

Finite-time synchronization of drive-response systems via periodically intermittent adaptive control

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Abstract: In this paper, the finite-time synchronization between two complex dynamical networks via the periodically intermittent adaptive control and periodically intermittent feedback control is studied. The finite-time synchronization criteria are derived based on finite-time stability theory, the differential inequality and the analysis technique. Since the traditional synchronization criteria for some models are improved in the convergence time by using the novel periodically intermittent adaptive control and periodically intermittent feedback control, the results of this paper are important. Numerical examples are finally presented to illustrate the effectiveness and correctness of the theoretical results.

Keywords: Drive-response systems; finite-time synchronization; periodically intermittent adaptive control; periodically intermittent feedback control.

1 Introduction

For the last decade or so, models of complex networks have been widely used to research real systems in nature, society and engineering. A complex network consists of a large set of interconnected nodes, where a node is a fundamental unit with detailed contents. These nodes may have different meanings in different situations, such as microprocessors, computers, companies, and so on. However, the network structure facilitates and constrains the network dynamical behaviors.

In the past decades, the control and synchronization of complex dynamical networks has attracted much attention and some relevant theoretical results have been established [1-7],

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due to its many potential practical applications. Many effective control methods including adaptive control [8-14], feedback control [15-18], observer control [19-20], pinning control [21-23], impulsive control [24-28] and intermittent control [29-37] have been proposed to drive the network to achieve synchronization. Among these control approaches, the discontinuous control methods which include impulsive control and intermittent control have attracted much interest due to its practical and easy implementation in engineering fields. But, the intermittent control is different from the impulsive control since impulsive control is activated only at some isolated instants, while intermittent control has a nonzero control width, the explain can be found in the map (see Fig. 1). Obviously, when $u = v$, the intermittent control becomes the general continuous control, while $u = 0$, the intermittent control becomes the impulsive control. Besides, using intermittent control is more effective and robust [38]. Therefore, in recent years, many synchronization criteria for complex dynamical networks with or without time delays via intermittent control have been presented, see [30-31, 39-40].

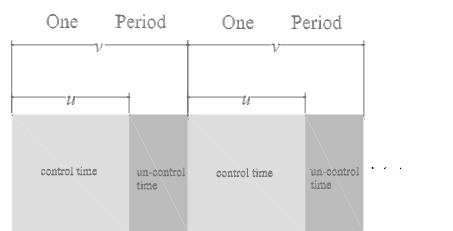


Fig.1. Sketch map of the intermittent control, i.e., the control time is periodic, and in any period, the time is composed of "control time" and "un-control time".

Recently, most of the current researches were primarily concerned with asymptotical or exponential synchronization of networks via intermittent control [29-31]. This indicates that the intermittent control can derive the slave system to synchronization the master system after the infinite horizon. However, in reality, the networks might always be expected to achieve synchronization or stability under a finite time, particularly in engineering fields. In order to achieve synchronization in a given time, the useful and efficient intermittent control methods might be finite-time intermittent control. Finite-time synchronization of complex networks via intermittent control means the optimality in convergence time dealing with intermittent control problems. However, to our best knowledge, there is few published

paper considering finite-time synchronization of complex dynamical networks via periodically intermittent control. Considering the important role of synchronization of complex networks, it is worth studying the finite-time synchronization of complex dynamical networks with intermittent control.

The main contribution of this paper lies in the following aspects. Firstly, a central lemma is proved by using analysis method. Additionally, two different intermittent controllers are designed to synchronize the addressed complex networks and some useful finite-time criteria are obtained. Subsequently, some sufficient conditions are also derived in terms of linear matrix inequality, which is very easy to verify. Besides, a very interesting fact is revealed that the more the control gains are, the longer in time to achieve finite-time synchronization will be.

The rest of this paper is organized as follows. In Section 2, the synchronization problem of a general complex network is formulated, and some useful lemmas and preliminaries are given. Finite-time synchronization of two complex dynamical networks by intermittent adaptive control theory is rigorously derived in Section 3. Section 4 studies the finite-time synchronization of drive-response networks by using the intermittent updated laws of coupling strengths. Some numerical examples are given for supporting the theory results in Section 5. Conclusions are drawn in Section 6.

2 Preliminaries

In this paper, we firstly introduce a complex network model considered in our work and give some useful mathematical preliminaries.

Consider a complex dynamical network consisting of N nodes, in which each node is an n -dimensional dynamical system. The state equation of the entire network is given as:

$$\dot{x}_i(t) = f(x_i(t)) + c \sum_{j=1}^N a_{ij} \Gamma x_j(t), \quad i = 1, 2, \dots, N, \quad (2.1)$$

where $x_i(t) = (x_{i1}(t), x_{i2}(t), \dots, x_{in}(t))^T \in R^n$ is the state vector of the i th dynamical node, $f: R^n \rightarrow R^n$ standing for the activity of an individual subsystem is a vector value function, the constant $c > 0$ is a coupling strength. $\Gamma \in R^{n \times n}$ is a positive definite matrix which describe the individual coupling between node i and node j and $A = (a_{ij}) \in R^{N \times N}$ is the coupling configuration matrix. If there is a connection from the nodes i to j ($j \neq i$), then $a_{ij} > 0$; otherwise, $a_{ij} = 0$ ($j \neq i$), and the diagonal elements of matrices A is defined as

$$a_{ii} = - \sum_{j=1, j \neq i}^N a_{ij}.$$

Throughout this paper, we have the following assumption.

Assumption 1. ([44]) Assume that there exists a positive definite diagonal matrix $P = \text{diag}(p_1, \dots, p_n)$ and a diagonal matrix $\Theta = \text{diag}(\theta_1, \dots, \theta_n)$, such that $f(\cdot)$ satisfies the following inequality:

$$(y - x)^T P(f(y) - f(x) - \Theta(y - x)) \leq -\xi(y - x)^T (y - x), \quad (2.2)$$

for some $\xi > 0$, all $x, y \in R^n$ and $t > 0$.

We can obtain that many well-known chaotic systems such as the Chua's oscillators, Rössler systems, Lorenz systems, Chen systems, and Lü systems satisfy Assumption 1.

Lemma 1 ([41]). Assume that a continuous, positive-definite function $V(t)$ satisfies the following inequality:

$$\dot{V}(t) \leq -\alpha V^\eta(t), \quad \forall t \geq t_0, \quad V(t_0) \geq 0,$$

where $\alpha > 0, 0 < \eta < 1$ are two constants. Then, for any given $t_0, V(t)$ satisfies the following inequality:

$$V^{1-\eta}(t) \leq V^{1-\eta}(t_0) - \alpha(1-\eta)(t-t_0), \quad t_0 \leq t \leq t_1,$$

and

$$V(t) \equiv 0, \quad \forall t \geq t_1,$$

with t_1 given by

$$t_1 = t_0 + \frac{V^{1-\eta}(t_0)}{\alpha(1-\eta)}.$$

Lemma 2 ([48]). Let $x_1, x_2, \dots, x_n \in R^m$ are any vectors and $0 < q < 2$ is a real number satisfying:

$$\|x_1\|^q + \|x_2\|^q + \dots + \|x_n\|^q \geq (\|x_1\|^2 + \|x_2\|^2 + \dots + \|x_n\|^2)^{q/2}. \quad (2.3)$$

Lemma 3. Suppose that function $V(t)$ is continuous and non-negative when $t \in [0, \infty)$ and satisfies the following conditions:

$$\begin{cases} \dot{V}(t) \leq -\alpha V^\eta(t), & lT \leq t \leq lT + \theta T, \\ \dot{V}(t) \leq 0, & lT + \theta T \leq t < (l+1)T, \end{cases} \quad (2.4)$$

where $\alpha > 0, T > 0, 0 < \eta, \theta < 1, l \in \ell = \{0, 1, 2, \dots, p\}$ is a finite natural number set and p is a positive integer, then the following inequality holds:

$$V^{1-\eta}(t) \leq V^{1-\eta}(0) - \alpha\theta(1-\eta)t, \quad 0 \leq t \leq \bar{T}, \quad (2.5)$$

for the constant \bar{T} is the setting time.

