Research Article

Multistability and Instability of Competitive Neural Networks with Mexican-Hat-Type Activation Functions

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We investigate the existence and dynamical behaviors of multiple equilibria for competitive neural networks with a class of general Mexican-hat-type activation functions. The Mexican-hat-type activation functions are not monotonously increasing, and the structure of neural networks with Mexican-hat-type activation functions is totally different from those with sigmoidal activation functions or nondecreasing saturated activation functions, which have been employed extensively in previous multistability papers. By tracking the dynamics of each state component and applying fixed point theorem and analysis method, some sufficient conditions are presented to study the multistability and instability, including the total number of equilibria, their locations, and local stability and instability. The obtained results extend and improve the very recent works. Two illustrative examples with their simulations are given to verify the theoretical analysis.

1. Introduction

In the past decades, some famous neural network models, including Hopfield neural networks, cellular neural networks, Cohen-Grossberg neural networks, and bidirectional associative memory neural networks, had been proposed in order to solve some practical problems. It should be mentioned that in the above network models only the neuron activity is taken into consideration. That is, there exists only one type of variables, the state variables of the neurons in these models. However, in a dynamical network, the synaptic weights also vary with respect to time due to the learning process, and the variation of connection weights may have influences on the dynamics of neural network. Competitive neural networks (CNNs) constitute an important class of neural networks, which model the dynamics of cortical cognitive maps with unsupervised synaptic modifications. The CNNs can be written in the following form:

\[
\begin{align*}
\text{STM}: & \quad \dot{x}_i(t) = -d_ix_i(t) + \sum_{j=1}^{N} \alpha_{ij} f_j(x_j(t)) \\
& \quad + B_i \sum_{j=1}^{P} m_{ij}(t) y_j + I_i,
\end{align*}
\]

\[
\text{LTM}: \quad \dot{m}_{ij}(t) = -\alpha m_{ij}(t) + y_j \beta f_i(x_i(t)),
\]

where \(i = 1, 2, \ldots, N\), \(j = 1, 2, \ldots, P\), \(x_i(t)\) is the neuron current activity level, \(f_j(x_j(t))\) is the output of neurons, \(m_{ij}(t)\) is the synaptic efficiency, \(y_j\) is the constant external stimulus, \(\alpha_{ij}\) represents the connection weight between the \(i\)th neuron and the \(j\)th neuron, \(B_i\) is the strength of the external stimulus, \(I_i\) is the constant input, and \(\alpha > 0\) and \(\beta_i \geq 0\) denote disposable scaling constants.
After setting \( S_i(t) = \sum_{j=1}^P m_{ij}(t) y_j = y^T m_i(t) \), where \( y = (y_1, y_2, \ldots, y_P)^T \) and \( m_i(t) = (m_{i1}(t), m_{i2}(t), \ldots, m_{ip}(t))^T \) and assuming the input stimulus \( y \) to be normalized with unit magnitude \( |y|^2 = y_1^2 + \cdots + y_P^2 = 1 \), then the above networks are simplified as

\[
\begin{align*}
\dot{x}_i(t) &= -d_i x_i(t) + \sum_{j=1}^N a_{ij} f_j(x_j(t)) + B_i S_i(t) + I_i, \\
\dot{S}_i(t) &= -\alpha_i S_i(t) + \beta_i f_i(x_i(t)), & i = 1, 2, \ldots, N.
\end{align*}
\]  

(2)

The qualitative analysis of neural dynamics plays an important role in the design of practical neural networks. To solve problems of optimization and signal processing, neural networks have to be designed in such a way that, for a given external input, they exhibit only one globally stable state (i.e., monostability). This matter has been treated in [1–7]. On the other hand, if neural networks are used to analyze associative memories, the coexistence of multiple locally stable equilibria or periodic orbits is required (i.e., multistability or multiperiodicity), since the adrestable memories or patterns are stored as stable equilibria or stable periodic orbits. In monostability analysis, the objective is to derive conditions that guarantee that each network contains only one steady state, and all the trajectories of the network converge to it, whereas in multistability analysis, the networks are allowed to have multiple equilibria or periodic orbits (stable or unstable). In general, the usual global stability conditions are not adequately applicable to multistable networks.

Recently, the multistability or multiperiodicity of neural networks has attracted the attention of many researchers. In [8, 9], based on decomposition of state space, the authors investigated the multistability of delayed Hopfield neural networks and showed that the \( n \)-neuron neural networks can have \( 2^n \) stable orbits located in \( 2^n \) subsets of \( \mathbb{R}^n \). Cao et al. [10] extended the above method to the Cohen-Grossberg neural networks with nondecreasing saturated activation functions with two corner points. In [11, 12], the multistability of almost-periodic solution in delayed neural networks was studied. Kaslik and Sivasundaram [13, 14] firstly revealed the effect of impulse on the multistability of neural networks. In [15–17], high-order synaptic connectivity was introduced into neural networks and the multistability and multiperiodicity were considered, respectively, for high-order neural networks based on decomposition of state space, Cauchy convergence principle, and inequality technique. In [18–22], the authors indicated that under some conditions, there exist \( 3^n \) equilibria for the \( n \)-neuron neural networks and \( 2^n \) of which are locally exponentially stable. In [23], the Hopfield neural networks with nondecreasing piecewise linear activation functions with \( 2r \) corner points were considered. It was proved that under some conditions, the \( n \)-neuron neural networks can have and only have \( (2r+1)^n \) equilibria, \((r+1)^n \) of which are locally exponentially stable and others are unstable. In [24], the multistability of neural networks with \( k + m \) step stair activation functions was discussed based on an appropriate partition of the \( n \)-dimensional state space. It was shown that the \( n \)-neuron neural networks can have \((2k+2m-1)^n \) equilibria, \((k+m)^n \) of which are locally exponentially stable.

**Figure 1:** Mexican-hat-type activation functions (3).

In particular, the case of \( k = m \) was previously discussed in [25]. For more references, see [26–32] and references therein.

It is well known that the type of activation functions plays a very important role in the multistability analysis of neural networks. In the abovementioned and most existing works, the activation functions employed in multistability analysis were mainly focused on sigmoidal activation functions and nondecreasing saturated activation functions, which are all monotonously increasing. In this paper, we will consider a class of continuous Mexican-hat-type activation functions, which are defined as follows (see Figure 1):

\[
f_i(x) = \begin{cases} 
 u_i, & -\infty < x < p_i, \\
 l_{1i} x + c_{1i}, & p_i \leq x \leq r_i, \\
 l_{2i} x + c_{2i}, & r_i < x \leq q_i, \\
 u_i, & q_i < x < +\infty,
\end{cases}
\]

(3)

where \( p_i, r_i, q_i, u_i, l_{1i}, l_{2i}, c_{1i}, c_{2i} \) are constants with \( -\infty < p_i < r_i < q_i < +\infty \), \( l_{1i} > 0 \), and \( l_{2i} < 0 \), \( i = 1, 2, \ldots, N \). In particular, when \( p_i = 1, r_i = q_i = 3, u_i = -1, l_{1i} = 1, l_{2i} = -1, c_{1i} = 0, \) and \( c_{2i} = 2 \) \((i = 1, 2, \ldots, N)\), the above activation functions \( f_i \) reduce to the following special activation functions employed in [33]:

\[
f_i(x) = \begin{cases} 
 -1, & -\infty < x < -1, \\
 x, & -1 \leq x \leq 1, \\
 -x + 2, & 1 < x \leq 3, \\
 -1, & 3 < x < +\infty.
\end{cases}
\]

(4)

It is necessary to point out that the Mexican-hat-type activation functions are not monotonously increasing, which are totally different from sigmoidal activation functions and nondecreasing saturated activation functions. Hence, the results and methods mentioned above cannot be applied to neural networks with activation functions (3). Very recently, the multistability and instability of Hopfield neural networks with activation functions (4) were studied in [33]. Inspired by [33], in this paper, we will investigate the multistability and instability of CNNs with activation functions (3). It should be noted that the structure of CNNs differs from and is more complex than that in [33]. Moreover, the activation functions...
(3) employed in this paper are more general than activation functions (4). More precisely, the contributions of this paper are three-fold as follows.

Firstly, we define four index subsets and present sufficient condition under which the CNNs with activation functions (3) have multiple equilibria, by tracking the dynamics of each state component and applying fixed point theorem. The index subsets are defined in terms of maximum and minimum values, which are different from and less restrictive than those given in [33]. Furthermore, we discuss the exact existence of equilibria for CNNs.

Secondly, based on some analysis method, we analyze the dynamical behaviors of each equilibrium point for CNNs, including local stability and instability. The dynamical behaviors of such system are much more complex than those of Hopfield neural networks considered in [33], due to the complexity of the networks structure and generality of activation functions.

Thirdly, specializing the model and activation functions to those in [33], we show that the obtained results extend and improve the very recent works in [33].

Finally, two examples with their simulations are given to verify and illustrate the validity of the obtained results.

2. Main Results

Firstly, we define the four index subsets as follows:

\[ \mathbb{N}_1 = \left\{ i : I_i < - \min \left\{ -d_i p_i + a_{ii} f_i (p_i), -d_i r_i + a_{ii} f_i (r_i) \right\} - \sum_{j \neq i, j=1}^N \max \{ a_{ij} u_{ij}, a_{ij} v_j \} \right\} \]

\[ - \max \left\{ \alpha_i^{-1} \beta_i u_i B_i, \alpha_i^{-1} \beta_i v_i B_i \right\} \right\}, \]

\[ \mathbb{N}_2 = \left\{ i : d_i r_i - a_{ii} f_i (r_i) - \sum_{j \neq i, j=1}^N \min \{ a_{ij} u_{ij}, a_{ij} v_j \} < I_i < d_i p_i - a_{ii} f_i (p_i) - \sum_{j \neq i, j=1}^N \max \{ a_{ij} u_{ij}, a_{ij} v_j \} \right\} \]

\[ - \max \left\{ \alpha_i^{-1} \beta_i u_i B_i, \alpha_i^{-1} \beta_i v_i B_i \right\} \right\}, \]

\[ \mathbb{N}_3 = \left\{ i : - \min \left\{ -d_i p_i + a_{ii} f_i (p_i), -d_i r_i + a_{ii} f_i (r_i) \right\} - \sum_{j \neq i, j=1}^N \min \{ a_{ij} u_{ij}, a_{ij} v_j \} \right\} \]

\[ - \sum_{j \neq i, j=1}^N \min \{ a_{ij} u_{ij}, a_{ij} v_j \} \]

\[ \mathbb{N}_4 = \left\{ i : I_i > - \min \left\{ -d_i q_i + a_{ii} f_i (q_i), -d_i r_i + a_{ii} f_i (r_i) \right\} - \sum_{j \neq i, j=1}^N \min \{ a_{ij} u_{ij}, a_{ij} v_j \} \right\} \]

\[ - \max \left\{ \alpha_i^{-1} \beta_i u_i B_i, \alpha_i^{-1} \beta_i v_i B_i \right\} \right\}, \]

\[ \mathbb{N}_5 = \left\{ i : I_i > - \min \left\{ -d_i q_i + a_{ii} f_i (q_i), -d_i r_i + a_{ii} f_i (r_i) \right\} - \sum_{j \neq i, j=1}^N \min \{ a_{ij} u_{ij}, a_{ij} v_j \} \right\} \]

\[ - \max \left\{ \alpha_i^{-1} \beta_i u_i B_i, \alpha_i^{-1} \beta_i v_i B_i \right\} \right\}, \]

where \( v_i = f_i (r_i) \). It is easy to see that \( u_i \leq f_i (x) \leq v_i \) for \( x \in \mathbb{R} \).

