

Exponential stability for stochastic interval delayed Hopfield neural networks

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Abstract: A type of stochastic interval delayed Hopfield neural network had been studied. By using Itô formula and Lyapunov function, some new delay-dependent and delay-independent sufficient conditions of its global exponential stability had been given. All the results obtained were generalizations of some recent results reported in the literature for stochastic neural networks with constant delays or their certain cases with variable-delays.

Key words: stochastic interval delayed Hopfield neural network; Brownian motion; Itô formula; Lyapunov function; robust stability

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带有时滞的随机区间 Hopfield 神经网络的指数稳定性

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摘要: 讨论了带有可变时滞的随机区间 Hopfield 神经网络的指数稳定性, 利用 Itô 公式和 Lyapunov 函数, 得到了几个关于其指数稳定时滞无关和时滞相关的充分性条件, 推广了现有文献中关于定常时滞随机神经网络及其确定形式的许多结果.

关键词: 随机区间时滞 Hopfield 神经网络; 布朗运动; Itô 公式; Lyapunov 函数; 鲁棒稳定性

1 Introduction

The theoretical research on neural network has made great progress since it was born. In many networks, such as in electronic neural networks, time delay can not be avoided. In fact, the stochastic perturbations can not be avoided either^[1~13]. On the other hand, the system is unavoidable uncertainty, which is due to the existence of modeling errors, can also destroy the stability of the neural networks. So it is very important to discuss the stability and robustness of network against such error and fluctuation^[4,7,8]. To overcome this difficulty, we will propose the stochastic interval delayed Hopfield neural networks (SIDHNN), and derive some robust stability criteria for the networks.

Consider an SIDHNN state equation as follows:

$$\begin{cases} du(t) = [-A_1 u(t) + W_1^T f(u^\tau(t))]dt + \\ \quad \sigma(t, u(t), u^\tau(t))dw(t), \quad t \geq 0, \\ u(s) = \xi(s), \quad -\bar{\tau} \leq s \leq 0. \end{cases} \quad (1)$$

Where $A_j := \{A = \text{diag}(a_{ii})_{n \times n} \in \mathbb{R}^{n \times n} : \underline{A} \leq A \leq \bar{A}, \text{ i.e. } \underline{a}_{ii} \leq a_{ii} \leq \bar{a}_{ii}, i = 1, 2, \dots, n\}$, $W_j^\tau := \{W^\tau = (w_{ij}^\tau)_{n \times n} \in \mathbb{R}^{n \times n} : \underline{W}^\tau \leq W^\tau \leq \bar{W}^\tau, \text{ i.e. } \underline{w}_{ij}^\tau \leq w_{ij}^\tau \leq \bar{w}_{ij}^\tau, i, j = 1, 2, \dots, n\}$ denote weight matrix, $w(t) = (w_1(t), w_2(t), \dots, w_m(t))^T (m \leq n)$ is an m -dimensional Brownian motion defined on a complete probability space (Ω, F, P) with a natural filtration $\{F_t\}_{t \geq 0}$ (i.e. $F_t = \sigma\{w(s) : 0 \leq s \leq t\}$), and $\xi \in C([-2\bar{\tau}, 0]; \mathbb{R}^n)$, $\xi \in L^2_{F_0}([-2\bar{\tau}, 0]; \mathbb{R}^n)$ is F_0 -measurable, $\int_{-\bar{\tau}}^0 E|\xi(s)|^2 ds < \infty$. $u(t) = (u_1(t), u_2(t), \dots, u_n(t))$, $u^\tau(t) = (u_1(t - \tau_1(t)), u_2(t - \tau_2(t)), \dots, u_n(t - \tau_n(t)))$ (where $-\bar{\tau} \leq -\tau_i(t) \leq 0, i = 1, 2, \dots, n$), $\sigma: t \times \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}^{n \times m}$, that is $\sigma(t, u, u^\tau) = (\sigma_{ij}(t, u, u^\tau))_{n \times m}$ (where $\tau'_i(t)$ be similar to $\tau_i(t)$). Define $f(u(t)) = (f_1(u_1(t)), f_2(u_2(t)), \dots, f_n(u_n(t)))$, similar to $f(u^\tau(t))$. Assume, in the whole paper, tho-

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se $\sigma(t, \cdot, \cdot)$ and $f(\cdot)$ (nonlinear sigmoidal activation function) satisfy

H1) There exist constants $L_j, 0 < L_j < \infty, j = 1, 2, \dots, n$ such that the incremental ratio for $f_j: \mathbb{R} \rightarrow \mathbb{R}$ satisfies

$$0 \leq f_j(x_j) - f_j(y_j) \leq L_j(x_j - y_j), \forall x_j, y_j \in \mathbb{R}.$$