Proof. Take $M_0 = V^{1-\eta}(0)$ and $W(t) = V^{1-\eta}(t) + \alpha(1-\eta)t$, where $t \geq 0$. Let $Q(t) = W(t) - hM_0$, where $h > 1$ is a constant. It is easy to see that

$$Q(t) < 0, \quad \text{for } t = 0. \quad (2.6)$$

Next, we will prove that

$$Q(t) < 0, \quad \text{for all } t \in [0, \theta T]. \quad (2.7)$$

Otherwise, there exists a $t_0 \in [0, \theta T)$ such that

$$Q(t_0) = 0, \quad \dot{Q}(t_0) > 0, \quad (2.8)$$

$$Q(t) < 0, \quad 0 \leq t < t_0. \quad (2.9)$$

Using Eqs. (2.6), (2.8) and (2.9), we obtain

$$\begin{aligned} \dot{Q}(t_0) &= (1 - \eta)V^{-\eta}(t)\dot{V}(t) + \alpha(1 - \eta) \\ &\leq (1 - \eta)V^{-\eta}(t)(-\alpha V^\eta(t)) + \alpha(1 - \eta) \\ &= -\alpha(1 - \eta) + \alpha(1 - \eta) = 0 \end{aligned} \quad (2.10)$$

This contradicts the second inequality in (2.8), and so (2.7) holds.

Now, we prove that for $t \in [\theta T, T)$

$$H(t) = W(t) - hM_0 - \alpha(1 - \eta)(t - \theta T) < 0. \quad (2.11)$$

Otherwise, there exists a $t_1 \in [\theta T, T)$ such that

$$H(t_1) = 0, \quad \dot{H}(t_1) > 0, \quad (2.12)$$

$$H(t) < 0, \quad \theta T \leq t < t_1. \quad (2.13)$$

Using Eqs. (2.12) and (2.13), we have

$$\begin{aligned} \dot{H}(t_1) &= (1 - \eta)V^{-\eta}(t_1)\dot{V}(t_1) + \alpha(1 - \eta) - \alpha(1 - \eta) \\ &\leq \alpha(1 - \eta) - \alpha(1 - \eta) \\ &= 0, \end{aligned} \quad (2.14)$$

which contradicts the second inequality in (2.12). Hence (2.11) holds.

Consequently, on the one hand, for $t \in [\theta T, T)$,

$$W(t) < hM_0 + \alpha(1 - \eta)(t - \theta T) \leq hM_0 + \alpha(1 - \eta)(1 - \theta)T.$$

On the other hand, it follows from Eqs. (2.6) and (2.7) that for $t \in [0, \theta T)$

$$W(t) < hM_0 < hM_0 + \alpha(1 - \eta)(1 - \theta)T.$$

So

$$W(t) < hM_0 + \alpha(1 - \eta)(1 - \theta)T, \quad \text{for all } t \in [0, T).$$

Similarly, we can prove that for $t \in [T, (1 + \theta)T)$,

$$W(t) < hM_0 + \alpha(1 - \eta)(1 - \theta)T,$$

and for $t \in [(1 + \theta)T, 2T)$

$$\begin{aligned} W(t) &< hM_0 + \alpha(1 - \eta)(1 - \theta)T + \alpha(1 - \eta)(t - \theta T - T) \\ &= hM_0 + \alpha(1 - \eta)(t - 2\theta T). \end{aligned}$$

In the following, we can derive the following estimation of $W(t)$ for any integers l .

For $lT \leq t < (l + \theta)T$,

$$W(t) < hM_0 + \alpha(1 - \eta)(1 - \theta)lT, \quad (2.15)$$

and for $(l + \theta)T \leq t < (l + 1)T$,

$$W(t) < hM_0 + \alpha(1 - \eta)[t - (l + 1)\theta T]. \quad (2.16)$$

Since for any $t \geq 0$, there exists a natural integer $k \in \mathbb{N}$, such that $kT \leq t < (k + 1)T$, by induction, we can conclude the following estimation of $W(t)$ for any t by Eqs. (2.15) and (2.16).

For $kT \leq t < (k + \theta)T$,

$$W(t) < hM_0 + \alpha(1 - \eta)(1 - \theta)kT \leq hM_0 + \alpha(1 - \eta)(1 - \theta)t,$$

and for $(k + \theta)T \leq t < (k + 1)T$,

$$W(t) < hM_0 + \alpha(1 - \eta)[t - (k + 1)\theta T] \leq hM_0 + \alpha(1 - \eta)(1 - \theta)t.$$

Let $h \rightarrow 1$, from the definition of $W(t)$, we obtain

$$V^{1-\eta}(t) \leq V^{1-\eta}(0) - \alpha(1 - \eta)t + \alpha(1 - \eta)(1 - \theta)t = V^{1-\eta}(0) - \alpha\theta(1 - \eta)t, \quad t \geq 0.$$

Lemma 4. When $a \leq b$ and $c < 1$ are all positive numbers, then the following inequality holds:

$$a^c \leq b^c.$$

The proof is easily obtained, and this proofs is omitted.

3. Finite-time synchronization of complex networks via intermittent adaptive control

In this section, we consider the finite-time synchronization of general complex dynamical networks via periodically intermittent control.

For simplicity, regarding model (2.1) as the master (or drive) system, and the response (or slave) system is given by:

$$\dot{y}_i(t) = f(y_i(t)) + c \sum_{j=1}^N a_{ij} \Gamma y_j(t) + u_i(t), \quad i = 1, 2, \dots, N, \quad (3.1)$$

where $y_i(t) = (y_{i1}(t), y_{i2}, \dots, y_{in}(t))^T \in R^n$ is the response state vector of the node i . $u(t) = (u_1(t), u_2(t), \dots, u_N(t))^T$ is an intermittent controller defined by

$$\begin{cases} u_i(t) = r_i e_i - \bar{k} \frac{\sqrt{\lambda_{\max}(P)}}{\lambda_{\min}(P)} \text{sign}(e_i), & lT \leq t < lT + \delta, \\ u_i(t) = 0, & \|e_i\| = 0 \text{ or } lT + \delta \leq t < (l+1)T, \end{cases} \quad (3.2)$$

where r_i ($i = 1, 2, \dots, N$) are the adaptive update laws, $\text{sign}(\cdot)$ function $\text{sign}(e_i) = (\text{sign}(e_{i1}), \text{sign}(e_{i2}), \dots, \text{sign}(e_{in}))^T$. $l \in \ell = \{0, 1, 2, \dots, p\}$ is a finite natural number set and p is a positive integer, $\bar{k} > 0$ is a tunable constant and real number, $T > 0$ is the control period, $\delta > 0$ is called the control width (control duration). P is a positive definite diagonal matrix. Denote $\lambda_{\max}(P)$ ($\lambda_{\min}(P)$) as the maximum (minimum) eigenvalue of the positive definite diagonal matrix P .