Remark 1. In this paper, the index subsets are defined in terms of maximum and minimum values, which are different from those given in [33], where they are defined in terms of absolute values. In general, our conditions are less restrictive, which have been shown in [16].

Remark 2. The inequality \( I_i - (a_{ii} + \alpha_i^{-1} \beta_i B_i) l_{i,1} < 0 \) holds for all \( i \in \mathbb{N}_2 \).

Proof. By the definition of index subset \( \mathbb{N}_2 \), \( u_i = f_i (p_i) \) and \( v_i = f_i (r_i) \), we obtain

\[ -d_i p_i + (a_{ii} + \alpha_i^{-1} \beta_i B_i) f_i (p_i) + \sum_{j \neq i, j=1}^N \max \{ a_{ij} u_{ij}, a_{ij} v_j \} + I_i < 0, \]

\[ -d_i r_i + (a_{ii} + \alpha_i^{-1} \beta_i B_i) f_i (r_i) + \sum_{j \neq i, j=1}^N \min \{ a_{ij} u_{ij}, a_{ij} v_j \} + I_i > 0. \]

It follows from (6) and (7) that

\[ -d_i p_i + (a_{ii} + \alpha_i^{-1} \beta_i B_i) f_i (p_i) - d_i r_i + (a_{ii} + \alpha_i^{-1} \beta_i B_i) f_i (r_i) \]

**\[ \mathbb{N}_5 = \left\{ i : I_i > - \min \left\{ -d_i q_i + a_{ii} f_i (q_i), -d_i r_i + a_{ii} f_i (r_i) \right\} - \sum_{j \neq i, j=1}^N \min \{ a_{ij} u_{ij}, a_{ij} v_j \} \right\} \]

\[ - \max \left\{ \alpha_i^{-1} \beta_i u_i B_i, \alpha_i^{-1} \beta_i v_i B_i \right\} \right\}, \]

\[ \mathbb{N}_3 = \left\{ i : - \min \left\{ -d_i p_i + a_{ii} f_i (p_i), -d_i r_i + a_{ii} f_i (r_i) \right\} - \sum_{j \neq i, j=1}^N \min \{ a_{ij} u_{ij}, a_{ij} v_j \} \right\} \]

\[ - \sum_{j \neq i, j=1}^N \min \{ a_{ij} u_{ij}, a_{ij} v_j \} \]

Noting that \( p_i < r_i \) and substituting \( f_i (p_i) = l_{i,1} p_i + e_{i,1} \) and \( f_i (r_i) = l_{i,1} r_i + e_{i,1} \) into (8), we can derive that \( d_i - (a_{ii} + \alpha_i^{-1} \beta_i B_i) l_{i,1} < 0 \) \((i \in \mathbb{N}_2)\).

Remark 3. The inequality \( d_i - (a_{ii} + \alpha_i^{-1} \beta_i B_i) l_{i,2} > 0 \) holds for all \( i \in \mathbb{N}_2 \cup \mathbb{N}_3 \).
Proof. From Remark 2, we get $a_{ii} + \alpha^{-1}_i B_i > 0$ ($i \in \mathbb{N}_2$). Thus, inequality $d_i - (a_{ii} + \alpha^{-1}_i B_i) l_{ii} > 0$ holds for all $i \in \mathbb{N}_2$, due to $l_{ii} < 0$.

By the definition of index subset $\mathbb{N}_3$ and equalities $u_i = f_j(q_i)$, we get

$$
\min \left[ -d_i p_i + a_{ii} f_i(p_i), -d_i r_i + a_{ii} f_i(r_i) \right] + \alpha^{-1}_i B_i v_i > -d_i q_i + a_{ii} f_i(q_i) + \alpha^{-1}_i B_i u_i,
$$

which implies that $-d_i r_i + (a_{ii} + \alpha^{-1}_i B_i) f_i(r_i) > -d_i q_i + (a_{ii} + \alpha^{-1}_i B_i) f_i(q_i)$. Using equalities $f_i(r_i) = l_{ii} r_i + \varepsilon_i$, $f_i(q_i) = l_{ii} q_i + \varepsilon_i$ and noting that $r_i < q_i$, the inequality $d_i - (a_{ii} + \alpha^{-1}_i B_i) l_{ii} > 0$ ($i \in \mathbb{N}_3$) can be proved easily.

It follows from the second equation of system (2) that

$$
-d_i x_i(t) + a_{ii} f_i(x_i(t))
$$

which leads to

$$
\alpha^{-1}_i B_i u_i + (S_i(0) - \alpha^{-1}_i B_i v_i) e^{-\alpha t} \leq S_i(t) \leq \alpha^{-1}_i B_i v_i + (S_i(0) - \alpha^{-1}_i B_i v_i) e^{-\alpha t}.
$$

Therefore, $S_i(0) = \left[ \alpha^{-1}_i B_i u_i, \alpha^{-1}_i B_i v_i \right]$ always implies that $S_i(t) \in \left[ \alpha^{-1}_i B_i u_i, \alpha^{-1}_i B_i v_i \right]$. That is, if $S(t) \in \prod_{i=1}^N \left[ \alpha^{-1}_i B_i u_i, \alpha^{-1}_i B_i v_i \right]$, then the solution $S(t)$ is bounded for all $t \geq 0$.

Let $(x(t), S(t)) \in \mathbb{R}^N \times \prod_{i=1}^N \left[ \alpha^{-1}_i B_i u_i, \alpha^{-1}_i B_i v_i \right]$. In the following, we will discuss the dynamics of state components $x_i(t)$ for $i \in \mathbb{N}_3$ ($i = 1, 2, 3, 4$), respectively.

Lemma 4. All the state components $x_i(t)$, $i \in \mathbb{N}_3$, will flow to the interval $(-\infty, p_i]$ when $t$ tends to $+\infty$.

Proof. According to the different location of $x_i(0)$, there are two cases for us to discuss.

Case (i). Consider $x_i(0) \in (-\infty, p_i]$. In this case, if there exists some $t^* \geq 0$ such that $x_i(t^*) = p_i$, $x_i(t) \leq p_i$ for $0 \leq t \leq t^*$, then it follows from system (2) and the definition of $\mathbb{N}_3$ that

$$
\dot{x}_i(t^*) = -d_i x_i(t^*) + a_{ii} f_i(x_i(t^*)) + \sum_{j \neq i}^N a_{ij} f_j(x_j(t^*)) + B_i S_i(t^*) + I_i
$$

$$
\leq -d_i p_i + a_{ii} f_i(p_i) + \sum_{j \neq i}^N \max \{a_{ij} u_j, a_{ij} v_j\} + \max \{\alpha^{-1}_i B_i u_i, \alpha^{-1}_i B_i v_i\} + I_i < 0.
$$

Hence, $x_i(t)$ would never get out of $(-\infty, p_i]$. Similarly, we can also conclude that once $x_i(T_0) \in (-\infty, p_i]$ for some $T_0 > 0$, then $x_i(t)$ would never escape from $(-\infty, p_i]$ for all $t \geq T_0$.

Case (ii). Consider $x_i(0) \in (p_i, +\infty)$. In this case, we claim that $x_i(t)$ would monotonically decrease until it reaches the interval $(-\infty, p_i]$ in some finite time.

In fact, when $x_i(t) \in (q_i, +\infty)$, noting that the definition of $N_3$ and $f_i(p_i) = f_i(q_i)$, we obtain

$$
\dot{x}_i(t) = -d_i x_i(t) + a_{ii} f_i(x_i(t)) + \sum_{j \neq i}^N a_{ij} f_j(x_j(t)) + B_i S_i(t) + I_i
$$

$$
\leq -d_i q_i + a_{ii} f_i(q_i) + \sum_{j \neq i}^N a_{ij} f_j(x_j(t)) + B_i S_i(t) + I_i
$$

$$
\leq -d_i q_i + a_{ii} f_i(q_i) + \sum_{j \neq i}^N \max \{a_{ij} u_j, a_{ij} v_j\} + \max \{\alpha^{-1}_i B_i u_i, \alpha^{-1}_i B_i v_i\} + I_i < 0;
$$

when $x_i(t) \in (r_i, q_i]$, by virtue of equalities $f_i(r_i) = l_{ii} r_i + \varepsilon_i$ and $f_i(q_i) = l_{ii} q_i + \varepsilon_i$, we get

$$
\dot{x}_i(t) = -d_i x_i(t) + a_{ii} (l_{ii} x_i(t) + \varepsilon_i)
$$

$$
\leq \sum_{j \neq i}^N a_{ij} f_j(x_j(t)) + B_i S_i(t) + I_i
$$

$$
\leq \sum_{j \neq i}^N \max \{a_{ij} u_j, a_{ij} v_j\} + \max \{\alpha^{-1}_i B_i u_i, \alpha^{-1}_i B_i v_i\} + I_i < 0;
$$

when $x_i(t) \in (p_i, r_i]$, it follows from $f_i(p_i) = l_{ii} p_i + \varepsilon_i$, $f_i(r_i) = l_{ii} r_i + \varepsilon_i$ and system (2) that

$$
\dot{x}_i(t) = -d_i x_i(t) + a_{ii} (l_{ii} x_i(t) + \varepsilon_i)
$$

$$
\leq \sum_{j \neq i}^N a_{ij} f_j(x_j(t)) + B_i S_i(t) + I_i
$$

$$
\leq \sum_{j \neq i}^N \max \{a_{ij} u_j, a_{ij} v_j\} + \max \{\alpha^{-1}_i B_i u_i, \alpha^{-1}_i B_i v_i\} + I_i < 0;
$$}