H2) There exist constants $b_{ik}, c_{ik}, i = 1, 2, \dots, n, k = 1, 2, \dots, m$ satisfy

$$\sigma_{ik}^2(t, x_i(t), x_i(t - \tau'_i(t))) \leq b_{ik} x_i^2(t) + c_{ik} x_i^2(t - \tau'_i(t)).$$

It is known^[6] that Eq. (1) has a unique global solution on $t \geq 0$, which is denoted by $u(t; \xi)$. Moreover, assume also that $\sigma(t, 0, 0) \equiv 0$ for the sake of stability. Thus system (1) admits an equation solution $u(t; 0) \equiv 0$.

$$\text{Let } w_{ij}^{\tau} = \max_{1 \leq j \leq n} \{ | \underline{w}_{ij}^{\tau} |, | \bar{w}_{ij}^{\tau} | \}, i = 1, 2, \dots, n.$$

Similarly we can define w_{ji}^{τ} .

2 Delay-dependent sufficient criteria

Theorem 1 Let $W = (\max \{ \underline{w}_{ij}^{\tau}, \bar{w}_{ij}^{\tau} \})_{n \times n}$. Assume that

$$\delta = \frac{n \max_{1 \leq i \leq n} \{ \underline{a}_i^{-2} \} \| W \|^2}{2(\min_{1 \leq i \leq n} \{ \underline{a}_i L_i^{-1} \} - \| W \|)} > 0, \quad (2)$$

$\epsilon =$

$$\max_{1 \leq i \leq n} \left\{ \left(\bar{a}_i \bar{\tau} + \sqrt{n \bar{\tau}^2 L_i^2 \sum_{j=1}^n w_{ji}^{\tau * 2} + m \bar{\tau} \sum_{k=1}^m (b_{ik} + c_{ik})} \right)^2 \right\},$$

$$\alpha = \min_{1 \leq i \leq n} \left\{ 1 - \left(\frac{1}{\underline{a}_i} + \delta L_i \right)^2 - \sum_{k=1}^m \left(\frac{1}{\underline{a}_i} + \delta L_i \right) b_{ik} \right\},$$

$$\beta = \max_{1 \leq i \leq n} \left\{ 3\epsilon n L_i^2 \sum_{j=1}^n w_{ji}^{\tau * 2} + \sum_{k=1}^m \left(\frac{1}{\underline{a}_i} + \delta L_i \right) c_{ik} \right\}.$$

If $\alpha > \beta$, then for $\xi \in C([-2\bar{\tau}, 0]; \mathbb{R}^n)$ the zero-solution of system (1) is global 2nd moment exponential stable and almost surely exponential stable.

Proof Define the Lyapunov function as follows

$$V(t, u) = \sum_{i=1}^n (\lambda_i u_i^2 + 2\delta \int_0^{u_i} f_i(s) ds),$$

$$LV(t, u, u^{\tau}) =$$

$$2 \sum_{i=1}^n (\lambda_i u_i(t) + \delta f_i(u_i(t))) \times$$

$$(-a_i u_i(t) + \sum_{j=1}^n w_{ij}^{\tau} f_j(u_j(t - \tau_j(t)))) +$$

$$\sum_{i=1}^n \sum_{k=1}^m (\lambda_i + \delta f'_i) \sigma_{ik}^2(t, u_i(t), u_i(t - \tau'_i(t))) \leq$$

$$- \sum_{i=1}^n \left(1 - \left(\frac{1}{\underline{a}_i} + \delta L_i \right)^2 - \left(\frac{1}{\underline{a}_i} + \delta L_i \right) \sum_{k=1}^m b_{ik} \right) \times$$

$$E u_i^2(t) + \sum_{i=1}^n \left(3\epsilon n L_i^2 \sum_{j=1}^n w_{ji}^{\tau * 2} + \left(\frac{1}{\underline{a}_i} + \delta L_i \right) \sum_{k=1}^m c_{ik} \right) \times$$

$$\sup_{-2\bar{\tau} \leq \vartheta \leq 0} E u_i^2(t + \vartheta) \leq$$

$$- \alpha \sum_{i=1}^n E u_i^2(t) + \beta \sum_{i=1}^n \sup_{-2\bar{\tau} \leq \vartheta \leq 0} E u_i^2(t + \vartheta).$$

By the Razuminkhin-type theory in [14] and condition $\alpha > \beta$, we can conclude that for every $\xi \in C([-2\bar{\tau}, 0]; \mathbb{R}^n)$, the zero-solution of system (1) is globally 2nd moment exponential stable and almost surely exponential stable.

