

## Letter

## Synchronization of Drive-Response Networks With Delays on Time Scales

Yanxia Tan and Zhenkun Huang

Dear Editor,

This letter deals with the synchronization in drive-response networks with discrete and distributed delays on time scales. Based on the theory of calculus on time scales, Lyapunov functional method and inequality technique, we obtain new sufficient conditions to ensure synchronization criteria which are dependent on boundedness of graininess but independent of the types of delays. Numerical simulations are given to illustrate the effectiveness of our new results.

Recently, the synchronization problem has played a significant role in nonlinear science since its potential applications in image processing, ecological systems, secure communication and harmonic oscillation generation, formation control, etc. [1]–[5]. Ways of Synchronization for chaos systems which is applied to communication is investigated in [1]. In [2], the authors discuss the synchronization via pinning control on general complex networks. In [3], researchers have studied the sufficient criteria which can ensure the occurrence of synchronization for Lur'e systems. Outer synchronization between drive and response networks via adaptive impulsive pinning control is investigated in [6]. An effective control input and a feedback control with an updated law are designed to realize finite-time synchronization between two complex networks [7].

Among the various works of synchronization, continuous-time synchronization [4], [5] and discrete-time synchronization [8] cases are studied respectively. However, in most systems, the interaction among neurons would occur at any time, maybe discrete moments following with continuous-time intervals. Due to the trouble of studying synchronization for continuous-time and discrete-time respectively, it is necessary and meaningful to research them at the same time in neural networks on time scales. The theory of time scales [9] has received a lot of attention recently, which was introduced by Hilger in order to unify continuous and discrete analysis. The field of dynamic systems on time scales links and extends the classical theory of differential and difference equations. Various works have showed that the theory of time scales is an essential tool to deal with many practical problems. Many works concerning the nonlinear systems of differential equation on time scales have been reported in [9], [10] and references therein. However, there are few results dealing with the synchronization of drive-response systems with connection time delays for both discrete and distributed cases on time scales.

As encountered unavoidably in some systems, time delays should be taken into account in the modeling for practical networks. Discrete connection time delays in systems of delayed feedback provide a good approximation in some electronic networks which consist of a small number of neurons. However, due to the presence of parallel pathways with a variety of axon sizes and lengths, a spatial extent exists in biological neural networks which cannot be modeled with discrete time delays. In this case, distributed connection delays are

necessary to be used in the models [11]. Motivated by the above exposition and existing ones [12], we investigate the exponential synchronization in drive-response networks with discrete and distributed delays on time scales.

**Notations:** Throughout the paper,  $\mathbb{T}$  is a nonempty closed subset of  $\mathbb{R}$ ;  $\sigma(t) := \inf\{s \in \mathbb{T} : s > t\}$  for all  $t \in \mathbb{T}$ ;  $\mathbb{h}_i := \{i-1, i+1\}$ ,  $i \in \mathcal{N} := \{1, 2, \dots, n\}$ . If  $\mathbb{T}$  has a left-scattered maximum  $m$ , then  $\mathbb{T}^k = \mathbb{T} - \{m\}$ ; otherwise,  $\mathbb{T}^k = \mathbb{T}$ . The graininess  $\mu(t) := \sigma(t) - t$  and let  $\bar{\mu} := \sup_{t \in \mathbb{T}} \mu(t)$ .  $f^\Delta(t)$  is the delta derivative of  $f(t)$ ;  $\mathcal{R} = \{p : \mathbb{T} \rightarrow \mathbb{R} | 1 + \mu(t)p(t) \neq 0, t \in \mathbb{T}^k\}$ ;  $e_p(t, s) = \exp(\int_s^t \xi_{\mu(\tau)}(p(\tau))\Delta\tau)$  is the exponential function and  $\xi_h : \mathbb{C}_h \rightarrow \mathbb{Z}_h$  is the cylinder transformation [9], [10].