In fact, when $x_i(t) \in (q_i, +\infty)$, noting that the definition of $N_3$ and $f_i(p_i) = f_i(q_i)$, we obtain

$$
\dot{x}_i(t) = -d_i x_i(t) + a_{ii} f_i(x_i(t))
$$

$$
+ \sum_{j \neq i}^N a_{ij} f_j(x_j(t)) + B_i S_i(t) + I_i
$$

$$
\leq -d_i q_i + a_{ii} f_i(q_i) + \sum_{j \neq i}^N a_{ij} f_j(x_j(t)) + B_i S_i(t) + I_i
$$

$$
\leq -d_i q_i + a_{ii} f_i(q_i) + \sum_{j \neq i}^N \max \{a_{ij} u_j, a_{ij} v_j\} + \max \{\alpha^{-1}_i B_i u_i, \alpha^{-1}_i B_i v_i\} + I_i < 0;
$$

when $x_i(t) \in (r_i, q_i]$, by virtue of equalities $f_i(r_i) = l_{ii} r_i + \varepsilon_i$ and $f_i(q_i) = l_{ii} q_i + \varepsilon_i$, we get

$$
\dot{x}_i(t) = -d_i x_i(t) + a_{ii} (l_{ii} x_i(t) + \varepsilon_i)
$$

$$
\leq \sum_{j \neq i}^N a_{ij} f_j(x_j(t)) + B_i S_i(t) + I_i
$$

$$
\leq \sum_{j \neq i}^N \max \{a_{ij} u_j, a_{ij} v_j\} + \max \{\alpha^{-1}_i B_i u_i, \alpha^{-1}_i B_i v_i\} + I_i < 0;
$$

when $x_i(t) \in (p_i, r_i]$, it follows from $f_i(p_i) = l_{ii} p_i + \varepsilon_i$, $f_i(r_i) = l_{ii} r_i + \varepsilon_i$ and system (2) that

$$
\dot{x}_i(t) = -d_i x_i(t) + a_{ii} (l_{ii} x_i(t) + \varepsilon_i)
$$

$$
\leq \sum_{j \neq i}^N a_{ij} f_j(x_j(t)) + B_i S_i(t) + I_i
$$

$$
\leq \sum_{j \neq i}^N \max \{a_{ij} u_j, a_{ij} v_j\} + \max \{\alpha^{-1}_i B_i u_i, \alpha^{-1}_i B_i v_i\} + I_i < 0;
$$

when $x_i(t) \in (p_i, r_i]$, it follows from $f_i(p_i) = l_{ii} p_i + \varepsilon_i$, $f_i(r_i) = l_{ii} r_i + \varepsilon_i$ and system (2) that

$$
\dot{x}_i(t) = -d_i x_i(t) + a_{ii} (l_{ii} x_i(t) + \varepsilon_i)
$$

$$
\leq \sum_{j \neq i}^N a_{ij} f_j(x_j(t)) + B_i S_i(t) + I_i
$$

$$
\leq \sum_{j \neq i}^N \max \{a_{ij} u_j, a_{ij} v_j\} + \max \{\alpha^{-1}_i B_i u_i, \alpha^{-1}_i B_i v_i\} + I_i < 0.$$
\[\dot{x}_i(t) = -d_i x_i(t) + a_{ii} f_i(x_i(t)) + B_i S_i(t) + I_i + \sum_{j \neq i, j=1}^{N} a_{ij} f_j(x_j(t)) + \sum_{j \neq i, j=1}^{N} \max\{a_{ij} u_j, a_{ij} V_j\} + \max\{\alpha_i^{-1} \beta_i u_i B_i, \alpha_i^{-1} \beta_i V_i B_i\} + I_i \leq 0.\]  

(15)

In summary, wherever the initial state \(x_i(0) (i \in \mathbb{N}_i)\) is located in, \(x_i(t)\) would flow to and enter the interval \((-\infty, p_i]\) and stay in this interval forever.

**Lemma 5.** All the state components \(x_i(t), i \in \mathbb{N}_i\), will flow to the interval \([r_i, q_i]\) when \(t\) tends to \(+\infty\).

**Proof.** We prove it in the following three cases due to the different location of \(x_i(0)\).

**Case (i).** Consider \(x_i(0) \in [r_i, q_i]\). In this case, if there exists some \(t^* \geq 0\) such that \(x_i(t^*) = r_i, r_i \leq x_i(t) \leq q_i\) for \(0 \leq t \leq t^*\), then we have

\[\dot{x}_i(t^*) = -d_i x_i(t^*) + a_{ii} f_i(x_i(t^*)) + B_i S_i(t^*) + I_i + \sum_{j \neq i, j=1}^{N} a_{ij} f_j(x_j(t^*)) \geq 0.\]  

(16)

Similarly, if there exists some \(t^{**} \geq 0\) such that \(x_i(t^{**}) = q_i, r_i \leq x_i(t) \leq q_i\) for \(0 \leq t \leq t^{**}\), then we get

\[\dot{x}_i(t^{**}) \leq -d_i q_i + a_{ii} f_i(q_i) + \sum_{j \neq i, j=1}^{N} \max\{a_{ij} u_j, a_{ij} V_j\} + \max\{\alpha_i^{-1} \beta_i u_i B_i, \alpha_i^{-1} \beta_i V_i B_i\} + I_i < 0.\]  

(17)

From the above two inequalities, we know that if \(x_i(0) \in [r_i, q_i]\), then \(x_i(t)\) would never get out of this interval. In the same way, we can also obtain that if there exists some \(T_0 > 0\) such that \(x_i(T_0) \in [r_i, q_i]\), then \(x_i(t)\) would stay in it for all \(t \geq T_0\).

**Case (ii).** Consider \(x_i(0) \in (-\infty, r_i]\). When \(x_i(t) \in (-\infty, p_i]\), from the definition of index subset \(\mathbb{N}_i\), we get

\[\dot{x}_i(t) \geq -d_i p_i + a_{ii} f_i(p_i) + \sum_{j \neq i, j=1}^{N} \min\{a_{ij} u_j, a_{ij} V_j\} + \min\{\alpha_i^{-1} \beta_i u_i B_i, \alpha_i^{-1} \beta_i V_i B_i\} + I_i > 0; \]  

(18)

when \(x_i(t) \in (p_i, r_i]\), it follows from the definition of index subset \(\mathbb{N}_i\), equalities \(f_i(p_i) = l_{i1} p_i + c_{i1}, f_i(r_i) = l_{i1} r_i + c_{i1}\) that

\[\dot{x}_i(t) \geq -d_i x_i(t) + a_{ii} f_i(l_{i1}, x_i(t) + c_{i1}) + \sum_{j \neq i, j=1}^{N} a_{ij} f_j(x_j(t)) + B_i S_i(t) + I_i + \sum_{j \neq i, j=1}^{N} \max\{a_{ij} u_j, a_{ij} V_j\} + \max\{\alpha_i^{-1} \beta_i u_i B_i, \alpha_i^{-1} \beta_i V_i B_i\} + I_i > 0.\]  

(19)

Thus, in this case, \(x_i(t)\) would monotonously increase until it reaches the interval \([r_i, q_i]\).

**Case (iii).** Consider \(x_i(0) \in (q_i, +\infty)\). When \(x_i(t) \in (q_i, +\infty)\), it follows that

\[\dot{x}_i(t) \leq -d_i q_i + a_{ii} f_i(q_i) + \sum_{j \neq i, j=1}^{N} \max\{a_{ij} u_j, a_{ij} V_j\} + \max\{\alpha_i^{-1} \beta_i u_i B_i, \alpha_i^{-1} \beta_i V_i B_i\} + I_i < 0.\]  

(20)

Therefore, \(x_i(t)\) would monotonously decrease until it enters the interval \([r_i, q_i]\).

In summary, wherever the initial state \(x_i(0) (i \in \mathbb{N}_i)\) is located in, \(x_i(t)\) would flow to and enter the interval \([r_i, q_i]\) and stay in it finally.

**Lemma 6.** All the state components \(x_i(t), i \in \mathbb{N}_i\), will flow to the interval \([q_i, +\infty)\) when \(t\) tends to \(+\infty\).

**Proof.** Similar to the proof of Lemmas 4 and 5, we will prove it in the following two cases.

**Case (i).** Consider \(x_i(0) \in (q_i, +\infty)\). If there exists some \(t^* \geq 0\) such that \(x_i(t^*) = q_i, x_i(t) \geq q_i\) for \(0 \leq t \leq t^*\), then

\[\dot{x}_i(t^*) \geq -d_i q_i + a_{ii} f_i(q_i) + \sum_{j \neq i, j=1}^{N} \min\{a_{ij} u_j, a_{ij} V_j\} + \min\{\alpha_i^{-1} \beta_i u_i B_i, \alpha_i^{-1} \beta_i V_i B_i\} + I_i > 0.\]  

(21)
Therefore, \(x_i(t)\) would never get out of \([q_i, +\infty)\). By the same method, we also get that once \(x_i(T_0) \in [q_i, +\infty)\) for some \(T_0 > 0\), then \(x_i(t)\) would stay in this interval for all \(t \geq T_0\).

Case (ii). Consider \(x_i(0) \in (−\infty, q_i)\). In this case, we claim that \(x_i(t)\) would monotonously increase until it enters the interval \([q_i, +\infty)\).

