

具有随机扰动和不确定性的 中立型耦合神经网络有限时间同步*

王柯杰, 陈巧玉, 童东兵, 毛琦

(上海工程技术大学 电子电气工程学院, 上海 201620)

摘要: 研究了具有时滞、不确定性和随机扰动的中立型耦合神经网络的有限时间同步问题。在 Lyapunov 稳定性理论的基础上, 结合不等式技术得到了有限时间同步判据。接着构造合适的状态反馈控制器, 使主从系统实现了有限时间同步。最后, 通过一个数值仿真验证了所提出理论的有效性。

关键词: 不确定性; 随机扰动; 耦合神经网络; 中立型; 有限时间同步

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Finite-Time Synchronization of Coupled Neutral-Type Neural Networks With Stochastic Disturbances and Uncertainties

WANG Kejie, CHEN Qiaoyu, TONG Dongbing, MAO Qi

(School of Electronic and Electrical Engineering, Shanghai University of Engineering Science, Shanghai 201620, P.R.China)

Abstract: The finite-time synchronization problems were solved for coupled neutral-type neural networks with stochastic disturbances and uncertainties. Based on the Lyapunov stability theory and the inequality techniques, the finite-time synchronization criterion was proposed for this system. Then the finite-time synchronization was realized for the master-slave system through the construction of an appropriate state feedback controller. At last, a numerical simulation was given to verify the effectiveness of the proposed theory.

Key words: uncertainty; stochastic disturbance; coupled neural network; neutral-type; finite-time synchronization

0 引 言

神经网络是模仿神经元信息传递过程所构建出的一种数学模型, 可以用来解决一些复杂的非线性问题。近年来, 神经网络已经被应用于生物信号检测、价格预测、风险评估和故障诊断等许多领域。随着研究的不断

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作者简介: 王柯杰(1996—), 男, 硕士生(E-mail: wangkejie0307@163.com);

陈巧玉(1984—), 女, 副教授, 博士(通讯作者, E-mail: goodluckqiaoyu@126.com);

童东兵(1979—), 男, 教授, 博士(E-mail: tongdongbing@163.com);

毛琦(1985—), 男, 讲师, 博士(E-mail: asdenglish@126.com)。

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深入,中立型神经网络吸引了越来越多的关注.它是神经网络的特殊形式,其时滞 in 系统现在的状态和状态的导数中同时存在,由此导致该类系统具有更加复杂的动态行为,也使得此类系统相较于一般时滞系统具有更好的普适性.像化学反应过程、涡轮喷气发动机的转动过程等都可以利用中立型时滞系统进行建模.通过设计新的 Lyapunov 泛函和提出新的分析技巧,文献[1]研究了一类具有混合时滞的中立型耦合神经网络的指数同步.文献[2]利用 M 矩阵方法和随机分析法,研究了具有 Markov 切换参数的中立型随机神经网络的自适应同步.基于 Lyapunov 稳定性理论和不等式技术,文献[3]针对一类具有多时滞的中立型神经网络,研究了其平衡点的存在性、唯一性和全局渐近稳定性.

同步是指复杂网络中节点达到相同状态的行为,它不仅是一种基本的自然现象,也是复杂网络科学领域的重要研究方向之一.在过去的几十年里,学者们进行了全面的研究,并取得了大量的成果.其中,有限时间同步由于其优异的暂态性能而受到越来越多的关注^[4].文献[5]利用 M 矩阵方法得到了具有 Markov 拓扑和分布脉冲效应的神经网络有限时间同步判据.针对一类模糊中立型耦合 Rayleigh 系统,文献[6]利用有限时间稳定性理论和不等式技术,给出了其有限时间同步判据.对于具有自适应状态耦合的多权重复杂网络,文献[7]通过 Lyapunov 稳定性理论,设计出合适的控制器,使其达到有限时间同步和 H_∞ 同步.

在神经网络中,不确定性和随机扰动往往会造成不稳定或抖振,这些因素不利于神经网络的实际应用.因此对不确定时滞系统的研究一直是控制理论研究中的难点和热点问题之一.本文研究了具有不确定性和随机扰动的中立型耦合神经网络有限时间同步问题,所构建的系统模型与文献[8-10]相比引入了不确定性和随机扰动,更具实际意义.在 Lyapunov 稳定性理论的基础上,结合不等式技术,推导出中立型耦合神经网络有限时间同步准则.为了解决不确定性和随机扰动造成的问题,本文构造了合适的状态反馈控制器来保证主从系统实现有限时间同步.以 Kronecker 积形式给出有限时间同步判据,易于使用 MATLAB 工具箱来检验.

1 问题描述

考虑一类具有时滞和不确定性及扰动的中立型神经网络,其驱动系统设计如下:

$$d[\mathbf{x}_i(t) - \mathbf{D}\mathbf{x}_i(t - \tau)] = [-(\mathbf{C} + \Delta\mathbf{C})\mathbf{x}_i(t) + (\mathbf{A} + \Delta\mathbf{A})\mathbf{f}(\mathbf{x}_i(t)) + (\mathbf{B} + \Delta\mathbf{B})\mathbf{f}(\mathbf{x}_i(t - \tau))] dt, \quad i = 1, 2, \dots, N, \quad (1)$$

其中, $\mathbf{x}_i(t) = [x_{i1}(t), \dots, x_{in}(t)]^T \in \mathbb{R}^n$ 代表第 i 个节点的状态向量, $\mathbf{D} = \text{diag}[d_1, \dots, d_n]$ 且 $|d_i| < 1$, $\mathbf{C} = \text{diag}[c_1, \dots, c_n]$ 是正定对角矩阵, $\mathbf{A} \in \mathbb{R}^{n \times n}$ 和 $\mathbf{B} \in \mathbb{R}^{n \times n}$ 分别为连接权矩阵和时滞连接权矩阵, $\mathbf{f}(\cdot)$ 是神经元激活函数且有界, τ 代表所考虑的时滞,未知矩阵 $\Delta\mathbf{C}, \Delta\mathbf{A}, \Delta\mathbf{B}$ 代表系统参数的不确定部分.