**Model description:** Consider  $n$ -neuron drive-response system with discrete delays on time scale  $\mathbb{T}$

$$x_i^\Delta(t) = -a_i x_i(t) + b_i f(x_i(t)) + \sum_{j \in \mathbb{h}_i} c_{ij} f(x_j(t - \tau_{ij})) \quad (1)$$

$$y_i^\Delta(t) = -a_i y_i(t) + b_i f(y_i(t)) + k_i(x_i(t) - y_i(t)) + \sum_{j \in \mathbb{h}_i} c_{ij} f(y_j(t - \tau_{ij})) \quad (2)$$

and corresponding drive-response system with distributed delays

$$x_i^\Delta(t) = -a_i x_i(t) + b_i f(x_i(t)) + \sum_{j \in \mathbb{h}_i} c_{ij} \int_0^{\tau_{ij}} K_{ij}(u) f(x_j(t-u)) \Delta u \quad (3)$$

$$y_i^\Delta(t) = -a_i y_i(t) + b_i f(y_i(t)) + k_i(x_i(t) - y_i(t)) + \sum_{j \in \mathbb{h}_i} c_{ij} \int_0^{\tau_{ij}} K_{ij}(u) f(y_j(t-u)) \Delta u \quad (4)$$

where  $t \in \mathbb{T} \subset \mathbb{R}$ ,  $\int_0^{\tau_{ij}} K_{ij}(u) \Delta u = 1$  and  $\bar{K}_{ij} = \sup_{t \in [0, \tau_{ij}]_{\mathbb{T}}} K_{ij}(t)$  if there exist  $\varepsilon > 0$  and  $M > 0$  such that

$$\sum_{i=1}^n w_i^2(t) \leq M \left( \sum_{i=1}^n \sup_{s \in [-\tau, 0]_{\mathbb{T}}} w_i^2(s) \right) e_{\ominus \varepsilon}(t, 0)$$

for all  $t \in [0, +\infty)_{\mathbb{T}}$ , then system (1) and (2) (or system (3) and (4)) is exponentially synchronized, where  $w_i(t) = x_i(t) - y_i(t)$  for  $i \in \mathcal{N}$ .

**Main results:** Now, we will give the criteria which are independent of delays for synchronization of systems (1)–(2) and (3)–(4).

**Theorem 1:** For drive-response system (1) and (2), if there exist positive constants  $\theta_1, \theta_2, \dots, \theta_n$  and  $\varepsilon > 0$  such that

$$(1 + \bar{\mu}\varepsilon)\bar{\mu}(|b_i|L + a_i + k_i)^2 + (|b_i|L - a_i - k_i) + \lambda_i < 0 \quad (5)$$

where

$$\lambda_i = \frac{\varepsilon}{2} + \frac{(1 + \bar{\mu}\varepsilon)L}{2} \sum_{j \in \mathbb{h}_i} \left[ |c_{ij}| + \frac{\theta_j}{\theta_i} |c_{ji}| e^{\varepsilon\tau_{ji}} (1 + 4\bar{\mu}L|c_{ji}|) \right]$$

and  $i \in \mathcal{N}$ , then system (1) and (2) is exponentially synchronized.

**Proof:** Let  $w_i(t) = x_i(t) - y_i(t)$  for  $i \in \mathcal{N}$  and  $t \in [-\tau, +\infty)_{\mathbb{T}}$ . It follows from (1) and (2) that:

$$w_i^\Delta(t) = -a_i w_i(t) + b_i[f(x_i(t)) - f(y_i(t))] - k_i w_i(t) + \sum_{j \in \mathbb{h}_i} c_{ij}[f(x_j(t - \tau_{ij})) - f(y_j(t - \tau_{ij}))] \quad (6)$$

which leads to

$$|w_i^\Delta(t)| \leq (|b_i|L + a_i + k_i)|w_i(t)| + L \sum_{j \in \mathbb{h}_i} |c_{ij}| w_j(t - \tau_{ij}).$$

Hence, we can get that

$$(w_i^\Delta(t))^2 \leq 2(|b_i|L + a_i + k_i)^2 (w_i(t))^2 + 4L^2 \sum_{j \in \mathbb{h}_i} c_{ij}^2 w_j^2(t - \tau_{ij}) \quad (7)$$

Corresponding author: Zhenkun Huang.

