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Exponential stability analysis of switched positive nonlinear systems with impulsive effects via multiple max-separable Lyapunov function

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Appendix A Preliminaries

Let $\mathbb{R}_+^n := \{\mathbf{x} \in \mathbb{R}^n, x_j \geq 0, 1 \leq j \leq n\}$, where \mathbb{R}^n stands for n -dimensional Euclidean space. $\mathbb{N}_0 = \{0, 1, 2, \dots\}$. I denotes the identity matrix. For $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$, define: $\mathbf{x} \gg \mathbf{y}$, if $x_j > y_j, 1 \leq j \leq n$; $\mathbf{x} \geq \mathbf{y}$, if $x_j \geq y_j$, for $1 \leq j \leq n$; $\mathbf{x} > \mathbf{y}$, if $\mathbf{x} \geq \mathbf{y}$ and $\mathbf{x} \neq \mathbf{y}$. $A = (a_{ij})_{n \times n}$ is Metzler matrix if $a_{ij} \geq 0 (i \neq j)$. The j th coordinate of vector \mathbf{x}_p is denoted by x_{pj} . The weighted l_∞ norm of $\mathbf{x} \in \mathbb{R}_+^n$ is given by

$$\|\mathbf{x}\|_\infty^v = \max_{1 \leq j \leq n} \frac{|x_j|}{v_j}.$$

For an n -tuple $\mathbf{r} = \{r_1, r_2, \dots, r_n\}$ and a constant $\lambda > 0$, the dilation map is defined by $\delta_\lambda^{\mathbf{r}}$

$$\delta_\lambda^{\mathbf{r}}(\mathbf{x}) = \begin{bmatrix} \lambda^{r_1} & 0 & 0 & \dots & 0 & 0 \\ 0 & \lambda^{r_2} & 0 & \dots & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 & \lambda^{r_n} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} \lambda^{r_1} x_1 \\ \lambda^{r_2} x_2 \\ \vdots \\ \lambda^{r_n} x_n \end{bmatrix},$$

where $\lambda_j > 0 (j = 1, 2, \dots, n)$. The dilation map $\delta_\lambda^{\mathbf{r}}$ is standard when $\mathbf{r} = \{1, 1, \dots, 1\}$.

Definition 1. A vector field $\mathbf{f} : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is cooperative if the Jacobian matrix $\frac{\partial \mathbf{f}}{\partial \mathbf{x}}(\mathbf{a})$ is Metzler matrix for all $\mathbf{a} \in \mathbb{R}_+^n \setminus \{\mathbf{0}\}$, where $\mathbf{f} : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is continuous and continuously differentiable on $\mathbb{R}_+^n \setminus \{\mathbf{0}\}$.

From ([1], Chapter 3, Remark 1.1, p.33), the cooperative vectors satisfy Kamke-condition.

Proposition 1. If $\mathbf{f} : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is a cooperative field, then given any two vectors \mathbf{x} and \mathbf{y} in \mathbb{R}_+^n satisfying $\mathbf{x} \geq \mathbf{y}$ and $x_j = y_j$, one has $f_j(\mathbf{x}) \geq f_j(\mathbf{y})$.

Definition 2. A vector field $\mathbf{g} : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is called order-preserving on \mathbb{R}_+^n , if $\mathbf{g}(\mathbf{x}) \geq \mathbf{g}(\mathbf{y})$ for any $\mathbf{x}, \mathbf{y} \in \mathbb{R}_+^n$ satisfying $\mathbf{x} \geq \mathbf{y}$.

If for any $t \in [0, +\infty)$, $\mathbf{f}_{\sigma(t)}$ is a cooperative field and $\mathbf{g}_{\sigma(t)}$ is order-preserving field, then system (1) is SPNSs. More precisely, in the first interval $[0, t_1)$, the solution of the first activated subsystem is nonnegative for any nonnegative initial condition. This implies that the initial condition is nonnegative for the second interval $[t_1, t_2)$. Hence, the solution of the system (1) is always nonnegative.

Definition 3. System (1) is exponentially stable (ES), if for nonnegative initial function $\boldsymbol{\theta}(t)$ and any switching signal from a fixed ADT class, there exist two constants $a_0 > 0$ and $b_0 > 0$ such that the solutions of system (1) satisfy

$$\|\mathbf{x}(t)\| \leq a_0 e^{-b_0 t} \|\boldsymbol{\theta}(t)\|, \quad t \geq 0,$$

where $\|\cdot\|$ is some norm in \mathbb{R}^n .