In fact, when \(x_i(t) \in (−\infty, p_i)\), we have

\[
\dot{x}_i(t) \geq -d_i p_i + a_{ii} f_i(p_i) + \sum_{j \neq i, j=1}^N \min\{a_{ij} u_j, a_{ij} v_j\} + \min\{\alpha^{-1} \beta_j u_j B_j, \alpha^{-1} \beta_j v_j B_j\} + I_i \geq -d_i q_i + a_{ii} f_i(q_i) + \sum_{j \neq i, j=1}^N \min\{a_{ij} u_j, a_{ij} v_j\} + \min\{\alpha^{-1} \beta_j u_j B_j, \alpha^{-1} \beta_j v_j B_j\} + I_i > 0;
\]

when \(x_i(t) \in [p_i, r_i]\), we get

\[
\dot{x}_i(t) = -d_i x_i(t) + a_{ii} \left(l_{ii} x_i(t) + c_{i1}\right) + \sum_{j \neq i, j=1}^N a_{ij} f_j(x_j(t)) + B_i S_i(t) + I_i \geq -d_i q_i + a_{ii} f_i(q_i) - d_i r_i + a_{ii} f_i(r_i) + \sum_{j \neq i, j=1}^N a_{ij} f_j(x_j(t)) + B_i S_i(t) + I_i \geq -d_i q_i + a_{ii} f_i(q_i) - d_i r_i + a_{ii} f_i(r_i) + \sum_{j \neq i, j=1}^N \min\{a_{ij} u_j, a_{ij} v_j\} + \min\{\alpha^{-1} \beta_j u_j B_j, \alpha^{-1} \beta_j v_j B_j\} + I_i > 0;
\]

when \(x_i(t) \in (r_i, q_i)\), then we obtain

\[
\dot{x}_i(t) = -d_i x_i(t) + a_{ii} \left(l_{ii} x_i(t) + c_{i2}\right) + \sum_{j \neq i, j=1}^N a_{ij} f_j(x_j(t)) + B_i S_i(t) + I_i \geq -d_i q_i + a_{ii} f_i(q_i) - d_i r_i + a_{ii} f_i(r_i) + \sum_{j \neq i, j=1}^N \min\{a_{ij} u_j, a_{ij} v_j\} + \min\{\alpha^{-1} \beta_j u_j B_j, \alpha^{-1} \beta_j v_j B_j\} + I_i > 0.
\]

In summary, wherever the initial state \(x_i(0) \in \mathbb{N}_i\) is located in, \(x_i(t)\) would flow to \([q_i, +\infty)\) when \(t \) is big enough and stay in it forever. □

Denote \((-\infty, p_i) = (-\infty, p_i)^1 \times [p_i, r_i)^0 \times [r_i, +\infty)^0\); \([p_i, r_i) = (-\infty, p_i)^0 \times [p_i, r_i)^1 \times [r_i, +\infty)^0\); \([r_i, +\infty) = (-\infty, p_i)^0 \times [p_i, r_i)^0 \times [r_i, +\infty)^1\). For any \(i \in \mathbb{N}_2\), let \(\delta_i = (\delta^{(1)}_i, \delta^{(2)}_i, \delta^{(3)}_i) = (1, 0, 0)\) or \((0, 1, 0)\) or \((0, 0, 1)\) and define

\[
\Omega_{\delta_i} = \prod_{i \in \mathbb{N}_2} (-\infty, p_i)^{\delta^{(1)}_i} \times [p_i, r_i)^{\delta^{(2)}_i} \times [r_i, +\infty)^{\delta^{(3)}_i},
\]

\[
\Omega = \prod_{i \in \mathbb{N}_4} (-\infty, p_i) \times \prod_{i \in \mathbb{N}_2} \Omega_{\delta_i} \times \prod_{i \in \mathbb{N}_4} (q_i, +\infty).
\]

It is easy to see that there exist \(3^{\|\mathbb{N}_2\|}\) \(\Omega_{\delta}\)-type regions, where \(\mathbb{N}_2\) denotes the number of elements in the set \(\mathbb{N}_2\). Now, we will prove the following theorem on the existence of multiple equilibria for system (2).

**Theorem 7.** Suppose that \(\mathbb{N}_1 \cup \mathbb{N}_2 \cup \mathbb{N}_3 \cup \mathbb{N}_4 \equiv \{1, 2, \ldots, N\}\). Then, system (2) with activation functions (3) has \(3^{\|\mathbb{N}_2\|}\) equilibria.

**Proof.** Pick a region arbitrarily \(\Omega_{\delta} \equiv \prod_{i \in \mathbb{N}_2} [\alpha^{-1}_i \beta_i u_i, \alpha^{-1}_i \beta_i v_i]\); we will show that there exists an equilibrium point located in each \(\Omega_{\delta} \equiv \prod_{i \in \mathbb{N}_2} [\alpha^{-1}_i \beta_i u_i, \alpha^{-1}_i \beta_i v_i]\).

Denote \(\mathbb{N}_{2,1} = \{i \in \mathbb{N}_2 : \delta^{(1)}_i = 1\}, \mathbb{N}_{2,2} = \{i \in \mathbb{N}_2 : \delta^{(2)}_i = 1\}, \mathbb{N}_{2,3} = \{i \in \mathbb{N}_2 : \delta^{(3)}_i = 1\}\). It is easy to see that \(\mathbb{N}_{2,1} \cup \mathbb{N}_{2,2} \cup \mathbb{N}_{2,3} = \mathbb{N}_2\).

Let \((x^T(t), S^T(t))\) be any solution of system (2) with initial state \((x^T(0), S^T(0)) \in \Omega_{\delta} \equiv \prod_{i \in \mathbb{N}_2} [\alpha^{-1}_i \beta_i u_i, \alpha^{-1}_i \beta_i v_i]\). Then, for \(i \in \mathbb{N}_{2,2}\), if there exists some \(t^* \geq 0\) such that \(x_i(t^*) = p_i\), then we get from the definition of index subset \(\mathbb{N}_2\) that

\[
\dot{x}_i(t^*) = -d_i x_i(t^*) + a_{ii} f_i(x_i(t^*)) + \sum_{j \neq i, j=1}^N a_{ij} f_j(x_j(t^*)) + B_i S_i(t^*) + I_i \leq -d_i p_i + a_{ii} f_i(p_i) + \sum_{j \neq i, j=1}^N \max\{a_{ij} u_j, a_{ij} v_j\} + \max\{\alpha^{-1}_i \beta_i u_i B_i, \alpha^{-1}_i \beta_i v_i B_i\} + I_i < 0;
\]

Similarly, for \(i \in \mathbb{N}_{2,3}\), if \(x_i(t) \geq q_i\), note that \(f_j(p_i) = f_j(q_i)\); then

\[
\dot{x}_i(t) \leq -d_i q_i + a_{ii} f_i(q_i) + \sum_{j \neq i, j=1}^N \max\{a_{ij} u_j, a_{ij} v_j\} + \max\{\alpha^{-1}_i \beta_i u_i B_i, \alpha^{-1}_i \beta_i v_i B_i\} + I_i \leq -d_i p_i + a_{ii} f_i(p_i) + \sum_{j \neq i, j=1}^N \max\{a_{ij} u_j, a_{ij} v_j\} + \max\{\alpha^{-1}_i \beta_i u_i B_i, \alpha^{-1}_i \beta_i v_i B_i\} + I_i < 0,
\]
and if there exists some $t^{**} \geq 0$ such that $x_i(t^{**}) = r_i$, then

$$
\dot{x}_i(t^{**}) \geq -d_ir_i + a_{ij}f_j(r_j) + \sum_{j \neq j=1}^N \min \{a_{ij}y_j, a_{ij}y_j^{-1}\}
+ \min \{a_i^{-1}\beta_i B_i, a_i^{-1}\beta_i y_i B_i\} + I_i > 0.
$$

Equivalently, the above equations can be rewritten as

$$
-d_i x_i + (a_i + a_i^{-1}\beta_i B_i) u_i + \sum_{j \neq j=1}^N a_{ij}f_j(x_j) + \sum_{j \neq j=1}^N a_{ij}f_j(x_j) + I_i = 0,
$$

$$
S_i = a_i^{-1}\beta_i f_i(x_i), \quad i = 1, 2, \ldots, N.
$$

In $\Omega_3 \times \prod_{i=1}^N [\alpha_i^{-1}\beta_i u_i, \alpha_i^{-1}\beta_i y_i]$, any equilibrium point of system (2) satisfies the following equations:

$$
-d_i x_i + (a_i + a_i^{-1}\beta_i B_i) u_i + \sum_{j \neq j=1}^N a_{ij}f_j(x_j) + \sum_{j \neq j=1}^N a_{ij}f_j(x_j) + I_i = 0,
$$

where $r_i$ is the equilibrium point.

Note that any equilibrium point of system (2) is a root of the following equations:

$$
-d_i x_i + (a_i + a_i^{-1}\beta_i B_i) u_i + \sum_{j \neq j=1}^N a_{ij}f_j(x_j) + \sum_{j \neq j=1}^N a_{ij}f_j(x_j) + I_i = 0,
$$

$$
S_i = a_i^{-1}\beta_i f_i(x_i), \quad i = 1, 2, \ldots, N.
$$

In the subset region $\prod_{i=1}^N [p_i, r_i] \times \prod_{i=1}^N [q_i, r_i]$, define a map $\Gamma$ as follows:

$$
\Gamma_i(x_i) = \frac{1}{d_i - (a_i + a_i^{-1}\beta_i B_i) l_i},
$$

$$
\times \left( (a_i + a_i^{-1}\beta_i B_i) c_i + \sum_{j \neq j=1}^N a_{ij}f_j(x_j)
+ \sum_{j \neq j=1}^N a_{ij}f_j(x_j) + I_i = 0,
\right), \quad i = 1, 2, \ldots, N.
$$

In the subset region $\prod_{i=1}^N [p_i, r_i] \times \prod_{i=1}^N [q_i, r_i]$, define a map $\Gamma$ as follows:

$$
\Gamma_i(x_i) = \frac{1}{d_i - (a_i + a_i^{-1}\beta_i B_i) l_i},
$$

$$
\times \left( (a_i + a_i^{-1}\beta_i B_i) c_i + \sum_{j \neq j=1}^N a_{ij}f_j(x_j)
+ \sum_{j \neq j=1}^N a_{ij}f_j(x_j) + I_i = 0,
\right), \quad i = 1, 2, \ldots, N.
$$

In the subset region $\prod_{i=1}^N [p_i, r_i] \times \prod_{i=1}^N [q_i, r_i]$, define a map $\Gamma$ as follows:

$$
\Gamma_i(x_i) = \frac{1}{d_i - (a_i + a_i^{-1}\beta_i B_i) l_i},
$$

$$
\times \left( (a_i + a_i^{-1}\beta_i B_i) c_i + \sum_{j \neq j=1}^N a_{ij}f_j(x_j)
+ \sum_{j \neq j=1}^N a_{ij}f_j(x_j) + I_i = 0,
\right), \quad i = 1, 2, \ldots, N.
$$
\[
\Gamma_i(x_i) = \frac{1}{d_i - (a_{ii} + \alpha_i^{-1} \beta_i B_i) l\{i\}} \times \left[ (a_{ii} + \alpha_i^{-1} \beta_i B_i) c_{i,1} + \sum_{j \in \mathbb{N}_i, j \neq i} a_j u_j + \sum_{j \in \mathbb{N}_i} a_j (L_{i,j} x_j + c_{j,1}) + \sum_{j \neq i} a_j (L_{i,j} x_j + c_{j,2}) + \sum_{j \neq i} a_j u_j + I_i \right], \quad i \in \mathbb{N}_{2,3} \cup \mathbb{N}_3.
\] (33)