其响应系统为

$$d[\mathbf{y}_i(t) - \mathbf{D}\mathbf{y}_i(t - \tau)] = \left[-(\mathbf{C} + \Delta\mathbf{C})\mathbf{y}_i(t) + (\mathbf{A} + \Delta\mathbf{A})\mathbf{f}(\mathbf{y}_i(t)) + (\mathbf{B} + \Delta\mathbf{B})\mathbf{f}(\mathbf{y}_i(t - \tau)) + c \sum_{j=1}^N g_{ij} \mathbf{\Gamma} \mathbf{y}_j(t) + \mathbf{u}_i(t) \right] dt + \boldsymbol{\sigma}_i(t, \mathbf{y}_i(t), \mathbf{y}_i(t - \tau)) d\boldsymbol{\omega}_i, \quad i = 1, 2, \dots, N, \quad (2)$$

其中,系统参数 $\mathbf{D}, \mathbf{C}, \mathbf{A}, \mathbf{B}, \Delta\mathbf{C}, \Delta\mathbf{A}, \Delta\mathbf{B}$ 的定义与式(1)相同, $\mathbf{y}_i(t) = [y_{i1}(t), \dots, y_{in}(t)]^T \in \mathbb{R}^n$, $\mathbf{u}_i(t)$ 为响应网络第 i 个节点的控制输入, c 代表耦合强度, $\mathbf{\Gamma} = \text{diag}[\gamma_1, \dots, \gamma_n]$ 为内耦合配置矩阵, $\boldsymbol{\sigma}_i: \mathbb{R}_+ \times \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}^n$ 是噪声强度函数, $\boldsymbol{\omega}_i(t) = [\omega_{i1}(t), \dots, \omega_{in}(t)]^T$ 是定义在完备概率空间的标准 Brown 运动, $\mathbf{G} = (g_{ij})_{N \times N}$ 是外耦合转置矩阵,满足

$$g_{ii} = - \sum_{j=1, j \neq i}^N g_{ij}, \quad i = 1, 2, \dots, N.$$

将同步误差信号定义为 $\mathbf{e}_i(t) = \text{col}\{e_{i1}(t), e_{i2}(t), \dots, e_{in}(t)\} = \mathbf{y}_i(t) - \mathbf{x}_i(t)$. 根据矩阵 \mathbf{G} 的性质,可得到

$$d[\mathbf{e}_i(t) - \mathbf{D}\mathbf{e}_i(t - \tau)] = \left[-(\mathbf{C} + \Delta\mathbf{C})\mathbf{e}_i(t) + (\mathbf{A} + \Delta\mathbf{A})\hat{\mathbf{f}}(\mathbf{e}_i(t)) + (\mathbf{B} + \Delta\mathbf{B})\hat{\mathbf{f}}(\mathbf{e}_i(t - \tau)) + c \sum_{j=1}^N g_{ij} \mathbf{\Gamma} \mathbf{e}_j(t) + \mathbf{u}_i(t) \right] dt + \boldsymbol{\sigma}_i(t, \mathbf{e}_i(t), \mathbf{e}_i(t - \tau)) d\boldsymbol{\omega}_i, \quad i = 1, 2, \dots, N, \quad (3)$$

其中

假设 3 未知矩阵 $\Delta C, \Delta A, \Delta B$ 满足下列结构:

$$(\Delta C, \Delta A, \Delta B) = PS(t)(H_1, H_2, H_3), \quad (6)$$

其中 P, H_1, H_2 和 H_3 是已知的实矩阵, 不确定矩阵 $S(t)$ 可以是时变不固定的, 且满足

$$S^T(t)S(t) \leq I. \quad (7)$$

注 2 式(6)和(7)中的参数不确定性结构广泛应用于不确定系统的随机过程. 式(7)可以对具有参数不确定性的实际系统进行精确建模. 值得注意的是, 式(6)中的可变矩阵 $S(t)$ 甚至可以是状态相关的, 即只要满足式(7), 则 $S(t) = S(t, \varphi(t))$ 成立.

引理 1^[13] 存在 $x, y \in \mathbb{R}^n$, 对任意 $\varepsilon > 0$, 不等式 $x^T y + y^T x \leq \varepsilon x^T x + \varepsilon^{-1} y^T y$ 成立.

引理 2^[14] 假设存在 $x_i \in \mathbb{R}^n, i = 1, \dots, n, 0 < \xi \leq 1$, 则下列不等式成立:

$$\left(\sum_{i=1}^n |x_i| \right)^\xi \leq \sum_{i=1}^n |x_i|^\xi.$$

引理 3^[15] 存在非负连续函数 $\chi(t) \in \mathbb{R}^n$, 常数 $\zeta \in (0, \infty)$ 和 $\xi \in (0, 1)$ 满足

$$\mathcal{L}\chi(t) \leq -\zeta\chi^\xi(t), \quad t \in \mathbb{R}^n \setminus \{0\},$$

则 $\chi(t)$ 的平凡零解在概率空间内是有限时间稳定的, 且随机设定时间满足 $E\{T_0\} \leq \chi^{1-\xi}(0)/(\zeta(1-\xi))$.