Assume that $x(t) = (x_1(t), \dots, x_N(t))^T$ and $y(t) = (y_1(t), \dots, y_N(t))^T$ are solutions of system (2.1) and (3.1). Further, $e_i(t) = y_i(t) - x_i(t)$ be synchronization error between the states of drive system (2.1) and the response system (3.1), and $\theta = \delta/T$ be the ratio of the control width δ to the control period T called control rate, then the error dynamical system can be derived as

$$\begin{cases} \dot{e}_i = f(y_i) - f(x_i) + c \sum_{j=1}^N a_{ij} \Gamma e_j + u_i, \\ lT \leq t < lT + \theta T, \quad i = 1, 2, \dots, N, \\ \dot{e}_i = f(y_i) - f(x_i) + c \sum_{j=1}^N a_{ij} \Gamma e_j, \\ lT + \theta T \leq t < (l+1)T, \quad i = 1, 2, \dots, N. \end{cases} \quad (3.3)$$

Meanwhile, the intermittent updated laws are taken as follows:

$$\begin{cases} \dot{r}_i = -\lambda_i \left[\frac{\eta_i r_i e_i^T P e_i}{r_i^2} + e_i^T P e_i + \frac{\bar{k}}{\sqrt{\lambda_i}} \text{sign}(r_i) \right], & lT \leq t < lT + \delta, \\ \dot{r}_i = 0, & \|r_i\| = 0 \text{ or } lT + \delta \leq t < (l+1)T, \end{cases} \quad (3.4)$$

where $\eta_i > 0$ ($i = 1, 2, \dots, N$) is a positive constant called control gain, $\lambda_i > 0$ is an arbitrary positive constant.

Then we can derive the following periodically intermittent synchronization criteria.

Theorem 1. Suppose Assumption 1 holds. If there exist a positive definite matrix $\Xi =$

$diag(\eta_1, \eta_2, \dots, \eta_N) > 0$ and a positive definite diagonal matrix $P > 0$ such that the following conditions hold:

$$\theta_j I_N - \Xi + c\gamma_j A \leq 0, \quad j = 1, 2, \dots, n, \quad (3.5)$$

$$\theta_j I_N - \frac{\xi}{\lambda_{\max}(P)} I_N + c\gamma_j A \leq 0, \quad j = 1, 2, \dots, n, \quad (3.6)$$

where $\Gamma = diag(\gamma_1, \gamma_2, \dots, \gamma_n)$, $\Theta = diag(\theta_1, \theta_2, \dots, \theta_n)$, ξ is a positive constant satisfies Assumption 1, and I_N is an appropriate identity matrix. Then the error system (3.3) is synchronized under the periodically intermittent controller (3.2) and the intermittent updated law (3.4) in a finite time:

$$T_1 = \frac{2V^{1/2}(0)}{\sqrt{2k\theta}}, \quad (3.7)$$

where $V(0) = \frac{1}{2} \sum_{i=1}^N e_i^T(0) P e_i(0) + \frac{1}{2} \sum_{i=1}^N \frac{1}{\lambda_i} (r_i(0))^2$, $e_i(0)$, $r_i(0)$ are the initial conditions of e_i , r_i .

Proof. We design the following Lyapunov function:

$$V(t) = \frac{1}{2} \sum_{i=1}^N e_i^T P e_i + \frac{1}{2} \sum_{i=1}^N \frac{1}{\lambda_i} r_i^2. \quad (3.8)$$

Then the derivative of $V(t)$ with respect to time t along the solutions of Eq. (3.3) can be calculated as follows:

When $lT \leq t < (l + \theta)T$, for $l \in \ell$,

$$\begin{aligned} \dot{V}(t) &= \sum_{i=1}^N e_i^T P \dot{e}_i + \sum_{i=1}^N \frac{1}{\lambda_i} r_i \dot{r}_i \\ &= \sum_{i=1}^N e_i^T P [f(y_i) - f(x_i) + c \sum_{j=1}^N a_{ij} \Gamma e_j + u_i] - \sum_{i=1}^N r_i \left[\frac{\eta_i r_i e_i^T P e_i}{r_i^2} + \right. \\ &\quad \left. \frac{\bar{k}}{\sqrt{\lambda_i}} \text{sign}(r_i) + e_i^T P e_i \right] \\ &= \sum_{i=1}^N \{ e_i^T P [f(y_i) - f(x_i) - \Theta e_i] + e_i^T P \Theta e_i + e_i^T P u_i + \\ &\quad c e_i^T P \sum_{j=1}^N a_{ij} \Gamma e_j \} - \sum_{i=1}^N r_i \left[\frac{\eta_i r_i e_i^T P e_i}{r_i^2} + \frac{\bar{k}}{\sqrt{\lambda_i}} \text{sign}(r_i) + e_i^T P e_i \right] \\ &\leq -\xi \sum_{i=1}^N e_i^T e_i + \sum_{i=1}^N e_i^T (P \Theta - \eta_i P) e_i + c \sum_{i=1}^N e_i^T P \sum_{j=1}^N a_{ij} \Gamma e_j \\ &\quad - \sum_{i=1}^N e_i^T P r_i e_i + \sum_{i=1}^N e_i^T P r_i e_i - \bar{k} \sum_{i=1}^N \frac{\sqrt{\lambda_{\max}(P)}}{\lambda_{\min}(P)} e_i^T P \text{sign}(e_i) \\ &\quad - \sum_{i=1}^N \frac{\bar{k}}{\sqrt{\lambda_i}} |r_i| \end{aligned}$$

$$\begin{aligned} &\leq \sum_{j=1}^n p_j \tilde{e}_j^T (\theta_j I_N - \Xi + c\gamma_j A) \tilde{e}_j - \bar{k} \sum_{i=1}^N \frac{\sqrt{\lambda_{\max}(P)}}{\lambda_{\min}(P)} e_i^T P \text{sign}(e_i) \\ &\quad - \sum_{i=1}^N \frac{\bar{k}}{\sqrt{\lambda_i}} |r_i|, \end{aligned}$$

where $\tilde{e}_j(t) = [\tilde{e}_{j1}, \tilde{e}_{j2}, \dots, \tilde{e}_{jN}]^T$ be the column vector of the i th $e(t)$.

Using $\text{sign}(e_i) = (\text{sign}(e_{i1}), \text{sign}(e_{i2}), \dots, \text{sign}(e_{iN}))^T$ and Lemma 2, we have

$$\dot{V}(t) \leq \sum_{j=1}^n p_j \tilde{e}_j^T Z_j \tilde{e}_j - \bar{k} \sum_{i=1}^N \sqrt{\lambda_{\max}(P)} \|e_i\| - \sum_{i=1}^N \frac{\bar{k}}{\sqrt{\lambda_i}} |r_i|,$$

Z_j is defined as

$$Z_j = \theta_j I_N - \Xi + c\gamma_j A, \quad j = 1, 2, \dots, n.$$

Using Lemma 2 and Lemma 4, we have

$$\dot{V}(t) \leq \sum_{j=1}^n p_j \tilde{e}_j^T Z_j \tilde{e}_j - \sqrt{2} \bar{k} \left(\frac{1}{2} \sum_{i=1}^N e_i^T P e_i + \frac{1}{2} \sum_{i=1}^N \frac{1}{\lambda_i} r_i^2 \right)^{\frac{1}{2}}.$$

It follows from condition (3.5) that

$$Z_j \leq 0,$$

which shows that $\dot{V}(t) \leq -\sqrt{2} \bar{k} V^{\frac{1}{2}}(t)$.