For \( i \in \mathbb{N}_{2,3} \), substituting \( f_i(p_i) = l_{i,1} p_i + c_{i,1} \) into (6) and noting that \( d_i - (a_{ii} + \alpha_i^{-1} \beta_i B_i) l\{i\} < 0 \) (Remark 2), we get that
\[
\Gamma_i(x_i) \geq \frac{1}{d_i - (a_{ii} + \alpha_i^{-1} \beta_i B_i) l\{i\}} \times \left[ (a_{ii} + \alpha_i^{-1} \beta_i B_i) c_{i,1} + \sum_{j \neq i} \max \{a_j u_j, a_j v_j\} + I_i \right] > p_i. \quad (34)
\]

Similarly, substituting \( f_i(r_i) = l_{i,1} r_i + c_{i,1} \) into (7) results in
\[
\Gamma_i(x_i) \leq \frac{1}{d_i - (a_{ii} + \alpha_i^{-1} \beta_i B_i) l\{i\}} \times \left[ (a_{ii} + \alpha_i^{-1} \beta_i B_i) c_{i,1} + \sum_{j \neq i} \max \{a_j u_j, a_j v_j\} + I_i \right] < r_i. \quad (35)
\]

For \( i \in \mathbb{N}_{2,3} \), note that \(-d_i p_i > -d_i q_i, f_i(p_i) = f_i(q_i) = l_{i,2} q_i + c_{i,2}\), and \( f_i(r_i) = l_{i,2} r_i + c_{i,2}\); it follows from (6)-(7) and \( d_i - (a_{ii} + \alpha_i^{-1} \beta_i B_i) l\{i\} > 0 \) (Remark 3) that
\[
\Gamma_i(x_i) \geq \frac{1}{d_i - (a_{ii} + \alpha_i^{-1} \beta_i B_i) l\{i\}} \times \left[ (a_{ii} + \alpha_i^{-1} \beta_i B_i) c_{i,2} + \sum_{j \neq i} \min \{a_j u_j, a_j v_j\} + I_i \right] > r_i, \quad (36)
\]

For \( i \in \mathbb{N}_3 \), note that \( v_i = f_i(r_i) = l_{i,2} r_i + c_{i,2} \) and \( u_i = f_i(q_i) = l_{i,2} q_i + c_{i,2} \); we can get from the definition of index subset \( \mathbb{N}_3 \) that
\[
-d_i r_i + (a_{ii} + \alpha_i^{-1} \beta_i B_i) l\{i\} r_i + c_{i,2} > 0, \quad (37)
\]
\[
-d_i q_i + (a_{ii} + \alpha_i^{-1} \beta_i B_i) l\{i\} q_i + c_{i,2} > 0. \quad (38)
\]

That is,
\[
\left[-d_i + (a_{ii} + \alpha_i^{-1} \beta_i B_i) l\{i\}\right] r_i + (a_{ii} + \alpha_i^{-1} \beta_i B_i) c_{i,2} + \sum_{j \neq i} \min \{a_j u_j, a_j v_j\} + I_i > 0, \quad (39)
\]

Then combining inequality \( d_i - (a_{ii} + \alpha_i^{-1} \beta_i B_i) l\{i\} > 0 \) and (38) together gives
\[
\Gamma_i(x_i) \geq \frac{1}{d_i - (a_{ii} + \alpha_i^{-1} \beta_i B_i) l\{i\}} \times \left[ (a_{ii} + \alpha_i^{-1} \beta_i B_i) c_{i,2} + \sum_{j \neq i} \min \{a_j u_j, a_j v_j\} + I_i \right] > r_i, \quad (40)
\]
\[ \Gamma_i(x_i) \leq \frac{1}{d_i} - (a_i + \alpha_i^{-1}B_i) I_i \]
\[ \times \left( a_i + \alpha_i^{-1}B_i \right) \epsilon_{i,2} + \sum_{j \neq i} \max \{ a_{ij} u_j, a_{ij} v_j \} + I_i \]
\[ < q_i. \]

Therefore, the map \( \Gamma \) maps a bounded and closed set into itself. Applying Brouwer’s fixed point theorem, there exists one \( (x_i^*, x_{i+1}^*, \ldots, x_N^*) \in \prod_{i \in N_2} \{ p_i, r_i \} \times \prod_{i \in N_3 \cup N_4} [r_i, q_i] \) such that
\[ \Gamma \left( x_i^*, x_{i+1}^*, \ldots, x_N^* \right) = \left( x_i^*, x_{i+1}^*, \ldots, x_N^* \right), \]

where \( i, i+1, \ldots, i \) represent the elements of index subset \( N_2, N_3, N_4 \).

Then for \( i \in N_1 \cup N_{2,1} \), define
\[ F_i(x_i) = -d_i x_i + (a_i + \alpha_i^{-1}B_i) u_i \]
\[ + \sum_{j \in N_2} a_{ij} u_j + \sum_{j \in N_2} a_{ij} (l_{i,1} x_j^* + c_{j,1}) \]
\[ + \sum_{j \in N_3 \cup N_4} a_{ij} (l_{i,2} x_j^* + c_{j,2}) + \sum_{j \in N_4} a_{ij} u_j + I_i. \]

By virtue of the definition of index subsets \( N_1 \) and \( N_2 \), we have
\[ \lim_{k \to -\infty} F_i(x_i) = +\infty \]
and
\[ F_i(p_i) = -d_i p_i + (a_i + \alpha_i^{-1}B_i) u_i \]
\[ + \sum_{j \in N_2} a_{ij} u_j + \sum_{j \in N_2} a_{ij} (l_{i,1} x_j^* + c_{j,1}) \]
\[ + \sum_{j \in N_3 \cup N_4} a_{ij} (l_{i,2} x_j^* + c_{j,2}) + \sum_{j \in N_4} a_{ij} u_j + I_i < 0. \]

Therefore, there exists the unique \( x_i^* \in (-\infty, p_i) \) such that \( F_i(x_i^*) = 0 \).

Similarly, for \( i \in N_4 \), define
\[ F_i(x_i) = -d_i x_i + (a_i + \alpha_i^{-1}B_i) u_i \]
\[ + \sum_{j \in N_1 \cup N_2} a_{ij} u_j + \sum_{j \in N_2} a_{ij} (l_{i,1} x_j^* + c_{j,1}) \]
\[ + \sum_{j \in N_3 \cup N_4} a_{ij} (l_{i,2} x_j^* + c_{j,2}) + \sum_{j \in N_4} a_{ij} u_j + I_i, \]
and we can also derive that there exists the unique \( x_i^* \in (q_i, +\infty) \) such that \( F_i(x_i^*) = 0 \).

Remark 8. \( x^* \) is located in the interior of \( \Omega_3 \).

**Proof.** Note \( x^* \) satisfies the following equations:
\[ -d_i x_i^* + (a_i + \alpha_i^{-1}B_i) f_i(x_i^*) \]
\[ + \sum_{j \neq i} a_{ij} f_j(x_j^*) + I_i = 0, \quad i = 1, 2, \ldots, N. \]

In the following, we prove that \( x_i^* \neq p_i, i \in N_1 \). Otherwise, from the definition of \( N_1 \), we have
\[ 0 = -d_i p_i + (a_i + \alpha_i^{-1}B_i) f_i(p_i) \]
\[ + \sum_{j \neq i} a_{ij} f_j(p_j) + \sum_{j \neq i} \max \{ a_{ij} u_j, a_{ij} v_j \} \]
\[ + \max \{ \alpha_i^{-1}B_i u_i, \alpha_i^{-1}B_i v_i \} + I_i < 0, \]

that is a contradiction. Similarly, from the definition of \( N_2 \cup N_4 \), we can also get
\[ x_i^* \neq p_i, r_i, q_i, \quad i \in N_2, \]
\[ x_i^* \neq r_i, q_i, \quad i \in N_3, \]
\[ x_i^* \neq q_i, \quad i \in N_4. \]

Therefore, \( x^* \) is located in the interior of \( \Omega_3 \).

Remark 9. Suppose that \( N_1 \cup N_2 \cup N_3 \cup N_4 = \{1, 2, \ldots, N\} \). Furthermore, if the following conditions
\[ \sum_{j \in N_2} |a_{ij} l_{i,1}| + \sum_{j \in N_3 \cup N_4} |a_{ij} l_{i,2}| \]
\[ < (a_i + \alpha_i^{-1}B_i) l_{i,1} - d_i, \quad i \in N_2, \]
\[ \sum_{j \in N_2} |a_{ij} l_{i,1}| + \sum_{j \in N_3 \cup N_4} |a_{ij} l_{i,2}| \]
\[ < d_i - (a_i + \alpha_i^{-1}B_i) l_{i,2}, \quad i \in N_2 \cup N_3 \cup N_4, \]
hold, then system (2) with activation functions (3) can have and only have \( S^3 \) equilibria.

**Proof.** From the proof of Theorem 7, we only need to prove that the fixed point of \( \Gamma \) is unique. In fact, suppose that
there exists another fixed point \((x_{i^*}, x_{i^*}, \ldots, x_{i^*})^T \in \prod_{i \in \mathbb{N}_2} \{p_i, r_i\} \times \prod_{i \in \mathbb{N}_2 \cup \mathbb{N}_3} [r_i, q_i]\) such that
\[
\Gamma \left( x_{i^*}, x_{i^*}, \ldots, x_{i^*} \right) = (x_{i^*}, x_{i^*}, \ldots, x_{i^*}).
\] (49)

Denote the index \(i_i\) such that
\[
\left| x_{i_i} - x_{i_i^*} \right| = \max_{i \in \mathbb{N}_2 \cup \mathbb{N}_3} \left| x_i^* - x_i^* \right|.
\] (50)

If \(i_i \in \mathbb{N}_{2,2}\), from \((a_{i,i} + \alpha_i^{-1} \beta_i B_i) l_{i,i} - d_i > 0\) (Remark 2) and (47), we have
\[
\left| x_{i_i}^* - x_{i_i}^* \right| = \left| \Gamma_{i_i} \left( x_{i_i}^* \right) - \Gamma_{i_i} \left( x_{i_i}^* \right) \right|
= \frac{1}{(a_{i,i} + \alpha_i^{-1} \beta_i B_i) l_{i,i} - d_i}
\times \sum_{j \neq i} a_{i,j} l_{j,i} \left( x_j^* - x_j^* \right)
+ \sum_{j \neq i} a_{i,j} l_{j,i} \left( x_j^* - x_j^* \right)
\leq \sum_{j \neq i} a_{i,j} l_{j,i} \left( x_j^* - x_j^* \right)
\times \left| x_i^* - x_i^* \right| < \left| x_i^* - x_i^* \right|. \tag{51}