定义 1^[16] 假设存在任意常数 $T_0 \in (0, \infty)$ 满足

$$\lim_{t \rightarrow t_0 + T_0} \|e(t)\| = 0, \quad \|e(t)\| = 0,$$

则误差系统(4)可以实现有限时间同步, 其中 $t \geq t_0 + T_0$, 且 T_0 为同步的设定时间.

2 主要定理及证明

在这一节中, 我们将推导出主从系统实现有限时间同步的判据.

定理 1 在假设 1、2、3 满足的前提下, 对于任意 $\varepsilon_i > 0 (i = 1, 2, \dots, 10)$, 对称矩阵 R_1, R_2 满足

$$\begin{bmatrix} \Omega_1 & \Omega_3^T \\ \Omega_3 & \Omega_2 \end{bmatrix} \leq 0, \quad (8)$$

其中 $\Omega_1 = I_N \otimes (-2C + (\varepsilon_1 + \varepsilon_3 + \varepsilon_5)PP^T + (\varepsilon_1^{-1} + \varepsilon_6^{-1})H_1^T H_1 + \varepsilon_2 AA^T + (\varepsilon_2^{-1} + \varepsilon_3^{-1} \lambda_{\max}(I_N \otimes (H_2^T H_2))) + \varepsilon_7^{-1} + \varepsilon_8^{-1} \lambda_{\max}(I_N \otimes (H_2^T H_2)))U^T U + \varepsilon_4 BB^T) + 2cG \otimes \Gamma + M^T M - 2R_1, \Omega_2 = I_N \otimes ((\varepsilon_4^{-1} + \varepsilon_5^{-1} \lambda_{\max}(I_N \otimes (H_3^T H_3))) + \varepsilon_9^{-1} + \varepsilon_{10}^{-1} \lambda_{\max}(I_N \otimes (H_3^T H_3)))U^T U + (\varepsilon_6 + \varepsilon_8 + \varepsilon_{10})DPP^T D + \varepsilon_7 DAA^T D + \varepsilon_9 DBB^T D) + N^T N - 2(I_N \otimes D)R_2, \Omega_3 = I_N \otimes (DC) - cG \otimes (D\Gamma) + (I_N \otimes D)R_1 + R_2$, 则反馈控制器(5)可以使误差系统(4)达到有限时间同步.

证明 定义一个算子 $\mathcal{D}e(t) = e(t) - (I_N \otimes D)e(t - \tau)$, 并构造如下 Lyapunov 函数:

$$V(t) = (\mathcal{D}e(t))^T (\mathcal{D}e(t)). \quad (9)$$

根据 Itô 公式, 可以得到

$$dV(t) = \mathcal{L}V(t) + 2(\mathcal{D}e(t))^T \sigma(t, e(t), e(t - \tau)) d\omega(t), \quad (10)$$

其中

$$\begin{aligned} \mathcal{L}V(t) = & 2[(\mathcal{D}e(t))^T [- (I_N \otimes C)e(t) - (I_N \otimes \Delta C)e(t) + (I_N \otimes A)F(e(t)) + \\ & (I_N \otimes \Delta A)F(e(t)) + (I_N \otimes B)F(e(t - \tau)) + (I_N \otimes \Delta B)F(e(t - \tau)) + \\ & (cG \otimes \Gamma)e(t) + u(t)]] + \text{tr}(\sigma^T(t, e(t), e(t - \tau))\sigma(t, e(t), e(t - \tau))). \end{aligned} \quad (11)$$

根据假设 1 和假设 3, 将控制器(5)代入式(11), 能够获得

$$\begin{aligned} \mathcal{L}V(t) \leq & -2e^T(t)(I_N \otimes C)e(t) - 2e^T(t)(I_N \otimes \Delta C)e(t) + 2e^T(t)(I_N \otimes A)F(e(t)) + \\ & 2e^T(t)(I_N \otimes \Delta A)F(e(t)) + 2e^T(t)(I_N \otimes B)F(e(t - \tau)) + \\ & 2e^T(t)(I_N \otimes \Delta B)F(e(t - \tau)) + 2e^T(t)(cG \otimes \Gamma)e(t) - 2e^T(t)R_1 e(t) + \\ & 2e^T(t)R_2 e(t - \tau) - 2\eta e^T(t) \text{sign}(e(t) - (I_N \otimes D)e(t - \tau)) \times \\ & |e(t) - (I_N \otimes D)e(t - \tau)|^\delta + 2e^T(t - \tau)(I_N \otimes (DC))e(t) + \end{aligned}$$

$$\begin{aligned}
& 2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{D}\Delta\mathbf{C}))\mathbf{e}(t) - 2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{D}\mathbf{A}))\mathbf{F}(\mathbf{e}(t)) - \\
& 2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{D}\Delta\mathbf{A}))\mathbf{F}(\mathbf{e}(t)) - 2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{D}\mathbf{B}))\mathbf{F}(\mathbf{e}(t-\tau)) - \\
& 2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{D}\Delta\mathbf{B}))\mathbf{F}(\mathbf{e}(t-\tau)) - 2\mathbf{e}^T(t-\tau)(\mathbf{c}\mathbf{G} \otimes (\mathbf{D}\mathbf{I}^T))\mathbf{e}(t) + \\
& 2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes \mathbf{D})\mathbf{R}_1\mathbf{e}(t) - 2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes \mathbf{D})\mathbf{R}_2\mathbf{e}(t-\tau) + \\
& 2\eta\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes \mathbf{D})\text{sign}(\mathbf{e}(t) - (\mathbf{I}_N \otimes \mathbf{D})\mathbf{e}(t-\tau)) \mid \mathbf{e}(t) - (\mathbf{I}_N \otimes \mathbf{D})\mathbf{e}(t-\tau) \mid^\delta + \\
& \mathbf{e}^T(t)\mathbf{M}^T\mathbf{M}\mathbf{e}(t) + \mathbf{e}^T(t-\tau)\mathbf{N}^T\mathbf{N}\mathbf{e}(t-\tau). \tag{12}
\end{aligned}$$