When $(l + \theta)T \leq t < (l + 1)T$, for $l \in \ell$, we have

$$\begin{aligned} \dot{V}(t) &= \sum_{i=1}^N e_i^T P \dot{e}_i \\ &= \sum_{i=1}^N \{e_i^T P [f(y_i) - f(x_i) - \Theta e_i] + e_i^T P \Theta e_i + c \sum_{j=1}^N a_{ij} e_i^T P \Gamma e_j\} \\ &\leq -\xi \sum_{i=1}^N e_i^T e_i + \sum_{j=1}^n p_j \tilde{e}_j^T \theta_j I_N \tilde{e}_j + c \sum_{j=1}^n p_j \tilde{e}_j^T \gamma_j A \tilde{e}_j \\ &\leq -\sum_{j=1}^n p_j \tilde{e}_j^T \frac{\xi}{\lambda_{\max}(P)} I_N \tilde{e}_j + \sum_{j=1}^n p_j \tilde{e}_j^T \theta_j I_N \tilde{e}_j + c \sum_{j=1}^n p_j \tilde{e}_j^T \gamma_j A \tilde{e}_j \\ &= \sum_{j=1}^n p_j \tilde{e}_j^T S_j \tilde{e}_j, \end{aligned}$$

where $\tilde{e}_j(t) = [\tilde{e}_{j1}, \tilde{e}_{j2}, \dots, \tilde{e}_{jN}]^T$ be the column vector of the i th $e(t)$, S_j is defined as

$$S_j = \theta_j I_N - \frac{\xi}{\lambda_{\max}(P)} I_N + c\gamma_j A, \quad j = 1, 2, \dots, n.$$

It follows from condition (3.6) that

$$S_j \leq 0,$$

which shows that $\dot{V}(t) \leq 0$.

Namely, let $\alpha = \sqrt{2k}$, $1 - \eta = \frac{1}{2}$, we have

$$\begin{cases} \dot{V}(t) \leq -\alpha V^\eta(t), & lT \leq t \leq lT + \theta T, \\ \dot{V}(t) \leq 0, & lT + \theta T \leq t < (l+1)T. \end{cases} \quad (3.9)$$

This implies the conclusion by Lemma 3 and the proof is complete.

Remark 1. Obviously, when $\theta = 1$, the intermittent control (3.2) is degenerated to a continuous control input which has been extensively proposed in previous work (see [15, 41-48]). And in this case, it is clear from Theorem 1 that system (2.1) and system (3.1) are finite-time synchronized under Assumption 1 based on the continuous control and the main results in Refs. [41-48] can be easily analogized from the simplified criteria despite model differences.

Remark 2. In this paper, we introduce finite-time intermittent control technique to synchronize complex networks. In the literature, there are many results concerning exponential synchronization via intermittent control (see [29-32, 39-40]). However, to the best of our knowledge, there is few published paper dealing with finite-time synchronization between two complex networks via intermittent control. Obviously, our results can enrich the finite-time synchronization problem of network by the intermittent control. What's more, compare with the exponential synchronization between two complex networks via intermittent control, the setting time of synchronization by finite-time intermittent control is estimated in finite-time.

Remark 3. From (3.7), we can see the roles that the control rate θ and the tunable constant \bar{k} are played an important role in finite-time synchronization of complex networks. The equation (3.7) indicates that the convergence time is changing when the values of \bar{k} and θ increases. Denote $T(\theta) = \frac{2V^{\frac{1}{2}}(0)}{\sqrt{2\bar{k}\theta}}$, if we fixed the \bar{k} , then we have, $T'(\theta) = -\frac{2V^{\frac{1}{2}}(0)}{\sqrt{2\bar{k}\theta^2}} < 0$. Obviously, $T(\theta)$ ($0 < \theta < 1$) is a strictly monotone decreasing function. Hence, the larger the control rate is, the shorter in the convergence time to achieve finite-time synchronization of complex networks will be. In addition, the synchronization for the complex dynamical networks with a continuous control input (where $\theta = 1$) is faster in the convergence time than those of complex networks with periodically intermittent control ($0 < \theta < 1$).

Remark 4. In Ref. [15], U.E. Vincent et.al. studied the finite-time synchronization for a class of chaotic and hyperchaotic systems via adaptive feedback controller. To derive their results, $l = \frac{\lambda}{c} > 0$ is selected such that $V^\eta(w, k_1) \leq lV_1^\eta(w)$ in Ref. [15]. Evidently, the application scope and generalization domain of the results are limited due to a suitable

parameter should be found. Especially, the system is not a chaotic system, but a class of complex dynamical networks or neural networks, the method of Ref. [15] is not suitable. However, our results do not need this restriction by introducing a new controller and a novel updated law.

In Ref. [34], Zhu et al. investigated the stabilization and synchronization of chaotic systems via intermittent control. He only concerned asymptotically stable under adaptive intermittent controlling. In Theorem 1, we introduce a novel periodically intermittent adaptive controlling to guarantee finite-time synchronization.

Remark 5. From the inequality (3.5), we have $\lambda_{\max}(\Xi) \geq c\lambda_{\max}(\gamma_j A) + \max_{1 \leq j \leq n} \theta_j$, from the inequality (3.6), we have $\frac{\xi}{\lambda_{\max}(P)} \geq c\lambda_{\max}(\gamma_j A) + \max_{1 \leq j \leq n} \theta_j$. Since $\lambda_{\max}(\Xi)$ and $\frac{\xi}{\lambda_{\max}(P)}$ are two different values, then the two conditions are needed in Theorem 1. But, if the two different values satisfy dependent relation, (3.5) and (3.6) are no difference in some sense. If $\lambda_{\max}(\Xi) \geq \frac{\xi}{\lambda_{\max}(P)}$, then we have the following Corollary 1. If $\frac{\xi}{\lambda_{\max}(P)} \geq \lambda_{\max}(\Xi)$, then we have the following Corollary 2.

Corollary 1. Suppose Assumption 1 holds. If there exist a positive definite diagonal matrix $P > 0$ and $\lambda_{\max}(\Xi) \geq \frac{\xi}{\lambda_{\max}(P)}$ ($\Xi = \text{diag}(\eta_1, \eta_2, \dots, \eta_N) > 0$) such that the following conditions hold:

$$\theta_j I_N - \frac{\xi}{\lambda_{\max}(P)} I_N + c\gamma_j A \leq 0, \quad j = 1, 2, \dots, n, \quad (3.10)$$

where $\Gamma = \text{diag}(\gamma_1, \gamma_2, \dots, \gamma_n)$, $\Theta = \text{diag}(\theta_1, \theta_2, \dots, \theta_n)$, ξ is a positive constant satisfies Assumption 1, and I_N is an appropriate identity matrix. Then the error system (3.3) is synchronized under the periodically intermittent controller (3.2) and the intermittent updated law (3.4) in a finite time:

$$T_1 = \frac{2V^{1/2}(0)}{\sqrt{2k\theta}}, \quad (3.11)$$

where $V(0) = \frac{1}{2} \sum_{i=1}^N e_i^T(0) P e_i(0) + \frac{1}{2} \sum_{i=1}^N \frac{1}{\lambda_i} (r_i(0))^2$, $e_i(0)$, $r_i(0)$ are the initial conditions of e_i , r_i .