Definition 4. Denote $N_\sigma(T_1, T_2)$ be the number of switching times of $\sigma(t)$ over the interval $[T_1, T_2]$. If there exist two positive constants τ_a and N_0 such that

$$N_\sigma(T_1, T_2) \leq N_0 + \frac{T_2 - T_1}{\tau_a},$$

then τ_a is called an average dwell time (ADT) of $\sigma(t)$.

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Appendix B The proof of Lemma 1

Proof: For any $j \in \{1, 2, \dots, n\}$, one can get

$$\begin{aligned} \left(\frac{x_j(t)}{v_{pj}}\right)^{\frac{r_{max}}{r_j}} &= \left(\frac{x_j(t)}{v_{qj}} \frac{v_{qj}}{v_{pj}}\right)^{\frac{r_{max}}{r_j}} \\ &= \left(\frac{x_j(t)}{v_{qj}}\right)^{\frac{r_{max}}{r_j}} \left(\frac{v_{qj}}{v_{pj}}\right)^{\frac{r_{max}}{r_j}} \\ &\leq \alpha^l \left(\frac{x_j(t)}{v_{qj}}\right)^{\frac{r_{max}}{r_j}} \\ &\leq \alpha^l \left(\frac{x_j(t)}{v_{qj}}\right)^{\frac{r_{max}}{r_j}} \\ &= \beta \left(\frac{x_j(t)}{v_{qj}}\right)^{\frac{r_{max}}{r_j}}, \end{aligned}$$

then it follows from the definition of the multiple max-separate Lyapunov function that (2) holds.

Appendix C The proof of Theorem 1

Proof: For any $p \in M, j \in \{1, 2, \dots, n\}$, define function $z_{pj}(\lambda) = \lambda + \frac{f_{pj}(\mathbf{v}_p)}{v_{pj}} + e^{\lambda \tau_{max}} \frac{g_{pj}(\mathbf{v}_p)}{v_{pj}}$, then it is easily to check that $z_{pj}(0) < 0$ and $z_{pj}(\lambda_{pj}) = 0$. Note that $z_{pj}(\lambda)$ is monotonically increasing in λ , then picking any λ_0 with $0 < \lambda_0 < \min_{p \in M} \min_{1 \leq j \leq n} \lambda_{pj}$, the following inequality holds

$$z_{pj}(\lambda_0) < 0, \quad j \in \{1, 2, \dots, n\}. \quad (C1)$$

Set $\mathbf{x}(t) = e^{-\lambda_0 t} \mathbf{y}^*(t)$, then it follows from system (1) that

$$\dot{\mathbf{y}}^*(t) \leq \lambda_0 \mathbf{y}^*(t) + \mathbf{f}_{\sigma(t)}(\mathbf{y}^*(t)) + e^{\lambda_0 \tau(t)} \mathbf{g}_{\sigma(t)}(\mathbf{y}^*(t - \tau(t))), \quad t \geq 0. \quad (C2)$$

Next, consider the following impulsive switched positive system

$$\begin{cases} \dot{\mathbf{y}}(t) = \lambda_0 \mathbf{y}(t) + \mathbf{f}_{\sigma(t)}(\mathbf{y}(t)) + e^{\lambda_0 \tau(t)} \mathbf{g}_{\sigma(t)}(\mathbf{y}(t - \tau(t))), & t \geq 0, t \neq t_k, \\ \Delta \mathbf{y}(t_k) = C_{\sigma(t_k)} \mathbf{y}((t_k - d(t_k))^-), \\ \mathbf{y}(t) = e^{\lambda_0 t} \boldsymbol{\theta}(t), \end{cases} \quad t \in [-\tau_{max}, 0]. \quad (C3)$$