Citation: Y. Tan and Z. Huang, "Synchronization of drive-response networks with delays on time scales," *IEEE/CAA J. Autom. Sinica*, vol. 11, no. 4, pp. 1063–1065, Apr. 2024.

The authors are with the School of Science, Jimei University, Xiamen 361021, China (e-mail: cxm2023doc@qq.com; hzk974226@jmu.edu.cn).

Color versions of one or more of the figures in this paper are available online at <http://ieeexplore.ieee.org>.

Digital Object Identifier 10.1109/JAS.2016.7510043

and

$$w_i^\Delta(t)w_i(t) \leq (|b_i|L - a_i - k_i)w_i^2(t) + L \sum_{j \in \mathbb{B}_i} |c_{ij}| |w_i(t)w_j(t - \tau_{ij})|. \quad (8)$$

Choose a Lyapunov functional candidate as

$$V(t) = \frac{1}{2}e_\varepsilon(t, 0) \sum_{i=1}^n \theta_i w_i^2(t) + I_1(t) + I_2(t), \quad t \in [0, +\infty)_{\mathbb{T}}$$

where

$$I_1(t) = \frac{1 + \bar{\mu}\varepsilon}{2} L \sum_{i=1}^n \theta_i \sum_{j \in \mathbb{B}_i} c_{ij} \int_{t-\tau_{ij}}^t e_\varepsilon(s + \tau_{ij}, 0) w_j^2(s) \Delta s$$

$$I_2(t) = 2(1 + \bar{\mu}\varepsilon) \bar{\mu} L^2 \sum_{i=1}^n \theta_i \sum_{j \in \mathbb{B}_i} c_{ij}^2 \int_{t-\tau_{ij}}^t e_\varepsilon(s + \tau_{ij}, 0) w_j^2(s) \Delta s.$$

It follows from (6) that:

$$\begin{aligned} V^\Delta(t) &\leq (1 + \mu(t)\varepsilon) e_\varepsilon(t, 0) \sum_{i=1}^n \theta_i w_i^2(t) \\ &\quad \times \left[ \frac{\varepsilon}{2} + (|b_i|L - a_i - k_i) + \frac{(1 + \bar{\mu}\varepsilon)L}{2} \sum_{j \in \mathbb{B}_i} |c_{ij}| \right. \\ &\quad + \frac{(1 + \bar{\mu}\varepsilon)L}{2} \sum_{j \in \mathbb{B}_i} \frac{\theta_j}{\theta_i} |c_{ji}| e^{\varepsilon\tau_{ji}} \\ &\quad + (1 + \bar{\mu}\varepsilon) \bar{\mu} (|b_i|L + a_i + k_i)^2 \\ &\quad \left. + 2(1 + \bar{\mu}\varepsilon) \bar{\mu} L^2 \sum_{j \in \mathbb{B}_i} \frac{\theta_j}{\theta_i} c_{ji}^2 e^{\varepsilon\tau_{ji}} \right]. \end{aligned}$$

