 & 75.9861 & 11.9570 \end{array}\right]^{T}, \\ & v_{1}=\left[\begin{array}{lll} 0.6826 & 1.2278 & 0.6271 \end{array}\right]^{T}, \\ & v_{2}=\left[\begin{array}{lll} 0.4213 & 0.6719 & 0.3204 \end{array}\right]^{T}, \\ & \vartheta_{1}=\left[\begin{array}{lll} 7.1928 & 10.0483 & 6.9808 \end{array}\right]^{T}, \\ & \vartheta_{2}=\left[\begin{array}{lll} 5.0049 & 5.5576 & 3.9734 \end{array}\right]^{T} \end{aligned}$ | $\begin{aligned} & v_{1}=\left[\begin{array}{lll} 64.7231 & 22.8869 & 24.0552 \end{array}\right]^{T}, \\ & v_{2}=\left[\begin{array}{lll} 21.8226 & 67.5209 & 10.5841 \end{array}\right]^{T}, \\ & v_{1}=\left[\begin{array}{lll} 0.4451 & 1.0511 & 0.3853 \end{array}\right]^{T}, \\ & v_{2}=\left[\begin{array}{lll} 0.4103 & 0.5137 & 0.2134 \end{array}\right]^{T}, \\ & \vartheta_{1}=\left[\begin{array}{lll} 5.3587 & 9.0788 & 5.1133 \end{array}\right]^{T}, \\ & \vartheta_{2}=\left[\begin{array}{lll} 4.9672 & 4.7610 & 3.1748 \end{array}\right]^{T} \end{aligned}$ |
| Switching parameters | $\begin{aligned} & \lambda=0.14, \mu=3 \\ & \tau_{a}^{*}=7.8472 \end{aligned}$ | $\begin{aligned} & \alpha_{1}=0.16, \quad \alpha_{2}=0.14, \quad \mu_{1}=\mu_{2}=3 \\ & T_{a 1}^{*}=6.8663, T_{a 2}^{*}=7.8472 \end{aligned}$ |

$$
\begin{aligned}
& v_{1}=\left[\begin{array}{l}
0.5757 \\
0.7741
\end{array}\right], \quad v_{2}=\left[\begin{array}{l}
0.7487 \\
0.5872
\end{array}\right], \quad v_{1}=\left[\begin{array}{l}
1.2302 \\
1.1821
\end{array}\right], \quad v_{2}=\left[\begin{array}{l}
1.1858 \\
1.2650
\end{array}\right] \\
& \vartheta_{1}=\left[\begin{array}{l}
1.0186 \\
1.0186
\end{array}\right], \quad \vartheta_{2}=\left[\begin{array}{l}
1.0186 \\
1.0186
\end{array}\right], \quad h_{1}=\left[\begin{array}{l}
-8.6846 \\
-5.0069
\end{array}\right], \quad h_{2}=\left[\begin{array}{l}
-3.1003 \\
-1.7068
\end{array}\right] .
\end{aligned}
$$

By $h_{i} \succeq K_{i}^{T} B_{i}^{T} v_{i}$, the state feedback gain matrices can be obtained as follows:

$$
K_{1}=\left[\begin{array}{ll}
-20.4460 & -11.7876
\end{array}\right], \quad K_{2}=\left[\begin{array}{ll}
-16.1223 & -8.8759
\end{array}\right] .
$$

Obviously, $\bar{A}_{i}=A_{i}+B_{i} K_{i}$ and $\bar{A}_{i, j}=A_{j}+B_{j} K_{i}$ are Metzler matrices. Then, choosing $\beta_{1}=0.5$ and $\beta_{2}=0.6$, and solving (25), (27) and (36), we obtain

$$
\begin{aligned}
& v_{2,1}=\left[\begin{array}{l}
0.7916 \\
0.9499
\end{array}\right], \quad v_{1,2}=\left[\begin{array}{l}
0.4309 \\
0.3613
\end{array}\right], \quad v_{2,1}=\left[\begin{array}{l}
0.9653 \\
0.9630
\end{array}\right], \quad v_{1,2}=\left[\begin{array}{l}
0.9921 \\
1.0477
\end{array}\right], \\
& v_{2,1}=\left[\begin{array}{l}
0.8413 \\
0.8509
\end{array}\right], \quad v_{1,2}=\left[\begin{array}{l}
0.8680 \\
0.8689
\end{array}\right] .
\end{aligned}
$$

Then, according to (19), we have $\mu_{11}=1.2744, \mu_{21}=1.7375, \mu_{01}=1.0942, \mu_{12}=1.6187, \mu_{22}=0.8863$ and $\mu_{02}=1.0942$. From (18), it can be obtained that $T_{a 1}^{*}=4.4623$ and $T_{a 2}^{*}=3.0031$. Choosing $T_{a 1}=4.5$ and $T_{a 2}=3.1$, simulation results of the closedloop systems are shown in Figs. 1 and 2, where the initial conditions are $x(0)=\left[\begin{array}{ll}0.1 & 0.2\end{array}\right]^{T}$, and $x(t)=\left[\begin{array}{ll}0 & 0\end{array}\right]^{T}, t \in\left[\begin{array}{ll}-0.1 & 0\end{array}\right.$. It can be seen that the closed-loop system is positive and exponentially stable, which indicates that the proposed method is effective.