If \(i_i \in \mathbb{N}_{2,3} \cup \mathbb{N}_3\), from \(d_{i_i} - (a_{i_i} + \alpha_i^{-1} \beta_i B_i) l_{i,i} > 0\) (Remark 3) and (48), we have
\[
\left| x_{i_i}^* - x_{i_i}^* \right| = \left| \Gamma_{i_i} \left( x_{i_i}^* \right) - \Gamma_{i_i} \left( x_{i_i}^* \right) \right|
= \frac{1}{d_{i_i} - (a_{i_i} + \alpha_i^{-1} \beta_i B_i) l_{i,i}}
\times \sum_{j \neq i} a_{i,j} l_{j,i} \left( x_j^* - x_j^* \right)
+ \sum_{j \neq i} a_{i,j} l_{j,i} \left( x_j^* - x_j^* \right)
\leq \sum_{j \neq i} a_{i,j} l_{j,i} \left( x_j^* - x_j^* \right)
\times \left| x_i^* - x_i^* \right| < \left| x_i^* - x_i^* \right|. \tag{52}

By the above two inequalities, we can deduce that \(x_{i_i}^* = x_{i_i}^*\) for all \(i \in \mathbb{N}_{2,2} \cup \mathbb{N}_{2,3} \cup \mathbb{N}_3\). That is, the fixed point of map \(\Gamma\) is unique. \(\square\)

**Theorem 10.** Suppose that \(\mathbb{N}_1 \cup \mathbb{N}_2 \cup \mathbb{N}_3 \cup \mathbb{N}_4 = \{1, 2, \ldots, N\}\).

(i) If the following conditions hold for all \(i \in \mathbb{N}_{2,2}\), then the other \(3^{N_2,2}-2^{N_2}\) equilibria are unstable.

(ii) Furthermore, if the following conditions hold for all \(i \in \mathbb{N}_{2,2}\), then the other \(3^{N_2,2}-2^{N_2}\) equilibria are unstable.

Proof. In the following, we will discuss the dynamical behaviors of \(3^{N_2}\) equilibria in two cases, respectively.

Case (i). Consider \(\mathbb{N}_{2,2} = \emptyset\). In this case, \(\Omega_\delta\) and \(\Omega_\delta\) can be rewritten as
\[
\Omega_\delta = \prod_{i \in \mathbb{N}_1 \cup \mathbb{N}_3} (-\infty, p_i) \times \prod_{i \in \mathbb{N}_3} [r_i, +\infty)
\times \left[ r_i, q_i \right] \times \prod_{i \in \mathbb{N}_4} [q_i, +\infty),
\]
\[
\Omega_\delta = \prod_{i \in \mathbb{N}_1 \cup \mathbb{N}_3} (-\infty, p_i) \times \prod_{i \in \mathbb{N}_2 \cup \mathbb{N}_3} [r_i, q_i] \times \prod_{i \in \mathbb{N}_4} [q_i, +\infty).
\] (57)

From (53)-(54), we can choose a sufficiently small number \(\eta > 0\) such that
\[
d_{i_i} - a_{i_i} l_{i,i} - \sum_{j \neq i} |a_{i,j}| l_{j,i} - |B_i| > \eta > 0, \tag{58}
\]
\[
\alpha_i - \beta_i l_{i,i} > \eta > 0 \tag{59}
\]
hold for all \(i \in \mathbb{N}_{2,3} \cup \mathbb{N}_3\).

Let \((x(t), S(t))^T \in \Omega_\delta\) be any a solution of system (2) with initial state \((x^T(0), S^T(0))^T \in \Omega_\delta \times \prod_{i=1}^{N} [\alpha_i^{-1} \beta_j \mu_i, \alpha_i^{-1} \beta_j v_i].\) From Lemmas 4–6 and the proof of Theorem 7, we get that there exists some \(T_0 \geq 0\) such that \((x^T(T_0), S^T(T_0))^T \in \Omega_\delta \times \prod_{i=1}^{N} [\alpha_i^{-1} \beta_j \mu_i, \alpha_i^{-1} \beta_j v_i].\) By the positive invariance of
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\(\bar{\Omega}_\delta \times \prod_{i=1}^{N} [\alpha_i^{-1} \beta_i u_i, \alpha_i^{-1} \beta_i v_i]\), we know that the solution 
\((x^T(t), S^T(t)) \in \bar{\Omega}_\delta \times \prod_{i=1}^{N} [\alpha_i^{-1} \beta_i u_i, \alpha_i^{-1} \beta_i v_i]\) for all \(t \geq T_0\), and its dynamics can be described by

\[
\dot{x}_i(t) = -d_i x_i(t) + a_{ii} u_i + \sum_{j \in N_2 \cup N_3, j \neq i} a_{ij} (l_{j,2} x_j(t) + c_{j,2}) \\
+ \sum_{j \in N_2 \cup N_3} a_{ij} u_j + B_i S_i(t) + I_i, \quad i \in \mathbb{N}_1 \cup \mathbb{N}_{2,1},
\]

\[
\dot{x}_i(t) = -d_i x_i(t) + a_{ii} (l_{i,2} x_i(t) + c_{i,2}) \\
+ \sum_{j \in N_2 \cup N_3, j \neq i} a_{ij} u_j + \sum_{j \in N_2 \cup N_3} a_{ij} (l_{j,2} x_j(t) + c_{j,2}) \\
+ \sum_{j \in N_4} a_{ij} u_j + B_i S_i(t) + I_i, \quad i \in \mathbb{N}_4,
\]

\[
\dot{S}_i(t) = -\alpha_i S_i(t) + \beta_i f_i(x_i(t)), \quad i = 1, 2, \ldots, N.
\]

We get from model (60) that

\[
\frac{d}{dt}(x_i(t) - x^*_i) = -d_i (x_i(t) - x^*_i) \\
+ \sum_{j \in N_2 \cup N_3} a_{ij} l_{j,2} (x_j(t) - x^*_j) \\
+ B_i (S_i(t) - S^*_i), \quad i \in \mathbb{N}_1 \cup \mathbb{N}_{2,1} \cup \mathbb{N}_4,
\]

\[
\frac{d}{dt}(x_i(t) - x^*_i) = -(d_i + a_{ii} l_{i,2}) (x_i(t) - x^*_i) \\
+ \sum_{j \in N_2 \cup N_3, j \neq i} a_{ij} l_{j,2} (x_j(t) - x^*_j) \\
+ B_i (S_i(t) - S^*_i), \quad i \in \mathbb{N}_{2,3} \cup \mathbb{N}_3,
\]

\[
\frac{d}{dt}(S_i(t) - S^*_i) = -\alpha_i (S_i(t) - S^*_i), \quad i \in \mathbb{N}_1 \cup \mathbb{N}_{2,1} \cup \mathbb{N}_4,
\]

\[
\frac{d}{dt}(S_i(t) - S^*_i) = -\alpha_i (S_i(t) - S^*_i) \\
+ \beta_i l_{i,2} (x_i(t) - x^*_i), \quad i \in \mathbb{N}_{2,3} \cup \mathbb{N}_3.
\]

Let

\[
\begin{align*}
\varphi_i(t) &= e^{\eta t} |x_i(t) - x^*_i|, \\
\psi_i(t) &= e^{\eta t} |S_i(t) - S^*_i|,
\end{align*}
\]

then by using (61)-(62), we can derive that

\[
D^\top \varphi_i(t) = \eta e^{\eta t} |x_i(t) - x^*_i| + e^{\eta t} \sgn(x_i(t) - x^*_i)
\]

\[
\times \left[ -(d_i + a_{ii} l_{i,2}) (x_i(t) - x^*_i) \\
+ \sum_{j \in N_2 \cup N_3, j \neq i} a_{ij} l_{j,2} (x_j(t) - x^*_j) + B_i (S_i(t) - S^*_i) \right]
\]

\[
\leq (\eta - d_i + a_{ii} l_{i,2}) \varphi_i(t) + \sum_{j \in N_2 \cup N_3, j \neq i} |a_{ij}| l_{j,2} \psi_j(t) + |B_i| \psi_i(t).
\]

Similarly, we have

\[
D^\top \psi_i(t) = \eta e^{\eta t} |S_i(t) - S^*_i| + e^{\eta t} \sgn(S_i(t) - S^*_i)
\]

\[
\times \left[ -\alpha_i (S_i(t) - S^*_i) + \beta_i l_{i,2} (x_i(t) - x^*_i) \right]
\]

\[
\leq (\eta - \alpha_i) \psi_i(t) + \beta_i l_{i,2} \varphi_i(t).
\]

Let \(\sigma > 0\) be an arbitrary real number and

\[
l_0 = (1 + \sigma) e^{\eta T_0} \left( \max_{1 \leq i \leq N} |x_i(T_0) - x^*_i| + \max_{1 \leq i \leq N} |S_i(T_0) - S^*_i| \right);
\]

then

\[
\varphi_i(T_0) = e^{\eta T_0} |x_i(T_0) - x^*_i| \\
\leq e^{\eta T_0} \max_{1 \leq i \leq N} |x_i(T_0) - x^*_i| < l_0, \quad i \in \mathbb{N}_{2,3} \cup \mathbb{N}_3,
\]

\[
\psi_i(T_0) = e^{\eta T_0} |S_i(T_0) - S^*_i| \\
\leq e^{\eta T_0} \max_{1 \leq i \leq N} |S_i(T_0) - S^*_i| < l_0, \quad i \in \mathbb{N}_{2,3} \cup \mathbb{N}_3.
\]

In the following, we will prove that

\[
\varphi_i(t) \leq l_0, \quad \psi_i(t) \leq l_0, \quad t > T_0, \quad i \in \mathbb{N}_{2,3} \cup \mathbb{N}_3.
\]
If (67) is not true, then there exists some \( i \in N_{2,3} \cup N_3 \) and \( t^* > T_0 \) such that either
\[
\varphi_i(t^*) = l_0, \quad D^+ \varphi_i(t^*) > 0,
\]
\[
\varphi_i(t^*) \leq l_0, \quad (T_0 \leq t \leq t^*, \ j \in N_{2,3} \cup N_3),
\]
\[
\psi_j(t^*) \leq l_0, \quad (T_0 \leq t \leq t^*, \ j \in N_{2,3} \cup N_3).
\]

or
\[
\varphi_i(t^*) = l_0, \quad D^+ \varphi_i(t^*) > 0,
\]
\[
\psi_j(t^*) \leq l_0, \quad (T_0 \leq t \leq t^*, \ j \in N_{2,3} \cup N_3),
\]
\[
\psi_j(t^*) \leq l_0, \quad (T_0 \leq t \leq t^*, \ j \in N_{2,3} \cup N_3).
\]

For the first case, it follows from (58) and (63) that
\[
D^+ \varphi_i(t^*) \leq (\eta - d_i + a_{i1,2}) \varphi_i(t^*) + \sum_{j \in N_{2,3} \cup N_3} |a_{ij}| l_{ij,2} \varphi_j(t^*) + |B_i| \psi_i(t^*)
\]
\[
\leq \left[ \eta - d_i + a_{i1,2} + \sum_{j \in N_{2,3} \cup N_3} |a_{ij}| l_{ij,2} + |B_i| \right] l_0 < 0,
\]
which implies that
\[
|x_i(t) - x_i^*| \leq e^{-d_i(t-T_0)} |x_i(T_1) - x_i^*| + \frac{\varepsilon}{d_i} \left( 1 - e^{-d_i(t-T_1)} \right), \quad i \in N_1 \cup N_{2,3} \cup N_4.
\]

That is, \( \lim_{t \to +\infty} |x_i(t) - x_i^*| = 0 \) (i \( \in N_1 \cup N_{2,3} \cup N_4 \)).