根据假设 2, 可以得到

$$\mid \hat{\mathbf{f}}(e_{ij}(t)) \mid = \mid \mathbf{f}(e_{ij}(t) + x_{ij}(t)) - \mathbf{f}(x_{ij}(t)) \mid \leq l_j \mid e_{ij}(t) \mid = l_j e_{ij}(t), \tag{13}$$

且

$$\|\hat{\mathbf{f}}(\mathbf{e}_i(t))\| = \|\mathbf{f}(\mathbf{e}_i(t) + \mathbf{x}_i(t)) - \mathbf{f}(\mathbf{x}_i(t))\| \leq \|\mathbf{U}\mathbf{e}_i(t)\|, \tag{14}$$

其中

$$l_j = \max\{\mid l_j^-, \mid l_j^+ \mid\}, j = 1, \dots, n, \mathbf{U} = \text{diag}(l_1, \dots, l_n).$$

由式 (14), 能够获得

$$\begin{aligned}
\mathbf{F}^T(\mathbf{e}(t))\mathbf{F}(\mathbf{e}(t)) &= \sum_{i=1}^N \|\hat{\mathbf{f}}(\mathbf{e}_i(t))\|^2 \leq \sum_{i=1}^N \mathbf{e}_i^T(t)\mathbf{U}^T\mathbf{U}\mathbf{e}_i(t) = \\
&\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{U}^T\mathbf{U}))\mathbf{e}(t). \tag{15}
\end{aligned}$$

通过假设 3 和引理 1, 有

$$\begin{aligned}
-2\mathbf{e}^T(t)(\mathbf{I}_N \otimes \Delta\mathbf{C})\mathbf{e}(t) &= \\
-2\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{P}\mathbf{S}(t)\mathbf{H}_1))\mathbf{e}(t) &= \\
-2\mathbf{e}^T(t)(\mathbf{I}_N \otimes \mathbf{P})(\mathbf{I}_N \otimes (\mathbf{S}(t)\mathbf{H}_1))\mathbf{e}(t) &\leq \\
\varepsilon_1\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{P}\mathbf{P}^T))\mathbf{e}(t) + \varepsilon_1^{-1}\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{H}_1^T\mathbf{H}_1))\mathbf{e}(t), \tag{16}
\end{aligned}$$

$$\begin{aligned}
2\mathbf{e}^T(t)(\mathbf{I}_N \otimes \mathbf{A})\mathbf{F}(\mathbf{e}(t)) &\leq \\
\varepsilon_2\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{A}\mathbf{A}^T))\mathbf{e}(t) + \varepsilon_2^{-1}\mathbf{F}^T(\mathbf{e}(t))\mathbf{F}(\mathbf{e}(t)) &\leq \\
\varepsilon_2\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{A}\mathbf{A}^T))\mathbf{e}(t) + \varepsilon_2^{-1}\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{U}^T\mathbf{U}))\mathbf{e}(t), \tag{17}
\end{aligned}$$

$$\begin{aligned}
2\mathbf{e}^T(t)(\mathbf{I}_N \otimes \Delta\mathbf{A})\mathbf{F}(\mathbf{e}(t)) &= \\
2\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{P}\mathbf{S}(t)\mathbf{H}_2))\mathbf{F}(\mathbf{e}(t)) &= \\
2\mathbf{e}^T(t)(\mathbf{I}_N \otimes \mathbf{P})(\mathbf{I}_N \otimes (\mathbf{S}(t)\mathbf{H}_2))\mathbf{F}(\mathbf{e}(t)) &\leq \\
\varepsilon_3\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{P}\mathbf{P}^T))\mathbf{e}(t) + \varepsilon_3^{-1}\mathbf{F}^T(\mathbf{e}(t))(\mathbf{I}_N \otimes (\mathbf{H}_2^T\mathbf{H}_2))\mathbf{F}(\mathbf{e}(t)) &\leq \\
\varepsilon_3\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{P}\mathbf{P}^T))\mathbf{e}(t) + \varepsilon_3^{-1}\lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_2^T\mathbf{H}_2))\mathbf{F}^T(\mathbf{e}(t))\mathbf{F}(\mathbf{e}(t)) &\leq \\
\varepsilon_3\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{P}\mathbf{P}^T))\mathbf{e}(t) + \varepsilon_3^{-1}\lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_2^T\mathbf{H}_2))\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{U}^T\mathbf{U}))\mathbf{e}(t), \tag{18}
\end{aligned}$$

$$\begin{aligned}
2\mathbf{e}^T(t)(\mathbf{I}_N \otimes \mathbf{B})\mathbf{F}(\mathbf{e}(t-\tau)) &\leq \\
\varepsilon_4\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{B}\mathbf{B}^T))\mathbf{e}(t) + \varepsilon_4^{-1}\mathbf{F}^T(\mathbf{e}(t-\tau))\mathbf{F}(\mathbf{e}(t-\tau)) &\leq \\
\varepsilon_4\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{B}\mathbf{B}^T))\mathbf{e}(t) + \varepsilon_4^{-1}\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{U}^T\mathbf{U}))\mathbf{e}(t-\tau), \tag{19}
\end{aligned}$$