Fig. 1. Switching signal.


Fig. 2. State responses of the closed-loop system.

## 5. Conclusions

Using the MDADT approach, the stabilization problem of positive switched systems with time-varying delays under asynchronous switching has been investigated in this paper. We have designed a feedback controller and a class of switching signals such that the closed-loop system is exponentially stable and has a prescribed $L_{1}$-gain performance in presence of asynchronous switching. Our future work will focus on the $L_{1}$ fault detection observer design for positive switched systems under asynchronous switching.

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## References

[1] S.B. Attia, S. Salhi, M. Ksouri, Static switched output feedback stabilization for linear discrete-time switched systems, Int. J. Innovative Comput. Inf. and Control 8 (5A) (2012) 3203-3213.
[2] F. Blanchini, P. Colaneri, M.E. Valcher, Co-positive Lyapunov functions for the stabilization of positive switched systems, IEEE Trans. Autom. Control 57 (12) (2012) 3038-3050.
[3] M. Bolajraf, F. Tadeo, T. Alvarez, M. Ait Rami, State-feedback with memory for controlled positivity with application to congestion control, IET Control Theory Appl. 4 (10) (2010) 2041-2048.
[4] E.R. Carson, C. Cobelli, L. Finkelstein, Modeling and identification of metabolic systems, Am. J. Physiol. 240 (3) (1981) 120-129.
[5] H. Caswell, Matrix Population Models: Construction, Analysis and Interpretation., Sinauer Associates, Sunderland, MA, 2001.
[6] X. Chen, J. Lam, P. Li, Z. Shu, $l_{1}$-Induced norm and controller synthesis of positive systems, Automatica 49 (5) (2013) 1377-1385.
[7] Z. Chen, K. Yamada, T. Sakanushi, Y. Zhao, Linear matrix inequality-based repetitive controller design for linear systems with time-varying uncertainties to reject position-dependent disturbances, Int. J. Innovative Comput. Inf. Control 9 (8) (2013) 3241-3256.
[8] E. Fornasini, M.E. Valcher, Linear copositive Lyapunov functions for continuous-time positive switched systems, IEEE Trans. Autom. Control 55 (8) (2010) 1933-1937.
[9] E. Fornasini, M.E. Valcher, Stability and stabilizability criteria for discrete-time positive switched systems, IEEE Trans. Autom. Control 57 (5) (2012) 1208-1221.
[10] C.A. Ibanez, M.S. Suarez-Castanon, O.O. Gutierrez-Frias, A switching controller for the stabilization of the damping inverted pendulum cart system, Int. J. Innovative Comput. Inf. Control 9 (9) (2013) 3585-3597.
[11] Q. Liu, W. Wang, D. Wang, New results on model reduction for discrete-time switched systems with time delay, Int. J. Innovative Comput. Inf. Control 8 (5A) (2012) 3431-3440.
[12] X. Liu, C. Dang, Stability analysis of positive switched linear systems with delays, IEEE Trans. Autom. Control 56 (7) (2011) 1684-1690.
[13] O. Mason, R. Shorten, On linear copositive Lyapunov functions and the stability of switched positive linear systems, IEEE Trans. Autom. Control 52 (7) (2007) 1346-1349.
[14] R. Shorten, D. Leith, J. Foy, R. Kilduff, Towards an analysis and design framework for congestion control in communication networks, in: Proceedings of 12th Yale Workshop Adaptive Learning Systems, 2003.
[15] R. Shorten, F. Wirth, D. Leith, A positive systems model of TCP-like congestion control: asymptotic results, IEEE/ACM Trans. Networking 14 (3) (2006) 616-629.
[16] X. Su, P. Shi, L. Wu, Y.D. Song, A novel approach to filter design for T-S fuzzy discrete-time systems with time-varying delay, IEEE Trans. Fuzzy Syst. 20 (6) (2012) 1114-1129.
[17] X. Su, P. Shi, L. Wu, Y.D. Song, A novel control design on discrete-time Takagi-Sugeno fuzzy systems with time-varying delays, IEEE Trans. Fuzzy Syst. 21 (4) (2013) 655-671.
[18] X. Su, L. Wu, P. Shi, Senor networks with random link failures: distributed filtering for T-S fuzzy systems, IEEE Trans. Ind. Inf. 9(3) (2013) 1739-1750.