The switching sequence is denoted by $0 = t_0 < t_1 < t_2 < \dots < t_k < t_{k+1} < \dots$. Then for the first interval $[t_0, t_1)$, let $\sigma(t) = \sigma(t_0)$, $t \in [t_0, t_1)$. Define $\|\boldsymbol{\theta}\| = \sup_{-\tau_{max} \leq t \leq 0} V_{\underline{\mathbf{v}}}(\boldsymbol{\theta}(t))$, where $\underline{\mathbf{v}}$ is defined in Lemma 1. Hence for any $\rho > 1$ one can check

$$V_{\mathbf{v}_{\sigma(t_0)}}(\mathbf{y}(0)) = V_{\mathbf{v}_{\sigma(t_0)}}(\boldsymbol{\theta}(0)) \leq \|\boldsymbol{\theta}\| < \rho \|\boldsymbol{\theta}\|.$$

Next, the following inequality will be proved

$$V_{\mathbf{v}_{\sigma(t_0)}}(\mathbf{y}(t)) < \rho \|\boldsymbol{\theta}\|, \quad \forall t \in [t_0, t_1). \quad (C4)$$

By contradiction, assume that (C4) is not always correct. Note that $V_{\mathbf{v}_{\sigma(t_0)}}(\mathbf{y}(t))$ is continuous, then there exist a time t^* and at least one index $j_0 \in \{1, 2, \dots, n\}$ such that

$$\left(\frac{y_j(t)}{v_{\sigma(t_0)j}}\right)^{\frac{r_{max}}{r_j}} < \rho \|\boldsymbol{\theta}\|, \quad t \leq t^*, \quad j \in \{1, 2, \dots, n\}, \quad (C5)$$

$$\left(\frac{y_{j_0}(t^*)}{v_{\sigma(t_0)j_0}}\right)^{\frac{r_{max}}{r_{j_0}}} = \rho \|\boldsymbol{\theta}\|, \quad (C6)$$

$$D_- \left(\frac{y_{j_0}(t)}{v_{\sigma(t_0)j_0}}\right)^{\frac{r_{max}}{r_{j_0}}} \Big|_{t=t^*} \geq 0, \quad (C7)$$

where D_- denotes the left derivative.

Case 1: If $t^* - \tau(t^*) \geq 0$, then $t^* - \tau(t^*) \in [0, t^*]$. Combining (C5) with (C6) can lead to

$$y_j(t^* - \tau(t^*)) \leq (\rho \|\boldsymbol{\theta}\|)^{\frac{r_j}{r_{max}}} v_{\sigma(t_0)j}, \quad j \in \{1, 2, \dots, n\}.$$

Since $\rho > 1$, $(\rho \|\boldsymbol{\theta}\|)^{\frac{1}{r_{max}}} > (\|\boldsymbol{\theta}\|)^{\frac{1}{r_{max}}}$. Therefore, if $\|\boldsymbol{\theta}\| \frac{1}{r_{max}} > \lambda^*$, then $(\rho \|\boldsymbol{\theta}\|)^{\frac{1}{r_{max}}} > \lambda^*$. Furthermore, it follows from Assumption 2 and Assumption 3 that

$$\begin{aligned} \mathbf{g}_{\sigma(t_0)}(\mathbf{y}(t^* - \tau(t^*))) &\leq \mathbf{g}_{\sigma(t_0)} \left((\rho \|\boldsymbol{\theta}\|)^{\frac{r_1}{r_{max}}} v_{\sigma(t_0)1}, (\rho \|\boldsymbol{\theta}\|)^{\frac{r_2}{r_{max}}} v_{\sigma(t_0)2}, \dots, (\rho \|\boldsymbol{\theta}\|)^{\frac{r_n}{r_{max}}} v_{\sigma(t_0)n} \right) \\ &= \mathbf{g}_{\sigma(t_0)} \left(\delta_{\lambda_{\theta}}^r \mathbf{v}_{\sigma(t_0)} \right) \\ &\leq \delta_{\lambda_{\theta}}^r \mathbf{g}_{\sigma(t_0)}(\mathbf{v}_{\sigma(t_0)}), \end{aligned}$$

where $\lambda_{\theta} = (\rho \|\boldsymbol{\theta}\|)^{\frac{1}{r_{max}}}$. Noting that $f_{\sigma(t_0)}$ is cooperative, this in turn implies

$$\begin{aligned} f_{\sigma(t_0)j_0}(\mathbf{y}(t^*)) &\leq f_{\sigma(t_0)j_0} \left((\rho \|\boldsymbol{\theta}\|)^{\frac{r_1}{r_{max}}} v_{\sigma(t_0)1}, (\rho \|\boldsymbol{\theta}\|)^{\frac{r_2}{r_{max}}} v_{\sigma(t_0)2}, \dots, (\rho \|\boldsymbol{\theta}\|)^{\frac{r_n}{r_{max}}} v_{\sigma(t_0)n} \right) \\ &= f_{\sigma(t_0)j_0} \left(\delta_{\lambda_{\boldsymbol{\theta}}}^* \mathbf{v}_{\sigma(t_0)} \right) \\ &\leq \lambda_{\boldsymbol{\theta}}^{r_{j_0}} f_{\sigma(t_0)j_0}(\mathbf{v}_{\sigma(t_0)}). \end{aligned}$$