$$\begin{aligned}
2\mathbf{e}^T(t)(\mathbf{I}_N \otimes \Delta\mathbf{B})\mathbf{F}(\mathbf{e}(t-\tau)) &= \\
2\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{P}\mathbf{S}(t)\mathbf{H}_3))\mathbf{F}(\mathbf{e}(t-\tau)) &= \\
2\mathbf{e}^T(t)(\mathbf{I}_N \otimes \mathbf{P})(\mathbf{I}_N \otimes (\mathbf{S}(t)\mathbf{H}_3))\mathbf{F}(\mathbf{e}(t-\tau)) &\leq \\
\varepsilon_5\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{P}\mathbf{P}^T))\mathbf{e}(t) + \varepsilon_5^{-1}\mathbf{F}^T(\mathbf{e}(t-\tau))(\mathbf{I}_N \otimes (\mathbf{H}_3^T\mathbf{H}_3))\mathbf{F}(\mathbf{e}(t-\tau)) &\leq \\
\varepsilon_5\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{P}\mathbf{P}^T))\mathbf{e}(t) + \varepsilon_5^{-1}\lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_3^T\mathbf{H}_3))\mathbf{F}^T(\mathbf{e}(t-\tau))\mathbf{F}(\mathbf{e}(t-\tau)) &\leq \\
\varepsilon_5\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{P}\mathbf{P}^T))\mathbf{e}(t) + \\
\varepsilon_5^{-1}\lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_3^T\mathbf{H}_3))\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{U}^T\mathbf{U}))\mathbf{e}(t-\tau), \tag{20}
\end{aligned}$$

$$2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{D}\Delta\mathbf{C}))\mathbf{e}(t) =$$

$$\begin{aligned}
& 2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DPS}(t)\mathbf{H}_1))\mathbf{e}(t) = \\
& 2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DP}))(\mathbf{I}_N \otimes (\mathbf{S}(t)\mathbf{H}_1))\mathbf{e}(t) \leq \\
& \varepsilon_6 \mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DPP}^T\mathbf{D}))\mathbf{e}(t-\tau) + \varepsilon_6^{-1} \mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{H}_1^T\mathbf{H}_1))\mathbf{e}(t), \tag{21}
\end{aligned}$$

$$\begin{aligned}
& -2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DA}))\mathbf{F}(\mathbf{e}(t)) \leq \\
& \varepsilon_7 \mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DAA}^T\mathbf{D}))\mathbf{e}(t-\tau) + \varepsilon_7^{-1} \mathbf{F}^T(\mathbf{e}(t))\mathbf{F}(\mathbf{e}(t)) \leq \\
& \varepsilon_7 \mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DAA}^T\mathbf{D}))\mathbf{e}(t-\tau) + \varepsilon_7^{-1} \mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{U}^T\mathbf{U}))\mathbf{e}(t), \tag{22}
\end{aligned}$$

$$\begin{aligned}
& -2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{D}\Delta\mathbf{A}))\mathbf{F}(\mathbf{e}(t)) = \\
& -2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DPS}(t)\mathbf{H}_2))\mathbf{F}(\mathbf{e}(t)) = \\
& -2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DP}))(\mathbf{I}_N \otimes \mathbf{S}(t)\mathbf{H}_2)\mathbf{F}(\mathbf{e}(t)) \leq \\
& \varepsilon_8 \mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DPP}^T\mathbf{D}))\mathbf{e}(t-\tau) + \varepsilon_8^{-1} \mathbf{F}^T(\mathbf{e}(t))(\mathbf{I}_N \otimes (\mathbf{H}_2^T\mathbf{H}_2))\mathbf{F}(\mathbf{e}(t)) \leq \\
& \varepsilon_8 \mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DPP}^T\mathbf{D}))\mathbf{e}(t-\tau) + \varepsilon_8^{-1} \lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_2^T\mathbf{H}_2))\mathbf{F}^T(\mathbf{e}(t))\mathbf{F}(\mathbf{e}(t)) \leq \\
& \varepsilon_8 \mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DPP}^T\mathbf{D}))\mathbf{e}(t-\tau) + \\
& \varepsilon_8^{-1} \lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_2^T\mathbf{H}_2))\mathbf{e}^T(t)(\mathbf{I}_N \otimes (\mathbf{U}^T\mathbf{U}))\mathbf{e}(t), \tag{23}
\end{aligned}$$

$$\begin{aligned}
& -2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DB}))\mathbf{F}(\mathbf{e}(t-\tau)) \leq \\
& \varepsilon_9 \mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DBB}^T\mathbf{D}))\mathbf{e}(t-\tau) + \varepsilon_9^{-1} \mathbf{F}^T(\mathbf{e}(t-\tau))\mathbf{F}(\mathbf{e}(t-\tau)) \leq \\
& \varepsilon_9 \mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DBB}^T\mathbf{D}))\mathbf{e}(t-\tau) + \varepsilon_9^{-1} \mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{U}^T\mathbf{U}))\mathbf{e}(t-\tau), \tag{24}
\end{aligned}$$