Corollary 2. Suppose Assumption 1 holds. If there exist a positive definite diagonal matrix $P > 0$ and $\frac{\xi}{\lambda_{\max}(P)} \geq \lambda_{\max}(\Xi)$ ($\Xi = \text{diag}(\eta_1, \eta_2, \dots, \eta_N) > 0$) such that the following conditions hold:

$$\theta_j I_N - \Xi + c\gamma_j A \leq 0, \quad j = 1, 2, \dots, n, \quad (3.12)$$

where $\Gamma = \text{diag}(\gamma_1, \gamma_2, \dots, \gamma_n)$, $\Theta = \text{diag}(\theta_1, \theta_2, \dots, \theta_n)$, ξ is a positive constant satisfies Assumption 1, and I_N is an appropriate identity matrix. Then the error system (3.3) is synchronized under the periodically intermittent controller (3.2) and the intermittent updated

law (3.4) in a finite time:

$$T_1 = \frac{2V^{1/2}(0)}{\sqrt{2\bar{k}\theta}}, \quad (3.13)$$

where $V(0) = \frac{1}{2} \sum_{i=1}^N e_i^T(0) P e_i(0) + \frac{1}{2} \sum_{i=1}^N \frac{1}{\lambda_i} (r_i(0))^2$, $e_i(0)$, $r_i(0)$ are the initial conditions of e_i , r_i .

If we let the adaptive update laws $r_i = 0$ ($i = 1, 2, \dots, N$), for all $t \geq 0$, then we have the following Corollary 3.

Corollary 3. Suppose Assumption 1 holds. If there exist a positive definite matrix $\Xi = \text{diag}(\eta_1, \eta_2, \dots, \eta_N) > 0$ and a positive definite diagonal matrix $P > 0$ such that the following conditions hold:

$$\theta_j I_N - \Xi + c\gamma_j A < 0, \quad j = 1, 2, \dots, n, \quad (3.14)$$

$$\theta_j I_N - \frac{\xi}{\lambda_{\max}(P)} I_N + c\gamma_j A < 0, \quad j = 1, 2, \dots, n, \quad (3.15)$$

where $\Gamma = \text{diag}(\gamma_1, \gamma_2, \dots, \gamma_n)$, $\Theta = \text{diag}(\theta_1, \theta_2, \dots, \theta_n)$, ξ is a positive constant satisfies Assumption 1, and I_N is an appropriate identity matrix. Then the error system (3.3) is synchronized under the periodically intermittent controller

$$\begin{cases} u_i(t) = -\eta_i e_i - \bar{k} \frac{\sqrt{\lambda_{\max}(P)}}{\lambda_{\min}(P)} \text{sign}(e_i), & lT \leq t < lT + \delta, \\ u_i(t) = 0, & \|e_i\| = 0 \text{ or } lT + \delta \leq t < (l+1)T, \end{cases} \quad (3.16)$$

in a finite time:

$$T_2 = \frac{2V^{1/2}(0)}{\sqrt{2\bar{k}\theta}}, \quad (3.17)$$

where $V(0) = \frac{1}{2} \sum_{i=1}^N e_i^T(0) P e_i(0)$, $e_i(0)$ is the initial condition of $e_i(t)$.

The proof of this corollary is easily obtained from Theorem 1. Thus it is omitted here.

4. Finite-time synchronization of complex networks via intermittent feedback control

In this section, we shall focus on analyzing finite-time synchronization between two complex dynamical networks via periodically intermittent feedback control, and found an adaptive law to automatically force the error systems converge to zero within a finite time and remain on it forever.

Now, we consider the following two complex dynamical networks:

$$\begin{cases} \dot{x}_i = f(x_i) + c \sum_{j=1}^N a_{ij} \Gamma x_j - r_i e_i, & i = 1, 2, \dots, N, \\ \dot{y}_i = f(y_i) + c \sum_{j=1}^N a_{ij} \Gamma y_j + k_i e_i, & i = 1, 2, \dots, N, \end{cases} \quad (4.1)$$

where r_i, k_i are the intermittent feedback gain which can be derived as:

$$\begin{cases} \dot{k}_i = -\psi_i[e_i^T P e_i + \frac{\bar{k} k_i \sqrt{\lambda_{\max}(P)} e_i^T}{k_i^2} \text{sign}(e_i) + \frac{\bar{k}}{\sqrt{\psi_i}} \text{sign}(k_i)], & lT \leq t < lT + \delta, \\ \dot{k}_i = 0, & lT + \delta \leq t < (l+1)T, \end{cases} \quad (4.2)$$

$$\begin{cases} \dot{r}_i = -\varphi_i[e_i^T P e_i + \frac{r_i \eta_i e_i^T P e_i}{r_i^2} + \frac{\bar{k}}{\sqrt{\varphi_i}} \text{sign}(r_i)], & lT \leq t < lT + \delta, \\ \dot{r}_i = 0, & lT + \delta \leq t < (l+1)T, \end{cases} \quad (4.3)$$

for $\psi_i, \varphi_i > 0$ ($i = 1, 2, \dots, N$) are arbitrary constants, $\bar{k} > 0$ is a tunable constant, $\Xi = \text{diag}(\eta_1, \eta_2, \dots, \eta_N) > 0$ is a positive definite matrix.

Then we can derive the following result.

Theorem 2. Suppose Assumption 1 holds. If there exist a positive definite matrix $\Xi = \text{diag}(\eta_1, \eta_2, \dots, \eta_N) > 0$ and a positive definite diagonal matrix $P > 0$ such that the following conditions hold:

$$\theta_j I_N - \Xi + c\gamma_j A \leq 0, \quad j = 1, 2, \dots, n, \quad (4.4)$$

$$\theta_j I_N - \frac{\xi}{\lambda_{\max}(P)} I_N + c\gamma_j A \leq 0, \quad j = 1, 2, \dots, n, \quad (4.5)$$

where $\Gamma = \text{diag}(\gamma_1, \gamma_2, \dots, \gamma_n), \Theta = \text{diag}(\theta_1, \theta_2, \dots, \theta_n)$, ξ is a positive constant satisfies Assumption 1, and I_N is an appropriate identity matrix. Then under the periodically intermittent feedback gain control (4.2) and (4.3), the two complex networks are synchronized in a finite time:

$$T_3 = \frac{2V^{1/2}(0)}{\sqrt{2} \bar{k} \theta}, \quad (4.6)$$

where $V(0) = \frac{1}{2} \sum_{i=1}^N e_i^T(0) P e_i(0) + \frac{1}{2} \sum_{i=1}^N \frac{1}{\psi_i} (k_i(0))^2 + \frac{1}{2} \sum_{i=1}^N \frac{1}{\varphi_i} (r_i(0))^2$, $e_i(0), k_i(0), r_i(0)$ are the initial conditions of e_i, k_i, r_i .