From (5), one gets  $V^\Delta(t) \leq 0$  for all  $t \in [0, +\infty)_{\mathbb{T}}$ . So, we have  $V(t) \leq V(0)$  and hence

$$\sum_{i=1}^n w_i^2(t) \leq M_1 \left( \sum_{i=1}^n \sup_{s \in [-\tau, 0]_{\mathbb{T}}} w_i^2(s) \right) e_{\ominus\varepsilon}(t, 0) \quad (9)$$

where

$$\begin{aligned} M_1 &= \frac{\max_{i \in \mathcal{N}} \{\theta_i\}}{\min_{i \in \mathcal{N}} \{\theta_i\}} \cdot \max_{i \in \mathcal{N}} \left\{ 1 + (1 + \bar{\mu}\varepsilon)L \sum_{j \in \mathbb{B}_i} |c_{ji}| \right. \\ &\quad \left. \times \tau_{ji} e^{\varepsilon\tau_{ji}} (1 + 4\bar{\mu}L|c_{ji}|) \right\}. \end{aligned}$$

Next, we give the criteria to ensure the occurrence for the exponential synchronization of driver-response system (3) and (4) where the connection time delay is of the distributed case.

Theorem 2: For system (3) and (4), if there exist positive constants  $\theta_1, \theta_2, \dots, \theta_n$  and  $\varepsilon > 0$  such that

$$(1 + \bar{\mu}\varepsilon) \bar{\mu} (|b_i|L + a_i + k_i)^2 + (|b_i|L - a_i - k_i) + \eta_i < 0 \quad (10)$$

where

$$\begin{aligned} \eta_i &= \frac{\varepsilon}{2} + \frac{(1 + \bar{\mu}\varepsilon)L}{2} \sum_{j \in \mathbb{B}_i} |c_{ij}| \\ &\quad + \frac{\theta_j}{\theta_i} |c_{ji}| e^{\varepsilon\tau_{ji}} (1 + 4\bar{\mu}L|c_{ji}| \tau_{ji} \bar{K}_{ji}) \Big], \quad i \in \mathcal{N} \end{aligned}$$

then, system (3) and (4) is exponentially synchronized.

Proof: Similarly to Theorem 1, we have the error system

$$\begin{aligned} w_i^\Delta(t) &= -a_i w_i(t) + b_i [f(x_i(t)) - f(y_i(t))] - k_i w_i(t) \\ &\quad + \sum_{j \in \mathbb{B}_i} c_{ij} \int_0^{\tau_{ij}} K_{ij}(u) [f(x_j(t-u)) - f(y_j(t-u))] \Delta u \end{aligned} \quad (11)$$

which leads to

$$\begin{aligned} |w_i^\Delta(t)| &\leq (|b_i|L + a_i + k_i) |w_i(t)| \\ &\quad + L \sum_{j \in \mathbb{B}_i} |c_{ij}| \int_0^{\tau_{ij}} K_{ij}(u) |w_j(t-u)| \Delta u. \end{aligned}$$

Hence, we get

$$\begin{aligned} (w_i^\Delta(t))^2 &\leq 2(|b_i|L + a_i + k_i)^2 w_i^2(t) \\ &\quad + 4L^2 \sum_{j \in \mathbb{B}_i} c_{ij}^2 \tau_{ij} \int_0^{\tau_{ij}} K_{ij}^2(u) w_j^2(t-u) \Delta u \end{aligned} \quad (12)$$

and

$$\begin{aligned} w_i(t)w_i^\Delta(t) &\leq (|b_i|L - a_i - k_i) w_i^2(t) \\ &\quad + L \sum_{j \in \mathbb{B}_i} |c_{ij}| \int_0^{\tau_{ij}} K_{ij}(u) |w_i(t)w_j(t-u)| \Delta u. \end{aligned} \quad (13)$$