$$\begin{aligned}
& -2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{D}\Delta\mathbf{B}))\mathbf{F}(\mathbf{e}(t-\tau)) = \\
& -2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DPS}(t)\mathbf{H}_3))\mathbf{F}(\mathbf{e}(t-\tau)) = \\
& -2\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DP}))(\mathbf{I}_N \otimes (\mathbf{S}(t)\mathbf{H}_3))\mathbf{F}(\mathbf{e}(t-\tau)) \leq \\
& \varepsilon_{10} \mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DPP}^T\mathbf{D}))\mathbf{e}(t-\tau) + \varepsilon_{10}^{-1} \mathbf{F}^T(\mathbf{e}(t-\tau))(\mathbf{I}_N \otimes (\mathbf{H}_3^T\mathbf{H}_3))\mathbf{F}(\mathbf{e}(t-\tau)) \leq \\
& \varepsilon_{10} \mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DPP}^T\mathbf{D}))\mathbf{e}(t-\tau) + \varepsilon_{10}^{-1} \lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_3^T\mathbf{H}_3))\mathbf{F}^T(\mathbf{e}(t-\tau))\mathbf{F}(\mathbf{e}(t-\tau)) \leq \\
& \varepsilon_{10} \mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{DPP}^T\mathbf{D}))\mathbf{e}(t-\tau) + \\
& \varepsilon_{10}^{-1} \lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_3^T\mathbf{H}_3))\mathbf{e}^T(t-\tau)(\mathbf{I}_N \otimes (\mathbf{U}^T\mathbf{U}))\mathbf{e}(t-\tau). \tag{25}
\end{aligned}$$

将式(16)–(25)代入式(12), 可得

$$\begin{aligned}
\mathcal{L}V(t) & \leq \mathbf{e}^T(t)[\mathbf{I}_N \otimes (-2\mathbf{C} + (\varepsilon_1 + \varepsilon_3 + \varepsilon_5)\mathbf{PP}^T + (\varepsilon_1^{-1} + \varepsilon_6^{-1})\mathbf{H}_1^T\mathbf{H}_1 + \varepsilon_2\mathbf{AA}^T + \\
& (\varepsilon_2^{-1} + \varepsilon_3^{-1} \lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_2^T\mathbf{H}_2)) + \varepsilon_7^{-1} + \varepsilon_8^{-1} \lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_2^T\mathbf{H}_2)))\mathbf{U}^T\mathbf{U} + \varepsilon_4\mathbf{BB}^T) + \\
& 2c\mathbf{G} \otimes \mathbf{\Gamma} + \mathbf{M}^T\mathbf{M} - 2\mathbf{R}_1]\mathbf{e}(t) + 2\mathbf{e}^T(t-\tau)[\mathbf{I}_N \otimes (\mathbf{DC}) - \\
& c\mathbf{G} \otimes (\mathbf{D}\mathbf{\Gamma}) + (\mathbf{I}_N \otimes \mathbf{D})\mathbf{R}_1 + \mathbf{R}_2]\mathbf{e}(t) + \mathbf{e}^T(t-\tau)[\mathbf{I}_N \otimes ((\varepsilon_4^{-1} + \\
& \varepsilon_5^{-1} \lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_3^T\mathbf{H}_3)) + \varepsilon_9^{-1} + \varepsilon_{10}^{-1} \lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_3^T\mathbf{H}_3)))\mathbf{U}^T\mathbf{U} + \\
& (\varepsilon_6 + \varepsilon_8 + \varepsilon_{10})\mathbf{DPP}^T\mathbf{D} + \varepsilon_7\mathbf{DAA}^T\mathbf{D} + \varepsilon_9\mathbf{DBB}^T\mathbf{D}) + \mathbf{N}^T\mathbf{N} - \\
& 2(\mathbf{I}_N \otimes \mathbf{D})\mathbf{R}_2]\mathbf{e}(t-\tau) - 2\eta \|\mathbf{e}(t) - (\mathbf{I}_N \otimes \mathbf{D})\mathbf{e}(t-\tau)\|^{\delta+1}. \tag{26}
\end{aligned}$$

根据引理 2, 有

$$\begin{aligned}
& -2\eta \|\mathbf{e}(t) - (\mathbf{I}_N \otimes \mathbf{D})\mathbf{e}(t-\tau)\|^{\delta+1} = \\
& -2\eta \sum_{i=1}^N \sum_{j=1}^N |e_{ij}(t) - d_j e_{ij}(t-\tau)|^{\delta+1} = \\
& -2\eta \sum_{i=1}^N \sum_{j=1}^N |(e_{ij}(t) - d_j e_{ij}(t-\tau))^2|^{\delta+1/2} \leq \\
& -2\eta \left(\sum_{i=1}^N \sum_{j=1}^N (e_{ij}(t) - d_j e_{ij}(t-\tau)) \right)^{\delta+1/2} = \\
& -2\eta ((\underline{\mathbf{D}}\mathbf{e}(t)))^T (\underline{\mathbf{D}}\mathbf{e}(t))^{\delta+1/2} = \\
& -2\eta V^{\delta+1/2}(t). \tag{27}
\end{aligned}$$

接着对式(10)和(26)求数学期望, 可得

$$E[V(t)] \leq E \left[\begin{bmatrix} \mathbf{e}^T(t) & \mathbf{e}^T(t-\tau) \end{bmatrix} \begin{bmatrix} \mathbf{\Omega}_1 & \mathbf{\Omega}_3^T \\ \mathbf{\Omega}_3 & \mathbf{\Omega}_2 \end{bmatrix} \begin{bmatrix} \mathbf{e}(t) \\ \mathbf{e}(t-\tau) \end{bmatrix} \right] - 2\eta E[V(t)]^{(\delta+1)/2}, \quad (28)$$