Proof. According to the the systems (4.1), then error dynamical system can be derived as

$$\begin{cases} \dot{e}_i = f(y_i) - f(x_i) + c \sum_{j=1}^N a_{ij} \Gamma e_j(t) + k_i e_i + r_i e_i, \\ lT \leq t < lT + \theta T, \quad i = 1, 2, \dots, N, \\ \dot{e}_i = f(y_i) - f(x_i) + c \sum_{j=1}^N a_{ij} \Gamma e_j, \\ lT + \theta T \leq t < (l+1)T, \quad i = 1, 2, \dots, N. \end{cases} \quad (4.7)$$

We choose a positive definite function in the form of

$$V(t) = \frac{1}{2} \sum_{i=1}^N e_i^T P e_i + \frac{1}{2} \sum_{i=1}^N \frac{1}{\psi_i} k_i^2 + \frac{1}{2} \sum_{i=1}^N \frac{1}{\varphi_i} r_i^2. \quad (4.8)$$

Then the derivative of $V(t)$ with respect to time t along the solutions of Eq. (4.7) can be calculated as follows:

When $lT \leq t < (l + \theta)T$, for $l \in \ell$,

$$\begin{aligned}
\dot{V}(t) &= \sum_{i=1}^N e_i^T P e_i + \sum_{i=1}^N \frac{1}{\psi_i} k_i \dot{k}_i + \sum_{i=1}^N \frac{1}{\varphi_i} r_i \dot{r}_i \\
&= \sum_{i=1}^N e_i^T P [f(y_i) - f(x_i) + c \sum_{j=1}^N a_{ij} \Gamma e_j + k_i e_i + r_i e_i] - \sum_{i=1}^N k_i [e_i^T P e_i + \\
&\quad \frac{\bar{k}}{\sqrt{\psi_i}} \text{sign}(k_i) + \frac{\bar{k} k_i \sqrt{\lambda_{\max}(P)} e_i^T}{k_i^2} \text{sign}(e_i)] - \sum_{i=1}^N r_i [e_i^T P e_i + \frac{r_i \eta_i e_i^T P e_i}{r_i^2} + \\
&\quad \frac{\bar{k}}{\sqrt{\varphi_i}} \text{sign}(r_i)] \\
&= \sum_{i=1}^N e_i^T P [f(y_i) - f(x_i) + c \sum_{j=1}^N a_{ij} \Gamma e_j] - \sum_{i=1}^N \bar{k} \sqrt{\lambda_{\max}(P)} e_i^T \text{sign}(e_i) - \\
&\quad \sum_{i=1}^N \frac{\bar{k}}{\sqrt{\psi_i}} k_i \text{sign}(k_i) - \sum_{i=1}^N e_i^T \eta_i P e_i - \sum_{i=1}^N \frac{\bar{k}}{\sqrt{\varphi_i}} r_i \text{sign}(r_i) \\
&= \sum_{i=1}^N \{e_i^T P [f(y_i) - f(x_i) - \Theta e_i] + e_i^T P \Theta e_i + c \sum_{j=1}^N a_{ij} \Gamma e_j\} - \sum_{i=1}^N \frac{\bar{k}}{\sqrt{\psi_i}} k_i \text{sign}(k_i) \\
&\quad - \sum_{i=1}^N \bar{k} \sqrt{\lambda_{\max}(P)} e_i^T \text{sign}(e_i) - \sum_{i=1}^N e_i^T \eta_i P e_i - \sum_{i=1}^N \frac{\bar{k}}{\sqrt{\varphi_i}} r_i \text{sign}(r_i) \\
&\leq -\xi \sum_{i=1}^N e_i^T e_i + \sum_{i=1}^N e_i^T (P \Theta - \eta_i P) e_i + c \sum_{i=1}^N e_i^T P \sum_{j=1}^N a_{ij} \Gamma e_j - \sum_{i=1}^N \frac{\bar{k}}{\sqrt{\psi_i}} k_i \text{sign}(k_i) \\
&\quad - \sum_{i=1}^N \bar{k} \sqrt{\lambda_{\max}(P)} e_i^T \text{sign}(e_i) - \sum_{i=1}^N \frac{\bar{k}}{\sqrt{\varphi_i}} r_i \text{sign}(r_i),
\end{aligned}$$

where $\tilde{e}_j = [\tilde{e}_{j1}, \tilde{e}_{j2}, \dots, \tilde{e}_{jN}]^T$ be the column vector of the i th $e(t)$.

Using $\text{sign}(e_i) = (\text{sign}(e_{i1}), \text{sign}(e_{i2}), \dots, \text{sign}(e_{iN}))^T$ and Lemma 2, we have

$$\dot{V}(t) \leq \sum_{j=1}^n p_j \tilde{e}_j^T Z_j \tilde{e}_j - \bar{k} \sum_{i=1}^N \sqrt{\lambda_{\max}(P)} \|e_i\| - \sum_{i=1}^N \frac{\bar{k}}{\sqrt{\psi_i}} |k_i| - \sum_{i=1}^N \frac{\bar{k}}{\sqrt{\varphi_i}} |r_i|,$$

Z_j is defined as

$$Z_j = \theta_j I_N - \Xi + c \gamma_j A, \quad j = 1, 2, \dots, n.$$

Using Lemma 2 and Lemma 4, we have

$$\dot{V}(t) \leq \sum_{j=1}^n p_j \tilde{e}_j^T Z_j \tilde{e}_j - \sqrt{2} \bar{k} \left(\frac{1}{2} \sum_{i=1}^N e_i^T P e_i + \frac{1}{2} \sum_{i=1}^N \frac{1}{\psi_i} k_i^2 + \frac{1}{2} \sum_{i=1}^N \frac{1}{\varphi_i} r_i^2 \right)^{\frac{1}{2}}.$$

It follows from condition (4.4) that

$$Z_j \leq 0,$$

which shows that $\dot{V}(t) \leq -\sqrt{2} \bar{k} V^{\frac{1}{2}}(t)$.

When $(l + \theta)T \leq t < (l + 1)T$, for $l \in \ell$, we have

$$\begin{aligned} \dot{V}(t) &= \sum_{i=1}^N e_i^T P \dot{e}_i \\ &= \sum_{i=1}^N \{e_i^T P [f(y_i) - f(x_i) - \Theta e_i] + e_i^T P \Theta e_i + c \sum_{j=1}^N a_{ij} e_i^T P \Gamma e_j\} \\ &\leq -\xi \sum_{i=1}^N e_i^T e_i + \sum_{j=1}^n p_j \tilde{e}_j^T \theta_j I_N \tilde{e}_j + c \sum_{j=1}^n p_j \tilde{e}_j^T \gamma_j A \tilde{e}_j \\ &\leq -\sum_{j=1}^n p_j \tilde{e}_j^T \frac{\xi}{\lambda_{\max}(P)} I_N \tilde{e}_j + \sum_{j=1}^n p_j \tilde{e}_j^T \theta_j I_N \tilde{e}_j + c \sum_{j=1}^n p_j \tilde{e}_j^T \gamma_j A \tilde{e}_j \\ &= \sum_{j=1}^n p_j \tilde{e}_j^T S_j \tilde{e}_j, \end{aligned}$$

where $\tilde{e}_j = [\tilde{e}_{j1}, \tilde{e}_{j2}, \dots, \tilde{e}_{jN}]^T$ be the column vector of the i th $e(t)$, S_j is defined as

$$S_j = \theta_j I_N - \frac{\xi}{\lambda_{\max}(P)} I_N + c \gamma_j A, \quad j = 1, 2, \dots, n.$$

It follows from condition (4.5) that

$$S_j \leq 0,$$

which shows that $\dot{V}(t) \leq 0$.

Namely, let $\alpha = \sqrt{2} \bar{k}$, $1 - \eta = \frac{1}{2}$, we have

$$\begin{cases} \dot{V}(t) \leq -\alpha V^\eta(t), & lT \leq t \leq lT + \theta T, \\ \dot{V}(t) \leq 0, & lT + \theta T \leq t < (l + 1)T. \end{cases} \quad (4.9)$$

According to Lemma 4, Then this completes the proof.