Now construct another Lyapunov functional as follows:

$$V(t) = \frac{1}{2}e_\varepsilon(t, 0) \sum_{i=1}^n \theta_i w_i^2(t) + J_1(t) + J_2(t), \quad t \in [0, +\infty)_{\mathbb{T}}$$

where

$$\begin{aligned} J_1(t) &= \frac{(1 + \bar{\mu}\varepsilon)L}{2} \sum_{i=1}^n \theta_i \sum_{j \in \mathbb{B}_i} |c_{ij}| \\ &\quad \times \int_0^{\tau_{ij}} K_{ij}(u) \left( \int_{t-u}^t e_\varepsilon(s + u, 0) w_j^2(s) \Delta s \right) \Delta u \\ J_2(t) &= 2(1 + \bar{\mu}\varepsilon) \bar{\mu} L^2 \sum_{i=1}^n \theta_i \sum_{j \in \mathbb{B}_i} c_{ij}^2 \tau_{ij} \\ &\quad \times \int_0^{\tau_{ij}} K_{ij}^2(u) \left( \int_{t-u}^t e_\varepsilon(s + u, 0) w_j^2(s) \Delta s \right) \Delta u. \end{aligned}$$

The delta derivatives of  $V(t)$  can be derived as follows:

$$\begin{aligned} V^\Delta(t) &\leq (1 + \mu(t)\varepsilon) e_\varepsilon(t, 0) \sum_{i=1}^n \theta_i w_i^2(t) \\ &\quad \times \left[ \frac{\varepsilon}{2} + (|b_i|L - a_i - k_i) \right. \\ &\quad + (1 + \bar{\mu}\varepsilon) \bar{\mu} (|b_i|L + a_i + k_i)^2 \\ &\quad + \frac{(1 + \bar{\mu}\varepsilon)L}{2} \sum_{j \in \mathbb{B}_i} |c_{ij}| + \frac{(1 + \bar{\mu}\varepsilon)L}{2} \sum_{j \in \mathbb{B}_i} \frac{\theta_j}{\theta_i} |c_{ji}| e^{\varepsilon\tau_{ji}} \\ &\quad \left. + 2(1 + \bar{\mu}\varepsilon) \bar{\mu} L^2 \sum_{j \in \mathbb{B}_i} \frac{\theta_j}{\theta_i} c_{ji}^2 \tau_{ji} \bar{K}_{ji} e^{\varepsilon\tau_{ji}} \right]. \end{aligned}$$

From (10), one gets that  $V^\Delta(t) \leq 0$  for all  $t \in [0, +\infty)_{\mathbb{T}}$  which implies that  $V(t) \leq V(0)$  and hence

$$\sum_{i=1}^n w_i^2(t) \leq M_2 \left( \sum_{i=1}^n \sup_{s \in [-\tau, 0]_{\mathbb{T}}} w_i^2(s) \right) e_{\ominus\varepsilon}(t, 0) \quad (14)$$

where

$$\begin{aligned} M_2 &= \frac{\max_{i \in \mathcal{N}} \{\theta_i\}}{\min_{i \in \mathcal{N}} \{\theta_i\}} \times \max_{i \in \mathcal{N}} \left\{ 1 + (1 + \bar{\mu}\varepsilon)L \sum_{j \in \mathbb{B}_i} |c_{ji}| \right. \\ &\quad \left. \times \tau_{ji} e^{\varepsilon\tau_{ji}} (1 + 4\bar{\mu}L|c_{ji}| \tau_{ji} \bar{K}_{ji}) \right\}. \end{aligned}$$

Remark 1: As  $\tau_{ji} \bar{K}_{ji} = 1$ , (10) has the same form as (5) and it is obvious that exponential synchronization of system (1)–(2) and (3)–(4) on time scales is guaranteed by the same criteria.

Remark 2: Synchronization criteria (5) and (10) are dependent on upper bound of graininess. When upper bound of graininess  $\bar{\mu} = 0$ , our results includes previous ones [12] as its special cases. As  $\mathbb{T} = \mathbb{Z}$ , the synchronization of coupled systems or drive-response neural networks are investigated in [13]. As  $\mathbb{T} = \mathbb{R}$ , (1)–(2) and (3)–(4) reduce to continuous cases, our results lead to ones in [12].