Then

$$\begin{aligned} D_- \left(\frac{y_{j_0}(t)}{v_{\sigma(t_0)j_0}} \right)^{\frac{r_{max}}{r_{j_0}}} \Big|_{t=t^*} &= \frac{r_{max}}{r_{j_0}} \left(\frac{y_{j_0}(t^*)}{v_{\sigma(t_0)j_0}} \right)^{\frac{r_{max}}{r_{j_0}} - 1} \frac{y_{j_0}(t^*)}{v_{\sigma(t_0)j_0}} \\ &= \frac{r_{max}}{r_{j_0}} \left(\frac{y_{j_0}(t^*)}{v_{\sigma(t_0)j_0}} \right)^{\frac{r_{max}}{r_{j_0}} - 1} \frac{1}{v_{\sigma(t_0)j_0}} \left(\lambda_0 y_{j_0}(t^*) + f_{\sigma(t_0)j_0}(\mathbf{y}(t^*)) + e^{\lambda_0 \tau(t^*)} g_{\sigma(t_0)j_0}(\mathbf{y}(t^* - \tau(t^*))) \right) \\ &\leq \frac{r_{max}}{r_{j_0}} \lambda_{\boldsymbol{\theta}}^{r_{max} - r_{j_0}} \frac{1}{v_{\sigma(t_0)j_0}} \left(\lambda_0 \lambda_{\boldsymbol{\theta}}^{r_{j_0}} v_{\sigma(t_0)j_0} + \lambda_{\boldsymbol{\theta}}^{r_{j_0}} f_{\sigma(t_0)j_0}(\mathbf{v}_{\sigma(t_0)}) + e^{\lambda_0 \tau_{max}} \lambda_{\boldsymbol{\theta}}^{r_{j_0}} g_{\sigma(t_0)j_0}(\mathbf{v}_{\sigma(t_0)}) \right) \\ &= \frac{r_{max}}{r_{j_0}} \lambda_{\boldsymbol{\theta}}^{r_{max} - r_{j_0}} \lambda_{\boldsymbol{\theta}}^{r_{j_0}} \left(\lambda_0 + \frac{f_{\sigma(t_0)j_0}(\mathbf{v}_{\sigma(t_0)})}{v_{\sigma(t_0)j_0}} + e^{\lambda_0 \tau_{max}} \frac{g_{\sigma(t_0)j_0}(\mathbf{v}_{\sigma(t_0)})}{v_{\sigma(t_0)j_0}} \right) \\ &= \frac{r_{max}}{r_{j_0}} \lambda_{\boldsymbol{\theta}}^{r_{max}} \left(\lambda_0 + \frac{f_{\sigma(t_0)j_0}(\mathbf{v}_{\sigma(t_0)})}{v_{\sigma(t_0)j_0}} + e^{\lambda_0 \tau_{max}} \frac{g_{\sigma(t_0)j_0}(\mathbf{v}_{\sigma(t_0)})}{v_{\sigma(t_0)j_0}} \right) \\ &< 0, \end{aligned}$$

where the last inequality is from (C1). This contradicts (C7).