Corollary 1 Let $w^{\tau *} = \max_{1 \leq j \leq n} \{ w_{ji}^{\tau * } \}$, assume that

$$\delta = \frac{n \max_{1 \leq i \leq n} \{ \underline{a}_i^{-2} \} \| W \|^2}{2(\min_{1 \leq i \leq n} \{ \underline{a}_i L_i^{-1} \} - \| W \|)} > 0,$$

$$\epsilon = \max_{1 \leq i \leq n} \left\{ \left(\bar{a}_i \bar{\tau} + n \bar{\tau} w^{\tau * } L_i + \sqrt{m \bar{\tau} \sum_{k=1}^m (b_{ik} + c_{ik})} \right)^2 \right\},$$

$$\alpha = \min_{1 \leq i \leq n} \left\{ 1 - \left(\frac{1}{\underline{a}_i} + \delta L_i \right)^2 - \left(\frac{1}{\underline{a}_i} + \delta L_i \right) \sum_{k=1}^m b_{ik} \right\},$$

$$\beta = \max_{1 \leq i \leq n} \left\{ 3\epsilon n^2 L_i^2 w^{\tau * 2} + \left(\frac{1}{\underline{a}_i} + \delta L_i \right) \sum_{k=1}^m c_{ik} \right\}.$$

If $\alpha < \beta$, for $\xi \in C([-2\bar{\tau}, 0]; \mathbb{R}^n)$ the zero-solution of system (1) is globally 2nd moment exponential stable and almost surely exponential stable

Example 1 Let

$$r_1 = r_2 = 0.5, \bar{\tau} = 0.01, L_1 = L_2 = 0.5,$$

$$\underline{A} = \begin{pmatrix} 1.3 & 0 \\ 0 & 1.3 \end{pmatrix}, \bar{A} = \begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix},$$

$$\underline{W}^{\tau} = \begin{pmatrix} 0.1 & -0.1 \\ 0.05 & -0.15 \end{pmatrix}, \bar{W}^{\tau} = \begin{pmatrix} 0.33 & 0.14 \\ 0.1 & 0.15 \end{pmatrix},$$

$$B = \begin{pmatrix} 0.01 & 0.02 \\ 0.05 & 0.001 \end{pmatrix}, C = \begin{pmatrix} 0.03 & 0.04 \\ 0.05 & 0.2 \end{pmatrix}.$$

By Corollary 1, it is easy to find that $\alpha = 0.3342, \beta = 0.2471$. Obviously, $\alpha > \beta$, so this system is globally exponentially robust stable when $\xi \in C([-0.02, 0], \mathbb{R}^2)$.

Remark 1 As far as the existing papers are concerned, [15 ~ 17] have discussed constant time delay, and those systems are special case of our systems. [17] has studied stochastic Hopfield neural networks with two variable delays, and the delays are single respectively in the certain part and uncertain part. Though this paper

has discussed the same problem with multi-delays, when $\sigma(t, u, u^r) \equiv 0$, we have generalized those certainty systems such as [18, 19].

3 Delay-independent sufficient criteria

Theorem 2 Assume those $\lambda_1, \lambda_2, \dots, \lambda_n$ are non-negative numbers and, $r \in [0, 1]$, let

$$\alpha > \beta + \gamma, \quad (3)$$

where $\alpha = \min_{1 \leq i \leq n} \left\{ 2 \underline{a}_i - \sum_{j=1}^n (w_{ij}^{r*} L_j^{2r} + b_{ij}) \right\}$ ($b_{ij} =$

$0, j = k+1, k+2, \dots, n$), $\gamma = \max_{1 \leq i \leq n} \left\{ \sum_{k=1}^m c_{ik} \right\}$, $\beta =$

$\max_{1 \leq i \leq n} \left\{ \frac{L_i^{2(1-r)}}{\lambda_i} \sum_{j=1}^m \lambda_j w_{ji}^{r*} \right\}$. Then for every $\xi \in$

$C([- \bar{\tau}, 0]; \mathbb{R}^n)$, zero-solution of system (1) is globally 2nd moment exponential stable and almost surely exponential stable.

The proof is omitted.

Corollary 2 Define $w^* = \max_{1 \leq i \leq n} \{ w_{ij}^{r*} \}$ and

$$\alpha = \min_{1 \leq i \leq n} \left\{ 2 \underline{a}_i - w^* \sum_{j=1}^n (L_j^{2r} + b_{ij}) \right\},$$

$$\beta = \max_{1 \leq i \leq n} \left\{ \frac{L_i^{2(1-r)}}{\lambda_i} w^* \sum_{j=1}^m \lambda_j \right\},$$

$$\gamma = \max_{1 \leq i \leq n} \left\{ \sum_{k=1}^m c_{ik} \right\}.$$

If (3) is satisfied, Theorem 1 is also true.

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