To sum up, for all \( i = 1, 2, \ldots, N \), we have \( \lim_{t \to +\infty} |x_i(t) - x_i^*| = 0 \). That is, the equilibrium point \((x^{*T}, S^{*T})^T\) is locally stable in \( \Omega_3 \times \prod_{i=1}^{N_1}[a_i^0, b_i^0, a_i^{-1}, b_i^1] \).

Case (ii). Consider \( N_{2,3} \neq 0 \). In this case, let \((x^T(t), S^T(t))^T\) be a solution of system (2) with initial condition \((x^T(0), S^T(0))^T\) nearby \((x^{*T}, S^{*T})^T\) in \( \Omega_3 \times \prod_{i=1}^{N_1}[a_i^0, b_i^0, a_i^{-1}, b_i^1] \). Without loss of generality, we assume that \((x^T(t), S^T(t))^T \) gets out of region \( \Omega_3 \times \prod_{i=1}^{N_1}[a_i^0, b_i^0, a_i^{-1}, b_i^1] \) for all \( t \geq 0 \). In fact, we can conclude that if the solution \((x^T(t), S^T(t))^T\) is unstable according to the definition of instability. Thus, system (2) can be rewritten as
\[
\dot{x}_i(t) = -d_i x_i(t) + a_{i1} u_j + \sum_{j \in N_{2,3} \cup N_3} a_{ij} u_j
\]
\[
+ \sum_{j \in N_{2,3} \cup N_3} a_{ij} (l_{ij1} x_j(t) + c_{j1})
\]
\[
+ \sum_{j \in N_{2,3} \cup N_3} a_{ij} (l_{ij2} x_j(t) + c_{j2})
\]
\[
+ \sum_{j \in N_{2,3} \cup N_3} a_{ij} (l_{ij3} x_j(t) + c_{j3})
\]
\[
+ B_i S_i(t) + I_i, \quad i \in N_1 \cup N_{2,3} \cup N_4.
\]

Then we have
\[
D^+ \left( |x_i(t) - x_i^*| \right) \leq -d_i |x_i(t) - x_i^*| + \varepsilon, \quad i \in N_1 \cup N_{2,3} \cup N_4,
\]
which implies that
\[
\frac{|x_i(t) - x_i^*|}{t} \leq \frac{\varepsilon}{d_i} \left( 1 - e^{-d_i(t-T_1)} \right), \quad i \in N_1 \cup N_{2,3} \cup N_4.
\]
It follows from model (76) that
\[
\frac{d}{dt}(x_i(t) - x_i^*) = -d_i(x_i(t) - x_i^*) + \sum_{j \in N_{2,1} \cup N_2} a_{ij} \left( f_i(x_j(t)) + c_{ij} \right)
\]
\[
+ \sum_{j \in N_{2,1} \cup N_2} a_{ij} \left( \frac{1}{2}x_j(t) - x_j^* \right) + \sum_{j \in N_2, j \neq i} a_{ij} x_j(t) + B_i S_i(t) + I_i, \quad i \in N_4,
\]
\[
\frac{d}{dt}(S_i(t) - S_i^*) = -\alpha_i (S_i(t) - S_i^*), \quad i \in N_{2,1} \cup N_2, \quad i \neq \text{max}(i),
\]
\[
\frac{d}{dt}(S_i(t) - S_i^*) = -\alpha_i (S_i(t) - S_i^*) + \beta_{i,j} \left( x_j(t) - x_j^* \right), \quad i \in N_{2,3} \cup N_3,
\]
\[
\frac{d}{dt}(S_i(t) - S_i^*) = -\alpha_i (S_i(t) - S_i^*) + \beta_{i,j} \left( x_j(t) - x_j^* \right), \quad i \in N_{2,3} \cup N_3.
\]

Pick an initial state \((x^T(0), S^T(0))^T\) of system (2) such that
\[
\max_{i \leq N} \left( |x_i(0) - x_i^*|, |S_i(0) - S_i^*| \right) = \max_{i \leq N} |x_i(0) - x_i^*| > 0.
\]

(78)

In the following, we claim that
\[
\max_{i \in N_{2,2,1} \cup N_2} \left( |x_i(t) - x_i^*|, |S_i(t) - S_i^*| \right) = \max_{i \in N_{2,2}} |x_i(t) - x_i^*|
\]

holds for all \(t \geq 0\). In fact, if (79) is not true, then there exists some \(t_0 \geq 0\) such that
\[
\max_{i \in N_{2,2}} \left( |x_i(t_0) - x_i^*|, \max_{i \in N_{2,2 \cup N_2 \cup N_3}} |S_i(t_0) - S_i^*| \right) = \max_{i \in N_{2,2}} |x_i(t_0) - x_i^*|;
\]

(80)

then there exists an index \(i_0\) such that one of the following three cases holds:

(a) \(i_0 \in N_{2,2} \cup N_3\) and \(|x_{i_0}(t_0) - x_{i_0}^*| = \max_{i \in N_{2,2} \cup N_3} |x_i(t_0) - x_i^*|\),

(b) \(i_0 \in N_{2,2,1} \cup N_2\) and \(|S_{i_0}(t_0) - S_{i_0}^*| = \max_{i \in N_{2,2} \cup N_3} |x_i(t_0) - x_i^*|\),

(c) \(i_0 \in N_{2,2,1} \cup N_2\) and \(|S_{i_0}(t_0) - S_{i_0}^*| = \max_{i \in N_{2,2} \cup N_3} |x_i(t_0) - x_i^*|\).

For the case (a), it follows from (53) and (77) that
\[
D^+ \left( |x_{i_0}(t_0) - x_{i_0}^*| \right) \leq \left( -d_{i_0} + a_{i_0,1} \right) |x_{i_0}(t_0) - x_{i_0}^*| + \sum_{j \in N_{2,2}} |a_{i_0,j}| |l_{i,j} - x_j^*| + \sum_{j \in N_{2,2} \cup N_2} |a_{i_0,j}| |l_{i,j} - x_j^*| + |B_{i_0}| |S_{i_0}(t_0) - S_{i_0}^*| \leq 0.
\]

(81)

For the case (b), by using (54) and (77), we get that
\[
D^+ \left( |S_{i_0}(t_0) - S_{i_0}^*| \right) \leq \left( -d_{i_0} + a_{i_0,1} \right) |S_{i_0}(t_0) - S_{i_0}^*| < 0.
\]

(82)

Similarly, for the case (c), by means of (56) and (77), we have
\[
D^+ \left( |S_{i_0}(t_0) - S_{i_0}^*| \right) \leq \left( -d_{i_0} + a_{i_0,1} \right) |S_{i_0}(t_0) - S_{i_0}^*| < 0.
\]

(83)
Meanwhile, denote $i_1 \in \mathbb{N}_{2,2}$ such that $|x_{i_1}(t_0) - x^*_i| = \max_{i \in \mathbb{N}_{2,2}} |x_i(t_0) - x^*_i|$; then we obtain from (55) and (77) that
\[
D^+ \left( |x_{i_1}(t_0) - x^*_i| \right) \geq \left( -d_{i_1} + a_{i_1,i_1,i_1} \right) |x_{i_1}(t_0) - x^*_i| - \sum_{j \neq i_1} \left| a_{i_1,j} \right| |j_1(t_0) - x^*_j| - \sum_{j \in \mathbb{N}_{2,2} \cup \mathbb{N}_i} \left| a_{i_1,j} \right| |j_2(t_0) - x^*_j| - |B_i||S_{i_1}(t_0) - S^*_i| \geq -(d_{i_1} + a_{i_1,i_1,i_1}) |x_{i_1}(t_0) - x^*_i| - \sum_{j \neq i_1} \left| a_{i_1,j} \right| |j_1(t_0) - x^*_j|,
\]
(84)
Thus, combining inequalities (79) and (84) together gives
\[
\max_{i \in \mathbb{N}_{2,2}} \{ |x_i(t) - x^*_i|, |S_i(t) - S^*_i| \} = \max_{i \in \mathbb{N}_{2,2}} |x_i(t) - x^*_i| \geq \max_{i \in \mathbb{N}_{2,2}} |x_i(0) - x^*_i| > 0
\]
which implies that
\[
\max_{i \in \mathbb{N}_{2,2} \cup \mathbb{N}_i} |x_i(t) - x^*_i| = \max_{i \in \mathbb{N}_{2,2}} |x_i(t) - x^*_i| \geq \max_{i \in \mathbb{N}_{2,2}} |x_i(0) - x^*_i|
\]
holds for all $t \geq 0$.

Therefore, there must exist an index $i_2 \in \mathbb{N}_{2,2}$ and an increasing time sequence $\{t_k\}_{k=1}^{\infty}$ with $\lim_{k \to \infty} t_k = +\infty$ such that $|x_{i_2}(t_k) - x^*_i| \geq \max_{i \in \mathbb{N}_{2,2} \cup \mathbb{N}_i} |x_i(t_0) - x^*_i|$. That is, $x_{i_2}(t)$ would not converge to $x^*_i$ when $t$ tends to $+\infty$. In other words, the equilibrium point $x^*$ is unstable.