Case 2: If $t^* - \tau(t^*) < 0$, then

$$\mathbf{y}(t^* - \tau(t^*)) = e^{\lambda_0(t^* - \tau(t^*))} \boldsymbol{\theta}(t^* - \tau(t^*)) \leq \boldsymbol{\theta}(t^* - \tau(t^*)),$$

which means

$$V_{\mathbf{v}_{\sigma(t_0)}}(\mathbf{y}(t^* - \tau(t^*))) < \rho \|\boldsymbol{\theta}\|.$$

Similar to the analysis in Case 1, $D_- \left(\frac{y_{j_0}(t)}{v_{\sigma(t_0)j_0}} \right)^{\frac{r_{max}}{r_{j_0}}} \Big|_{t=t^*} < 0$ can be obtained, which also yields a contradiction. Thus, (C4) is true for any $\rho > 1$. Let ρ tend to 1^+ , then

$$V_{\mathbf{v}_{\sigma(t_0)}}(\mathbf{y}(t)) \leq \|\boldsymbol{\theta}\|, \quad \forall t \in [t_0, t_1].$$

This combines with Lemma 1 leads to

$$V_{\mathbf{v}_{\sigma(t_1)}}(\mathbf{y}(t)) \leq \beta \|\boldsymbol{\theta}\|, \quad \forall t \in [t_0, t_1].$$

That is, for any $t \in [t_0, t_1)$, $\left(\frac{y_j(t)}{v_{\sigma(t_1)j}} \right)^{\frac{r_{max}}{r_j}} \leq \beta \|\boldsymbol{\theta}\|$. Let t tend to t_1^- , then

$$y_j(t_1^-) \leq (\beta \|\boldsymbol{\theta}\|)^{\frac{r_j}{r_{max}}} v_{\sigma(t_1)j}. \tag{C8}$$

Since $-\tau_{max} \leq t_1 - d(t_1) \leq t_1$, the following inequality is always true

$$y_j \left((t_1 - d(t_1))^- \right) \leq (\beta \|\boldsymbol{\theta}\|)^{\frac{r_j}{r_{max}}} v_{\sigma(t_1)j}. \tag{C9}$$

Combining (C8) and (C9) gives

$$\begin{aligned} y_j(t_1) &= y_j(t_1^-) + C_{\sigma(t_1)jj} y_j \left((t_1 - d(t_1))^- \right) \\ &\leq (\beta \|\boldsymbol{\theta}\|)^{\frac{r_j}{r_{max}}} (1 + C_{\sigma(t_1)jj}) v_{\sigma(t_1)j} \\ &\leq (\beta \|\boldsymbol{\theta}\|)^{\frac{r_j}{r_{max}}} h v_{\sigma(t_1)j}. \end{aligned}$$

Then it follows that

$$V_{\mathbf{v}_{\sigma(t_1)}}(\mathbf{y}(t_1)) \leq \beta \eta \|\boldsymbol{\theta}\|.$$

Applying the similar analysis to the one of the first interval, one can verify that

$$V_{\mathbf{v}_{\sigma(t_1)}}(\mathbf{y}(t)) \leq \beta \eta \|\boldsymbol{\theta}\|, \quad \forall t \in [t_1, t_2)$$

and

$$V_{\mathbf{v}_{\sigma(t_2)}}(\mathbf{y}(t_2)) \leq (\beta \eta)^2 \|\boldsymbol{\theta}\|.$$

By induction, for each $k \in \{1, 2, \dots, n, \dots\}$, the following inequality can be obtained

$$V_{\mathbf{v}_{\sigma(t_k)}}(\mathbf{y}(t)) \leq (\beta \eta)^k \|\boldsymbol{\theta}\|, \quad t \in [t_k, t_{k+1}). \tag{C10}$$

It should be pointed out that even if $\lim_{t \rightarrow +\infty} t_k - d(t_k) = +\infty$ holds for the impulsive delay, (C10) is always true. The reasons are that $t_k - d(t_k) \leq t_k$ and $\beta \eta \geq 1$. Furthermore, for any $t \in [0, +\infty)$, assume that $t \in [t_k, t_{k+1})$, then we have

$$V_{\mathbf{v}_{\sigma(t_k)}}(\mathbf{y}(t)) \leq (\beta h)^k \|\boldsymbol{\theta}\| \leq (\beta \eta)^{N_0 + \frac{t}{\tau_a}} \|\boldsymbol{\theta}\|.$$

It follows from (C2) and (C3) that

$$\begin{aligned}
V_{\mathbf{v}_{\sigma(t_k)}}(\mathbf{x}(t)) &= e^{-\lambda_0 t} V_{\mathbf{v}_{\sigma(t_k)}}(\mathbf{y}^*(t)) \\
&\leq e^{-\lambda_0 t} V_{\mathbf{v}_{\sigma(t_k)}}(\mathbf{y}(t)) \\
&\leq (\beta\eta)^{N_0} e^{-\lambda_0 t} (\beta\eta)^{\frac{t}{\tau_a}} \|\boldsymbol{\theta}\| \\
&= (\beta\eta)^{N_0} e^{-(\lambda_0 - \frac{\ln(\beta\eta)}{\tau_a})t} \|\boldsymbol{\theta}\|.
\end{aligned}$$