其中

$$\begin{aligned} \mathbf{\Omega}_1 = & \mathbf{I}_N \otimes (-2\mathbf{C} + (\varepsilon_1 + \varepsilon_3 + \varepsilon_5)\mathbf{P}\mathbf{P}^T + (\varepsilon_1^{-1} + \varepsilon_6^{-1})\mathbf{H}_1^T\mathbf{H}_1 + \varepsilon_2\mathbf{A}\mathbf{A}^T + \\ & (\varepsilon_2^{-1} + \varepsilon_3^{-1}\lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_2^T\mathbf{H}_2)) + \varepsilon_7^{-1} + \varepsilon_8^{-1}\lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_2^T\mathbf{H}_2)))\mathbf{U}^T\mathbf{U} + \varepsilon_4\mathbf{B}\mathbf{B}^T) + \\ & 2c\mathbf{G} \otimes \mathbf{\Gamma} + \mathbf{M}^T\mathbf{M} - 2\mathbf{R}_1, \end{aligned}$$

$$\begin{aligned} \mathbf{\Omega}_2 = & \mathbf{I}_N \otimes ((\varepsilon_4^{-1} + \varepsilon_5^{-1}\lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_3^T\mathbf{H}_3)) + \varepsilon_9^{-1} + \varepsilon_{10}^{-1}\lambda_{\max}(\mathbf{I}_N \otimes (\mathbf{H}_3^T\mathbf{H}_3)))\mathbf{U}^T\mathbf{U} + \\ & (\varepsilon_6 + \varepsilon_8 + \varepsilon_{10})\mathbf{D}\mathbf{P}\mathbf{P}^T\mathbf{D} + \varepsilon_7\mathbf{D}\mathbf{A}\mathbf{A}^T\mathbf{D} + \varepsilon_9\mathbf{D}\mathbf{B}\mathbf{B}^T\mathbf{D}) + \mathbf{N}^T\mathbf{N} - 2(\mathbf{I}_N \otimes \mathbf{D})\mathbf{R}_2, \end{aligned}$$

$$\mathbf{\Omega}_3 = \mathbf{I}_N \otimes (\mathbf{D}\mathbf{C}) - c\mathbf{G} \otimes (\mathbf{D}\mathbf{\Gamma}) + (\mathbf{I}_N \otimes \mathbf{D})\mathbf{R}_1 + \mathbf{R}_2.$$

当定理1成立时, $E[dV(t)] \leq -2\eta E[V(t)]^{(\delta+1)/2}$. 通过引理3, 可以保证误差系统(4)可以在设定时间 $T_0 = V^{(1-\delta)/2}(0)/(\eta(1-\delta))$ 实现有限时间同步, 证明完成.

控制器算法的步骤如下:

- 步1 定义激活函数 $f(\cdot)$ 和 Gauss 函数的宽度 σ_i , 并选择矩阵 $\mathbf{D}, \mathbf{C}, \mathbf{B}, \mathbf{A}, \mathbf{\Gamma}, \mathbf{G}$;
- 步2 选择不确定参数矩阵 $\mathbf{P}, \mathbf{H}_1, \mathbf{H}_2, \mathbf{H}_3$ 及噪声强度矩阵 \mathbf{M}, \mathbf{N} ;
- 步3 通过 $\mathbf{e}(t) = \mathbf{y}(t) - \mathbf{x}(t)$ 计算 $\mathbf{e}(t)$;
- 步4 选择合适的参数 $\eta > 0, 0 < \delta < 1$, 且由式(8)可以确定对称矩阵 $\mathbf{R}_1, \mathbf{R}_2$;
- 步5 计算 $\mathbf{u}(t)$ 的值, 并将其用于产生控制信号.

3 仿 真

在本节中, 我们将通过以下的仿真来检验所设计的反馈控制器的有效性. 误差系统(4)的参数设计为

$$\mathbf{D} = \begin{bmatrix} -0.5 & 0 \\ 0 & -0.3 \end{bmatrix}, \mathbf{C} = \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix}, \mathbf{A} = \begin{bmatrix} 0.2 & -0.03 \\ -0.3 & 0.04 \end{bmatrix}, \mathbf{B} = \begin{bmatrix} 0.02 & -0.08 \\ -0.03 & 0.02 \end{bmatrix},$$

$$\mathbf{P} = \begin{bmatrix} 0.3 & 0 \\ 0 & 0.6 \end{bmatrix}, \mathbf{H}_1 = \begin{bmatrix} 0.01 & 0 \\ 0 & 0.03 \end{bmatrix}, \mathbf{H}_2 = \begin{bmatrix} 0.05 & 0 \\ 0 & 0.02 \end{bmatrix}, \mathbf{H}_3 = \begin{bmatrix} 0.02 & 0 \\ 0 & 0.01 \end{bmatrix},$$

$$\mathbf{U} = \begin{bmatrix} 0.02 & 0 \\ 0 & 0.01 \end{bmatrix}, \mathbf{G} = \begin{bmatrix} -1 & 1 \\ 2 & -2 \end{bmatrix}, \mathbf{\Gamma} = \begin{bmatrix} 0.03 & 0 \\ 0 & 0.05 \end{bmatrix},$$

$$\mathbf{M} = \begin{bmatrix} -0.01 & -0.01 & 0.03 & 0.04 \\ -0.01 & 0.01 & -0.03 & -0.04 \\ 0.03 & -0.03 & 0.04 & -0.01 \\ 0.04 & -0.04 & -0.01 & 0.06 \end{bmatrix}, \mathbf{N} = \begin{bmatrix} 0.01 & -0.02 & -0.01 & -0.01 \\ -0.02 & 0.02 & -0.01 & -0.03 \\ -0.01 & -0.01 & -0.02 & -0.01 \\ -0.01 & -0.03 & -0.01 & 0.02 \end{bmatrix},$$