Remark 6. The intermittent feedback gains k_i and r_i are automatically adapted to the suitable strength k_0 and r_0 , respectively, which depend on the initial values. In Ref. [17], Xu et. al. considered the two complex networks by the adjustment of feedback gains k_i and r_i such that the error system (8) asymptotically converge to zero. However, in Theorem 2 of this paper, we can consider the same model by using the adapted of feedback gains k_i and r_i drive the error system (4.7) in finite time converge to zero. What's more, the Theorem 2 of this paper is suitable to a great many complex dynamical networks, neural networks and chaotic systems concerning about intermittent control. For example, the method of Theorem

2 in this paper is suitable to the main results of Theorems and corollaries in Ref. [46].

Remark 7. According to the convergence time T_1, T_2, T_3 , we can conclude the convergence time satisfies: $T_2 \leq T_1 \leq T_3$. In Theorem 1, we did shorten the synchronization or convergence time by using the intermittent adaptive controller in contrast to the Theorem 2, which introduced the intermittent feedback controller, but using many control laws is very expensive in practical applications. Besides, a very interesting phenomenon is revealed that the more the control gains are, the longer in time to achieve finite-time synchronization will be.

5 Numerical examples

In the previous section, theorems essentially provide the criteria for finite-time synchronization. In this section, we use representative example to illustrate how these theorem can be applied to achieve finite-time synchronization in complex network.

We consider two complex networks, in which each subsystem is a Lorenz system. The dynamics of a single Lorenz system is described as

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \end{bmatrix} = \begin{bmatrix} -r_1 & r_1 & 0 \\ r_3 & -1 & 0 \\ 0 & 0 & -r_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + \begin{bmatrix} 0 \\ -x_1x_3 \\ x_1x_2 \end{bmatrix} = f(x), \quad (5.1)$$

where the parameters are selected as $r_1 = 10$, $r_3 = 28$, $r_2 = 8/3$, the single Lorenz system has a chaotic attractor (Fig. 2). We take $P = \text{diag}(5, 4, 2)$ and $\Theta = \text{diag}(3, 3, 3)$ such that the Lorenz system satisfies Assumption 1.

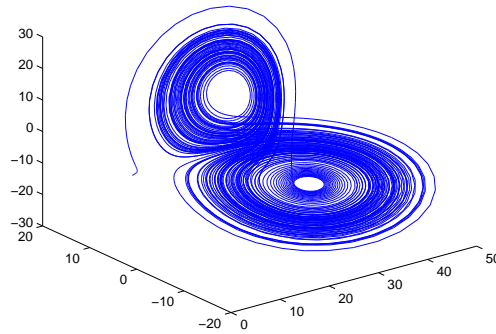


Fig.2. The chaotic trajectories of the Lorenz system.

5.1. The simulation with the periodically intermittent adaptive controllers

Example 1. We consider the directed network with 100 dynamical nodes. $A = (a_{ij})_{100 \times 100}$

is a symmetrically diffusive coupling matrix with $a_{ij} = 0$ or 1 ($j \neq i$), $\Gamma = \text{diag}\{1, 1, 1\}$ and the coupling strength $c = 1$.

According to Theorem 1 in Section 3, one can synchronize the error system (3.3) by design of the following intermittent adaptive controllers and intermittent updated laws

$$\begin{cases} u_i(t) = r_i e_i - \bar{k} \frac{\sqrt{\lambda_{\max}(P)}}{\lambda_{\min}(P)} \text{sign}(e_i), & lT \leq t < lT + \delta, \\ u_i(t) = 0, & \|e_i\| = 0 \text{ or } lT + \delta \leq t < (l+1)T, \end{cases} \quad (5.2)$$

and

$$\begin{cases} \dot{r}_i = -\lambda_i \left[\frac{\eta_i r_i e_i^T P e_i}{r_i^2} + e_i^T P e_i + \frac{\bar{k}}{\sqrt{\lambda_i}} \text{sign}(r_i) \right], & lT \leq t < lT + \delta, \\ \dot{r}_i = 0, & \|r_i\| = 0 \text{ or } lT + \delta \leq t < (l+1)T, \end{cases} \quad (5.3)$$

In the simulation, all parameters are selected as: $\bar{k} = 10$, $\lambda_i = 1$, $r_i(0) = 2 + 0.5i$, $x_i(0) = (3 + 0.2i, 4 + 0.3i, 6 + 0.4i)^T$, $y_i(0) = (-5 + 0.7i, -1 + 0.8i, -4 + 0.9i)^T$, where $1 \leq i \leq 100$. Further, select $T = 0.2$ and $\delta = 0.16$. Under above conditions, all assumptions of Theorem 1 are satisfied, then system (2.1) and system (3.1) are finite-time synchronized under the intermittent adaptive controllers (5.2) and the intermittent updated laws (5.3), which is shown in Fig 3.

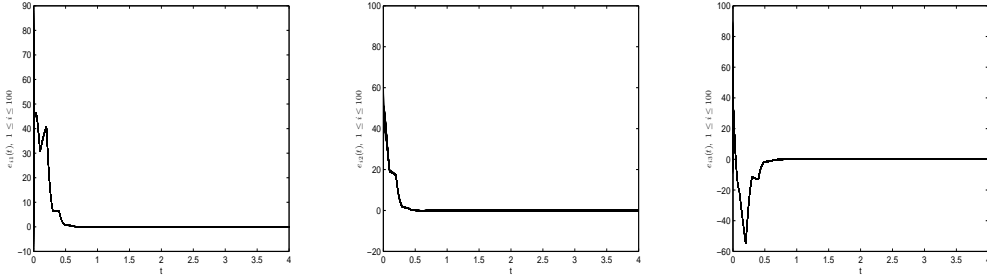


Fig.3. The errors e_{i1}, e_{i2}, e_{i3} , $i = 1, 2, \dots, 100$ under periodically intermittent adaptive controllers (5.2) and (5.3) with the parameters $T = 0.2$, $\delta = 0.16$.

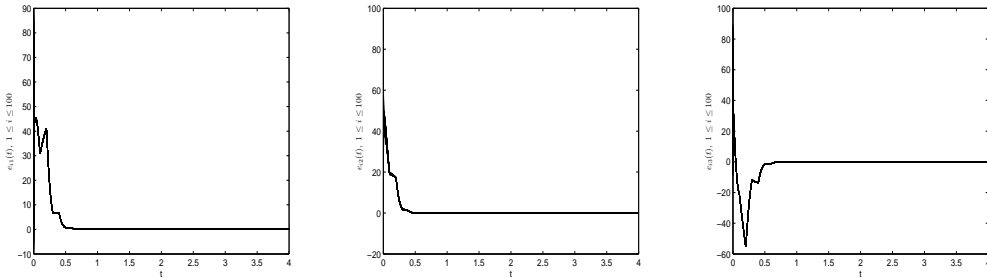


Fig.4. The errors e_{i1}, e_{i2}, e_{i3} , $i = 1, 2, \dots, 100$ under periodically intermittent adaptive controllers (5.2) and (5.3) with the parameters $T = 0.2$, $\delta = 0.08$.