Furthermore, from Lemma 1, we can get

$$V_{\overline{\mathbf{v}}}(\mathbf{x}(t)) \leq V_{\mathbf{v}_{\sigma(t_k)}}(\mathbf{x}(t)) \leq (\beta h)^{N_0} e^{-(\lambda_0 - \frac{\ln(\beta\eta)}{\tau_a})t} \|\boldsymbol{\theta}\|.$$

Since the switching signal satisfies $\tau_a > \frac{\ln(\beta\eta)}{\lambda_0}$, that is $\lambda_0 - \frac{\ln(\beta\eta)}{\tau_a} > 0$, this implies that system (1) is ES.

Appendix D The proof of Theorem 2

Proof: From (5), we have

$$\lambda_0 + \left(a_{jj}^p + \sum_{i=1, i \neq j}^{i=n} \frac{1}{v_{pj}} \tilde{a}_{ji}^p v_{pi} \right) + e^{\lambda_0 \tau_{max}} \sum_{i=1}^{i=n} \frac{1}{v_{pj}} |b_{ji}^p| v_{pi} < 0, \quad \forall p \in M, \quad j \in \{1, 2, \dots, n\}. \quad (D1)$$

For the linear system (4), since $r_1 = r_2 = \dots = r_n$, the multiple max-separable Lyapunov function can be written as

$$V_{\mathbf{v}_p}(\mathbf{x}) = \max_{1 \leq j \leq n} \left(\frac{|x_j|}{v_{pj}} \right)^{\frac{r_{max}}{r_j}} = \max_{1 \leq j \leq n} \left(\frac{|x_j|}{v_{pj}} \right).$$

Similar to the proof of (C4), suppose

$$\frac{|x_j(t)|}{v_{\sigma(t_0)j}} < \rho \|\boldsymbol{\theta}\|, \quad t \leq t^*, \quad j \in \{1, 2, \dots, n\}, \quad (D2)$$

$$\frac{|x_{j_0}(t^*)|}{v_{\sigma(t_0)j_0}} = \rho \|\boldsymbol{\theta}\|, \quad (D3)$$

$$D_- \frac{|x_{j_0}(t)|}{v_{\sigma(t_0)j_0}} \Big|_{t=t^*} \geq 0, \quad (D4)$$

Then the left derivative of $|x_{j_0}(t^*)|/v_{pj_0}$ needs to be calculated, that is

$$\begin{aligned}
D_- \frac{|x_{j_0}(t)|}{v_{pj_0}} \Big|_{t=t^*} &= \frac{\text{sign}(x_{j_0}(t^*)) \dot{x}_{j_0}(t^*)}{v_{pj_0}} \\
&= \frac{\text{sign}(x_{j_0}(t^*))}{v_{pj_0}} \left(\lambda_0 x_{j_0}(t^*) + \sum_{i=1}^{i=n} a_{j_0 i}^p x_i(t^*) + e^{\lambda_0 \tau(t^*)} \sum_{i=1}^{i=n} b_{j_0 i}^p x_i(t^*) \right) \\
&\leq \frac{1}{v_{pj_0}} \left(\lambda_0 |x_{j_0}(t^*)| + a_{j_0 j_0}^p |x_{j_0}(t^*)| + \sum_{i=1, i \neq j_0}^{i=n} \tilde{a}_{j_0 i}^p |x_i(t^*)| + e^{\lambda_0 \tau_{max}} \sum_{i=1}^{i=n} |b_{j_0 i}^p| |x_i(t^*)| \right).
\end{aligned}$$

It follows from (D1) that $D_- \frac{|x_{j_0}(t)|}{v_{pj_0}} \Big|_{t=t^*} < 0$, which contradicts (D4). Then the rest proof which shows similarities with the one of Theorem 1 is left out here.

References

- 1 Smith H. Monotone dynamical systems: an introduction to the theory of competitive and cooperative systems. American Mathematical Society: Providence, RI, 1995.