[19] L. Wu, X. Su, P. Shi, Output feedback control of Markovian jump repeated scalar nonlinear systems, IEEE Trans. Autom. Control (2013), http:// dx.doi.org/10.1109/TAC.2013.2267353.
[20] L. Wu, W.X. Zheng, Weighted $H_{\infty}$ model reduction for linear switched systems with time-varying delay, Automatica 45 (1) (2009) $186-193$.
[21] L. Wu, W.X. Zheng, H. Gao, Dissipativity-based sliding mode control of switched stochastic systems, IEEE Trans. Autom. Control 58 (3) (2013) 785-791.
[22] Z. Xiang, Y.N. Sun, Q. Chen, Robust reliable stabilization of uncertain switched neutral systems with delayed switching, Appl. Math. Comput. 217 (23) (2011) 9835-9844.
[23] Z. Xiang, R. Wang, Robust control for uncertain switched non-linear systems with time delay under asynchronous switching, IET Control Theory Appl. 3 (8) (2009) 1041-1050.
[24] M. Xiang, Z. Xiang, Stability, $L_{1}$-gain and control synthesis for positive switched systems with time-varying delay, Nonlinear Anal.: Hybrid Syst. 9 (1) (2013) 9-17.
[25] W. Xiang, J. Xiao, $H_{\infty}$ filtering for switched nonlinear systems under asynchronous switching, Int. J. Syst. Sci. 42 (5) (2011) $751-765$.
[26] W. Xiang, J. Xiao, M.N. Iqbal, Fault detection for switched nonlinear systems under asynchronous switching, Int. J. Control 84 (8) (2011) $1362-1376$.
[27] G. Xie, L. Wang, Stabilization of switched linear systems with time-delay in detection of switching signal, J. Math. Anal. Appl. 305 (6) (2005) 277-290.
[28] R. Yang, H. Gao, P. Shi, Delay-dependent robust $H_{\infty}$ control for uncertain stochastic time-delay systems, Int. J. Robust Nonlinear Control 20 (16) (2010) 1852-1865.
[29] R. Yang, P. Shi, G.P. Liu, Filtering for discrete-time networked nonlinear systems with mixed random delays and packet dropouts, IEEE Trans. Autom. Control 56 (11) (2011) 2655-2660.
[30] R. Yang, P. Shi, G.P. Liu, H. Gao, Network-based feedback control for systems with mixed delays based on quantization and dropout compensation, Automatica 47 (12) (2011) 2805-2809.
[31] A. Zappavigna, P. Colaneri, J. Geromel, R. Middleton, Dwell time analysis for continuous-time switched linear positive systems, in: Proceedings of American Control Conference, Marriott Waterfront, Baltimore, MD, USA, 2010, pp. 6256-6261.
[32] A. Zappavigna, P. Colaneri, J. Geromel, R. Middleton, Stabilization of continuous-time switched linear positive systems, in: Proceedings of American Control Conference, Marriott Waterfront, Baltimore, MD, USA, 2010, pp. 3275-3280.
[33] L. Zhang, N. Cui, M. Liu, Y. Zhao, Asynchronous filtering of discrete-time switched linear systems with average dwell time, IEEE Trans. Circ Syst.: Part I 58 (5) (2011) 1109-1118.
[34] L. Zhang, H. Gao, Asynchronously switched control of switched linear systems with average dwell time, Automatica 46 (5) (2010) 953-958.
[35] L. Zhang, P. Shi, Stability, $l_{2}$-gain and asynchronous $H_{\infty}$ control of discrete-time switched systems with average dwell time, IEEE Trans. Autom. Control 54 (9) (2009) 2193-2200.
[36] J. Zhang, Z. Han, F. Zhu, J. Huang, Stability and stabilization of positive switched systems with mode-dependent average dwell time, Nonlinear Anal.: Hybrid Syst. 9 (1) (2013) 42-45.
[37] X. Zhao, L. Zhang, P. Shi, M. Liu, Stability and stabilization of switched linear systems with mode-dependent average dwell time, IEEE Trans. Autom. Control 57 (7) (2012) 1809-1815.
[38] X. Zhao, L. Zhang, P. Shi, M. Liu, Stability of switched positive linear systems with average dwell time switching, Automatica 48 (6) (2012) $1132-1137$.
[39] X. Zhao, L. Zhang, P. Shi, Stability of a class of switched positive linear time-delay systems, Int. J. Robust Nonlinear Control 23 (5) (2013) $578-589$.