Remark 3: Synchronization of drive-response systems or coupled neural networks in continuous-time ( $\mu(t) = 0$ ) are studied in [4], [5], [12] and discrete-time ( $\mu(t) = 1$ ) cases are investigated in [13]. This is the first time that we apply the theory of time scales to unify and improve discrete-time and continuous-time neural network under the same frameworks of systems (1)–(2) and (3)–(4). It is well known

that few works have been done to study the synchronization of drive-response systems on time scales, thus the main technique that we use in Theorems 1 and 2 is novel.

**Numerical example:** For convenience, we let  $n = 6$  and  $x(t) = (x_1(t), x_2(t), \dots, x_6(t))^T$ ,  $y(t) = (y_1(t), y_2(t), \dots, y_6(t))^T$ ,  $t \in [0, +\infty)_{\mathbb{T}}$  for the systems (1)–(2) and (3)–(4) in the following example.

Example 1: Consider  $\mathbb{T} = \mathbb{P}_{1,1} = \bigcup_{k=0}^{\infty} [2k, 2k+1]$ . In this case, we have  $\mu(t) = 1$ , and  $\bar{\mu} = 1$ . For system (1) and (2), let  $a_i = 0.1$ ,  $b_i = -0.1$ ,  $c_{ij} = 0.05$ ,  $\tau_{ij} = 1.0$ ,  $k_i = 0.5$ ,  $\theta_i = 0.5$  for  $i = 1, 2, \dots, 6$ , and  $f(x) = 0.1 \sin x$ . For  $s \in [-\tau, 0]_{\mathbb{T}=\mathbb{P}_{1,1}}$ , the initial condition is

$$(x_1(s), x_2(s), \dots, x_6(s)) = (0.05, 0.1, 0.15, 0.2, 0.25, 0.3) \\ (y_1(s), y_2(s), \dots, y_6(s)) = -(0.05, 0.1, 0.15, 0.2, 0.25, 0.3).$$

It is easy to check that (5) holds. Then, for  $t \in [0, +\infty)_{\mathbb{T}=\mathbb{P}_{1,1}}$ , the state trajectory of  $x(t)$  and  $y(t)$  with exponential synchronization can be observed in Fig. 1.

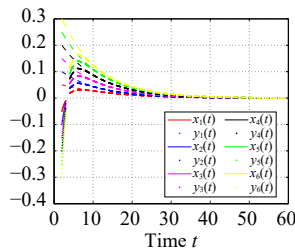


Fig. 1. The state trajectory of system (1)–(2) with  $\mathbb{T} = \mathbb{P}_{1,1}$ .

For system (3) and (4), let  $a_i = 0.2$ ,  $b_i = -0.1$ ,  $c_{ij} = 0.05$ ,  $\tau_{ij} = 1.0$ ,  $k_i = 0.5$ ,  $\theta_i = 0.5$  for  $i = 1, 2, \dots, 6$ , and  $f(x) = 0.1 \sin x$ . For  $s \in [-\tau, 0]_{\mathbb{T}=\mathbb{P}_{1,1}}$ , the initial condition is

$$(x_1(s), x_2(s), \dots, x_6(s)) = (0.1, 0.2, 0.3, 0.4, 0.5, 0.6) \\ (y_1(s), y_2(s), \dots, y_6(s)) = -(0.1, 0.2, 0.3, 0.4, 0.5, 0.6)$$

with the functions

$$K_{ij}(u) = 1, \quad i = 1, 2, \dots, 6, \quad j = i - 1, i + 1.$$

It is easy to check that (10) holds. Then, for  $t \in [0, +\infty)_{\mathbb{T}=\mathbb{P}_{1,1}}$ , the state trajectory of  $x(t)$  and  $y(t)$  with exponential synchronization can be observed in Fig. 2.