激活函数为 $f(\cdot) = \tanh(\cdot)$, $\eta = 1.13, \delta = 0.5, \mathbf{x}_1(0) = [1.2, 1.8]^T, \mathbf{x}_2(0) = [1.5, 0.8]^T, \mathbf{y}_1(0) = [1.7, -2.3]^T, \mathbf{y}_2(0) = [1.3, 0.4]^T, \tau = 0.1$. 通过使用 MATLAB 的 Yalmip 工具箱, 求得

$$\mathbf{R}_1 = \begin{bmatrix} 0.018 & 0 & 0.050 & 0 & 0.137 & 0 & 0.012 & 6 \\ 0.050 & 0 & 0.062 & 0 & 0.013 & 0 & 0.186 & 0 \\ 0.137 & 0 & 0.013 & 0 & 0.037 & 0 & 0.135 & 0 \\ 0.012 & 6 & 0.186 & 0 & 0.135 & 0 & 0.023 & 0 \end{bmatrix},$$

$$\mathbf{R}_2 = \begin{bmatrix} -0.170 & 0 & -0.150 & 0 & -0.001 & 0 & -0.001 & 0 \\ -0.150 & 0 & -1.800 & 0 & -0.001 & 5 & 0.000 & 37 \\ -0.001 & 0 & -0.001 & 5 & -0.170 & 0 & -0.122 & 8 \\ -0.001 & 0 & 0.000 & 37 & -0.122 & 8 & -1.800 & 0 \end{bmatrix},$$

$$\varepsilon_1 = 0.316, \varepsilon_2 = 0.512, \varepsilon_3 = 0.235, \varepsilon_4 = 0.613, \varepsilon_5 = 2.130,$$

$$\varepsilon_6 = 0.718, \varepsilon_7 = 0.223, \varepsilon_8 = 0.313, \varepsilon_9 = 0.524, \varepsilon_{10} = 0.413.$$

经过检验, 以上结果满足假设1、2、3和定理1中的条件. 因此误差系统(4)可以实现有限时间同步. 通过

仿真,可以得到以下仿真结果.图 2 为随机噪声,图 3 和图 4 分别描绘了无控制输入和控制器(5)作用下误差系统(4)的状态曲线,且同步时间 $t \approx 3.3 \text{ s} < T_0 = 4.12 \text{ s}$,图 5 是控制器(5)的信号轨迹.由此可以证明本文所设计的控制器的有效性.

注 3 与文献[17]的同步时间 $t \approx 5 \text{ s}$ 和文献[18]的同步时间 $t \approx 10 \text{ s}$ 相比,本文所设计的控制器可以使系统在 $t \approx 3.3 \text{ s}$ 内达到同步,同步时间更短,更具优越性.

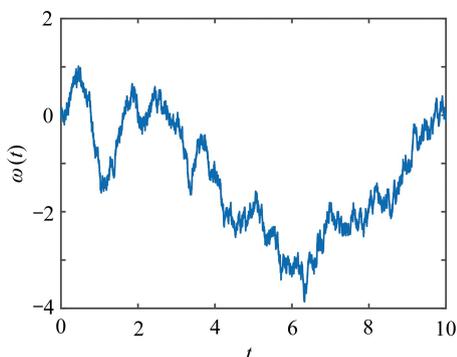


图 2 随机噪声

Fig. 2 Random noises

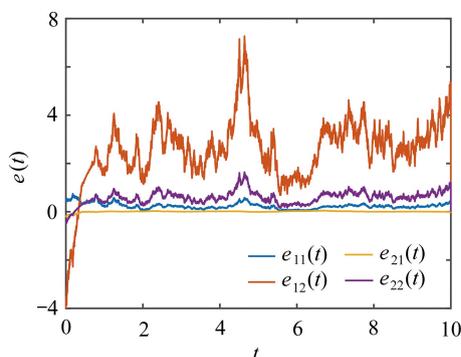


图 3 无控制器作用下的误差系统状态轨迹

Fig. 3 State trajectories of the error system

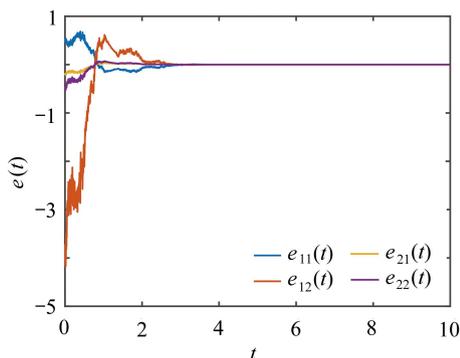


图 4 控制器(5)作用下的误差系统状态轨迹

Fig. 4 State trajectories of the error system with controller (5)

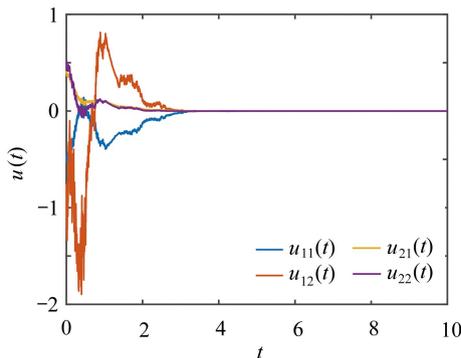


图 5 控制输入

Fig. 5 Control inputs

注 为了解释图中的颜色,读者可以参考本文的电子网页版本.

4 结 论

本文进一步研究了中立型耦合神经网络的有限时间同步问题,所使用的模型同时考虑了时滞、不确定性和随机扰动的影响,更具普遍性.通过构造合适的 Lyapunov 函数和运用不等式技术,推导出中立型耦合神经网络的有限时间同步准则.设计适当的状态反馈控制器,使所考虑的系统达到有限时间同步状态.最后通过仿真结果检验了所获结论的有效性.

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