5.2. The simulation with the periodically intermittent feedback controllers

Example 2. Consider the following example consisting of 100 identical nodes, which is described by:

$$\begin{cases} \dot{y}_{i1} = r_1(y_{i2} - y_{i1}) + r_3 \sum_{j=1}^{100} a_{ij} \Gamma y_{ij} + k_i e_{i1}, \\ \dot{y}_{i2} = r_3 y_{i1} - y_{i2} - y_{i1} y_{i3} + c \sum_{j=1}^{100} a_{ij} \Gamma y_{ij} + k_i e_{i2}, \\ \dot{y}_{i3} = -r_2 y_{i3} + y_{i1} y_{i2} + c \sum_{j=1}^{100} a_{ij} \Gamma y_{ij} + k_i e_{i3}, \end{cases} \quad (5.4)$$

$$\begin{cases} \dot{x}_{i1} = r_1(x_{i2} - x_{i1}) + c \sum_{j=1}^{100} a_{ij} \Gamma x_{ij} - r_i e_{i1}, \\ \dot{x}_{i2} = r_3 x_{i1} - x_{i2} - x_{i1} x_{i3} + c \sum_{j=1}^{100} a_{ij} \Gamma x_{ij} - r_i e_{i2}, \\ \dot{x}_{i3} = -r_2 x_{i3} + x_{i1} x_{i2} + c \sum_{j=1}^{100} a_{ij} \Gamma x_{ij} - r_i e_{i3}, \end{cases} \quad (5.5)$$

where $i = 1, 2, \dots, 100$, $A = (a_{ij})_{100 \times 100}$ is a symmetrically diffusive coupling matrix with $a_{ij} = 0$ or 1 ($j \neq i$), $\Gamma = \text{diag}\{1, 1, 1\}$ and the coupling strength $c = 1$.

According to Theorem 2, the intermittent updated laws of the coupling strengths can be described by (4.2)-(4.3).

For convenience, in the numerical simulation, all parameters are given as follows: $k_i(0) = -2 + 3i$, $r_i(0) = -1 + 2i$, $\psi_i = 1$, $\varphi_i = 1$, $x_i(0) = (4 + 0.2i, 2 + 0.3i, 5 + 0.4i)^T$, $y_i(0) = (-4 + 0.7i, -3 + 0.8i, -5 + 0.9i)^T$, $T = 0.2$, $\delta = 0.16$, where $1 \leq i \leq 10$. The discontinuous sign function is replaced by a continuous function (tanh). Then system (5.4) and (5.5) are finite-time synchronized under the periodically intermittent feedback controllers (4.2) and (4.3), which is shown in Fig. 5.

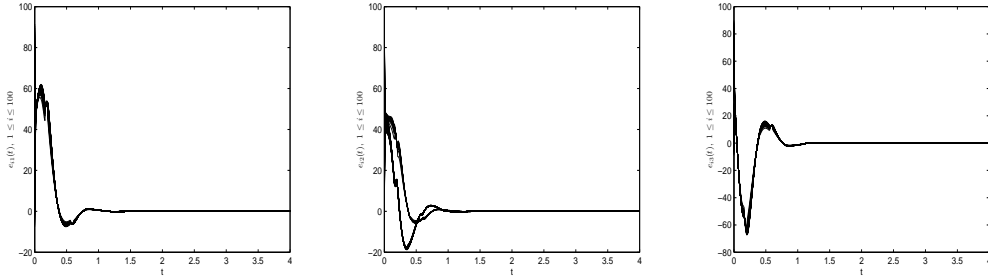


Fig.5. The synchronization errors e_{i1}, e_{i2}, e_{i3} , $i = 1, 2, \dots, 100$ of the complex networks (5.4) and (5.5) with $T = 0.2$, $\delta = 0.16$.

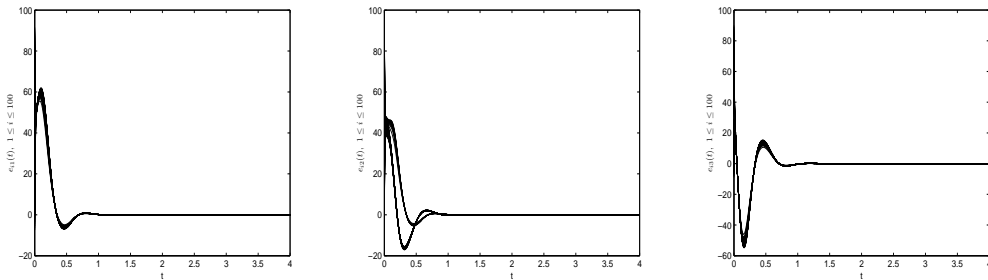


Fig.6. The synchronization errors e_{i1}, e_{i2}, e_{i3} , $i = 1, 2, \dots, 100$ of the complex networks (5.4) and (5.5) with full control (4.2) and (4.3).

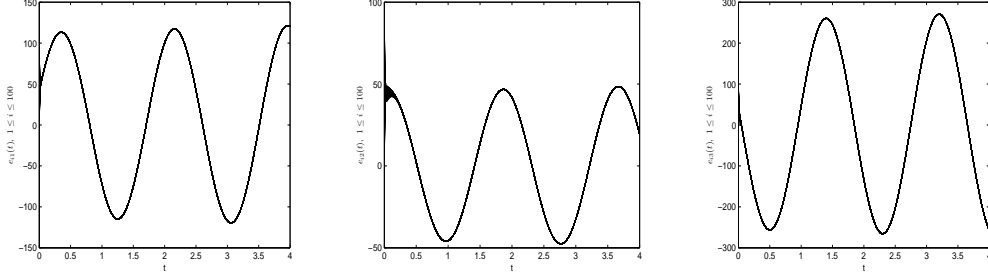


Fig.7. The synchronization errors e_{i1}, e_{i2}, e_{i3} , $i = 1, 2, \dots, 100$ of the complex networks (5.4) and (5.5) with no control inputs.

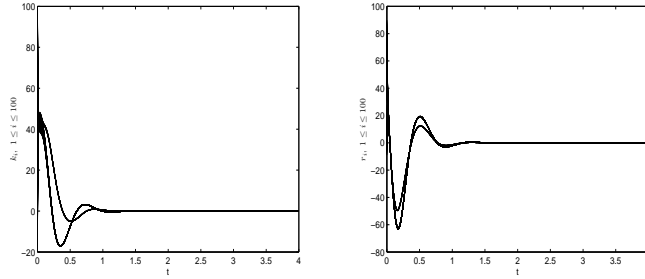


Fig.8. The feedback gains k_i, r_i , $i = 1, 2, \dots, 100$ of the complex networks (4.2) and (4.3) with $T = 0.2$, $\delta = 0.16$.

Fig.5 shows the synchronous errors e_{ij} ($1 \leq j \leq 3$) with intermittent control under $T = 0.2$, $\delta = 0.16$. Fig.6 shows the synchronous errors e_{ij} ($1 \leq j \leq 3$) with full control and the no control inputs are revealed in Fig.7, the feedback gains k_i, r_i are shown in Fig.8. It is clear that the synchronization of full control is faster than intermittent control.

Remark 8. Through simple computation by the above parameters, we get the setting time $(T_1, T_2, T_3) = (0.875, 0.786, 0.949)$, which came from Theorem 1, Corollaries 1 and Theorem 2, respectively. Hence, the datum shows the correction of the final sentence in Remark 7.

6 Conclusion

In this paper, the finite-time synchronization issue between two complex dynamical networks is discussed. Based on the Lyapunov stability, the periodically intermittent control schemes are proposed to synchronize such network in finite time by combining adaptive control method and feedback control technique, respectively. Some sufficient conditions ensuring the global finite-time stability of the synchronization process are derived. Numerical examples are presented to verify the effectiveness of the proposed synchronization criteria

finally.

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