**Conclusion:** In this letter, by using the Lyapunov functional method and inequality technique, we have derived new sufficient criteria to ensure the synchronization of systems (1)–(2) and (3)–(4) on time scales. Both continuous-time and discrete-time synchronization are studied in a unified framework (1)–(2) and (3)–(4) as well as the connection time delays are of discrete or distributed cases. Our numerical results have proved that our results obtained in this letter are effective and the method utilized in this letter can be served as a universal framework for the research of other dynamic systems on time scales.

**Acknowledgments:** This work was supported by the National Natural Science Foundation of China (61573005).

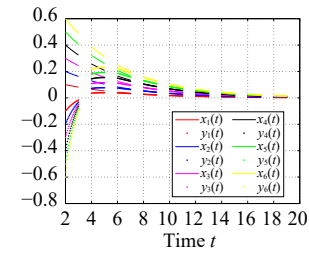


Fig. 2. The state trajectory of system (3) and (4) when  $\mathbb{T} = \mathbb{P}_{1,1}$ .

## References

- [1] L. Kocarev and U. Parlitz, "General approach for chaotic synchronization with applications to communication," *Physical Review Letters*, vol. 74, no. 25, pp. 5028–5031, 1995.
- [2] W. Yu, G. Chen, J. Lu, and J. Kurths, "Synchronization via pinning control on general complex networks," *SIAM J. Control Optimization*, vol. 51, no. 2, pp. 1395–1416, 2013.
- [3] J. Cao, H. Li, and D. Ho, "Synchronization criteria of Lur'e systems with time-delay feedback control," *Chaos Solutions Fractals*, vol. 23, no. 4, pp. 1285–1298, 2005.
- [4] J. Lu, D. Ho, J. Cao, and J. Kurths, "Exponential synchronization of linearly coupled neural networks with impulsive disturbances," *IEEE Trans. Neural Networks*, vol. 22, no. 2, pp. 329–335, 2011.
- [5] G. Chen, J. Zhou, and Z. Liu, "Global synchronization of coupled delayed neural networks and applications to chaos CNN models," *Int. J. Bifurcation Chaos*, vol. 14, no. 7, pp. 2229–2240, 2004.
- [6] Z. Wu, G. Chen, and X. Fu, "Outer synchronization of drive-response dynamical networks via adaptive impulsive pinning control," *J. the Franklin Institute*, vol. 352, no. 10, pp. 4297–4308, 2015.
- [7] J. Mei, M. Jiang, and J. Wang, "Finite-time structure identification and synchronization of drive-response systems with uncertain parameter," *Communi. in Nonlinear Science Numerical Simulation*, vol. 18, no. 4, pp. 999–1015, 2013.
- [8] M. Park, O. Kwon, J. Park, S. Lee, and E. Cha, "Synchronization of discrete-time complex dynamical networks with interval time-varying delays via non-fragile controller with randomly occurring perturbation," *J. the Franklin Institute*, vol. 351, no. 10, pp. 4850–4871, 2014.
- [9] M. Bohner, and A. Peterson, *Dynamic Equations on Time Scales: An Introduction With Applications*, Boston, USA: Birkhäuser, 2001.
- [10] Z. Huang, Y. Raffoul, and C. Cheng, "Scale-limited activating sets and multiperiodicity for threshold-linear networks on time scales," *IEEE Trans. Cyber.*, vol. 44, no. 4, pp. 488–499, 2014.
- [11] K. Gopalsamy and I. Leung, "Delay induced periodicity in a neural netlet of excitation and inhibition," *Physical D Nonlinear Phenomena*, vol. 89, no. 34, pp. 395–426, 1996.
- [12] C. Li and S. Yang, "Exponential synchronization in drive-response systems of hopfield-type neural networks with time delays," *Int. J. Bifurcation Chaos*, vol. 27, no. 11, pp. 4167–4176, 2007.
- [13] Z. Yuan, Z. Xu, and L. Guo, "Generalized synchronization of two bidirectionally coupled discrete dynamical systems," *Communi. Science Nonlinear Numerical Simulation*, vol. 17, no. 2, pp. 992–1002, 2012.