In the following, we will consider the dynamical behaviors of system (2) with activation functions (4). Note that $\max_{j \neq i,j} |a_{ij}|$ and $\min_{j \neq i,j} |a_{ij}| = -|a_{ij}|$; in this case, $\mathbb{N}_1 - \mathbb{N}_4$ reduce to the following index subsets:
\[
\mathbb{N}_1 = \left\{ i : I_i < -d_i + a_{i_1} - \sum_{j \neq i_1} |a_{ij}| - \alpha^{-1}_i \beta_i |B_i| \right\},
\]
\[
\mathbb{N}_2 = \left\{ i : |I_i| < -d_i + a_{i_1} - \sum_{j \neq i_1} |a_{ij}| - \alpha^{-1}_i \beta_i |B_i| \right\},
\]
\[
\mathbb{N}_3 = \left\{ i : -d_i + a_{i_1} + \sum_{j \neq i_1} |a_{ij}| + \alpha^{-1}_i \beta_i |B_i| < I_i < 3d_i + a_{i_1} - \sum_{j \neq i_1} |a_{ij}| - \alpha^{-1}_i \beta_i |B_i| \right\},
\]
\[
\mathbb{N}_4 = \left\{ i : I_i > \max \left\{ 3d_i + a_{i_1}, d_i - a_{i_1} \right\} + \sum_{j \neq i_1} |a_{ij}| + \alpha^{-1}_i \beta_i |B_i| \right\},
\]
(87)
Meanwhile, conditions (47)-(48) become
\[
\sum_{j \in \mathbb{N}_{2,2}} |a_{ij}| + \sum_{j \in \mathbb{N}_{2,2} \cup \mathbb{N}_i} |a_{ij}| < -d_i + a_{i_1} + \alpha^{-1}_i \beta_i |B_i|,
\]
\[
\sum_{j \in \mathbb{N}_{2,2}} |a_{ij}| + \sum_{j \in \mathbb{N}_{2,2} \cup \mathbb{N}_i} |a_{ij}| < d_i + a_{i_1} + \alpha^{-1}_i \beta_i |B_i|,
\]
(88)
holds for all $t \geq 0$.

Remark II. Under index subsets $\mathbb{N}_2, \mathbb{N}_3$, conditions (88)-(89) always hold.

Proof. From the definition of index subset $\mathbb{N}_2$, we get
\[
\sum_{j \neq i_1,j} |a_{ij}| < -d_i + a_{i_1} + \alpha^{-1}_i \beta_i |B_i| < d_i + a_{i_1} + \alpha^{-1}_i \beta_i |B_i|.
\]
(90)
Therefore, inequalities (88) and (89) hold for $i \in \mathbb{N}_{2,2}$ and $i \in \mathbb{N}_{2,3}$, respectively.
Figure 3: Transient behavior of $x_2$ in Example 1.

Figure 4: Transient behavior of $S_1$ and $S_2$ in Example 1.

Figure 5: Phase plot of state variable $(x_1, x_2, S_1)\,^T$ in Example 1.

Figure 6: Phase plot of state variable $(x_1, x_2, S_2)\,^T$ in Example 1.

Figure 7: Transient behavior of $x_1$ and $x_2$ near the equilibrium point $(-201/80, -5/8, -1/2, -5/16)\,^T$ in Example 1.

Figure 8: Transient behavior of $x_1$ and $x_2$ near the equilibrium point $(5/12, -241/120, 5/24, -1/2)\,^T$ in Example 1.
Similarly, it follows from the definition of $\overline{N}_3$ that
\[
d_i - a_{ii} + \sum_{j \neq i, j=1}^{N} \left|a_{ij}\right| - \alpha_i^{-1}\beta_i B_i < 3d_i + a_{ii} - \sum_{j \neq i, j=1}^{N} \left|a_{ij}\right| + \alpha_i^{-1}\beta_i B_i,
\]  
which implies that
\[
\sum_{j \neq i, j=1}^{N} \left|a_{ij}\right| < d_i + a_{ii} + \alpha_i^{-1}\beta_i B_i. \tag{92}
\]
Therefore, condition (89) holds for $i \in N_3$.

From Theorem 10 and Remark 11, we can obtain Corollary 12 as follows.
Corollary 12. Suppose that $\mathbb{N}_1 \cup \mathbb{N}_2 \cup \mathbb{N}_3 \cup \mathbb{N}_4 = \{1, 2, \ldots, N\}$. If the following conditions hold, then system (2) with activation functions (4) has exactly $3^{|\mathbb{N}_2|}$ equilibria, $2^{|\mathbb{N}_2|}$ of them are locally stable and others are unstable.

Proof. First of all, according to Theorem 7 and Remark 11, the exact existence of $3^{|\mathbb{N}_2|}$ equilibria for system (2) with activation functions (4) can be guaranteed under condition $\mathbb{N}_1 \cup \mathbb{N}_2 \cup \mathbb{N}_3 \cup \mathbb{N}_4 = \{1, 2, \ldots, N\}$. In addition, it is easy to see that conditions (93)–(95) imply the conditions (53)–(56) with $l_{i,1} = 1$, $l_{i,2} = -1$. From Theorem 10, we derive the result of Corollary 12. \hfill $\Box$

Let $B_i = \beta_i = S_i(t) = 0$ $(i = 1, 2, \ldots, N)$; then system (2) is transformed into the following neural networks which are investigated in [33]:

$$x_i(t) = -d_i x_i(t) + \sum_{j=1}^{N} a_{ij} f_j(x_j(t)) + I_i. \quad (96)$$
In this case, $\mathbb{N}_1 - \mathbb{N}_4$ turn into the following index subsets defined in [33]:

\[
\begin{align*}
\mathbb{N}_1 &= \left\{ i : I_i < -|d_i + a_{ii}| - \sum_{j \neq i, j = 1}^N |a_{ij}| \right\}, \\
\mathbb{N}_2 &= \left\{ i : |I_i| < -d_i + a_i - \sum_{j \neq i, j = 1}^N |a_{ij}| \right\}, \\
\mathbb{N}_3 &= \left\{ i : -d_i + a_i + \sum_{j \neq i, j = 1}^N |a_{ij}| < I_i \right\}, \\
\mathbb{N}_4 &= \left\{ i : I_i > \max \{3d_i + a_i, a_{ii} - 3d_i - a_i - \sum_{j \neq i, j = 1}^N |a_{ij}| \} \right\}.
\end{align*}
\]

Applying Corollary 12, we can obtain easily Corollary 13 as follows.

**Corollary 13.** Suppose that $\mathbb{N}_1 \cup \mathbb{N}_2 \cup \mathbb{N}_3 \cup \mathbb{N}_4 = \{1, 2, \ldots, N\}$ holds. Then system (96) with activation functions (4) has exactly $3^{\mathbb{N}_3}$ equilibria, $2^{\mathbb{N}_2}$ of them are locally stable and others are unstable.

**Proof.** First of all, it follows from the definition of index subset $\mathbb{N}_2$ that (93)-(94) with $B_i = 0$ hold for $i \in \mathbb{N}_{2,2}$ and $i \in \mathbb{N}_{4,3}$, respectively. From the definition of $\mathbb{N}_j$, we have

\[
\begin{align*}
d_i - a_{ii} + \sum_{j \neq i, j = 1}^N |a_{ij}| &< 3d_i + a_i - \sum_{j \neq i, j = 1}^N |a_{ij}|,
\end{align*}
\]

which implies that

\[
\begin{align*}
d_i + a_i - \sum_{j \neq i, j = 1}^N |a_{ij}| &> 0;
\end{align*}
\]

that is, inequality (94) with $B_i = 0$ holds for $i \in \mathbb{N}_{4,3}$. Inequality (95) is obvious, due to $a_{ii} > 0$ and $\beta_i = 0$. From Corollary 12, we derive the result of Corollary 13. \hfill $\Box$

**Remark 14.** In this paper, we study the multistability and instability of CNNs with activation functions (3). The models are different and more general than those in [33], and the considered activation functions (3) are also more general than those employed in [33]. Moreover, the index subsets defined in this paper are less restrictive than those defined in [33].

**Remark 15.** Compared with the results reported in [33], it can be seen that Corollary 13 above is consistent with Theorem 1 in [33]. That is, if we specialize the system and activation functions in Theorem 10 to those considered in [33], we can obtain the main result in [33]. Therefore, Theorem 10 extends and improves the main result in [33].

### 3. Two Illustrative Examples

For convenience, we consider the following two-dimensional CNNs:

\[
\begin{align*}
\dot{x}_i(t) &= -d_i x_i(t) + \sum_{j=1}^2 a_{ij} f_j(x_j(t)) + B_i S_i(t) + I_i, \\
\dot{S}_i(t) &= -\alpha_i S_i(t) + \beta_i f_i(x_i(t)), \quad i = 1, 2.
\end{align*}
\]

**Example 1.** For system (100), take $d_1 = d_2 = 1$, $a_{11} = a_{22} = 2$, $a_{12} = 0.5$, $a_{21} = 0.5$, $B_1 = 0.4$, $B_2 = -0.4$, $I_1 = I_2 = 0$, $\alpha_1 = \alpha_2 = 2$, $\beta_1 = \beta_2 = 1$, and

\[
\begin{align*}
f_i(x) &= \begin{cases} 
-1, & -\infty < x < -1, \\
-x, & -1 \leq x \leq 1, \\
-x + 2, & 1 < x \leq 3, \\
-1, & 3 < x < +\infty.
\end{cases} \quad (i = 1, 2).
\end{align*}
\]

It is easy to see that $\mathbb{N}_2 = \{1, 2\}$. In addition, by simple computations, we have

\[
\begin{align*}
-d_1 + a_{11} - |a_{12}| - |B_1| = 0.1 > 0, \\
-d_2 + a_{22} - |a_{21}| - |B_2| = 0.1 > 0.
\end{align*}
\]

Therefore, the conditions in Corollary 12 hold. According to Corollary 12, system (100) has exactly $3^2 = 9$ equilibria, $2^2 = 4$ equilibria are locally stable and others are unstable. In fact, by direct computations, we can obtain the nine equilibria $(-2.7, -1.3, -0.5, -0.5)^T$, $(-201/80, -5/8, -1/2, -5/16)^T$, $(-541/280, 41/28, -1/2, 15/56)^T$, $(5/12, -241/120, 5/24, -1/2)^T$, $(0, 0, 0, 0)^T$, $(-100/361, 482/361, -50/361, 120/361)^T$, $(39/32, -701/320, 25/64, -1/2)^T$, $(402/281, 100/281, 80/281, 50/281)^T$, and $(1385/921, 1102/921, 457/1842, 185/921)^T$. From Figures 2, 3, 4, 5, and 6, it can be seen that the four equilibria $(-2.7, -1.3, -0.5, -0.5)^T$, $(-541/280, 41/28, -1/2, 15/56)^T$, $(39/32, -701/320, 25/64, -1/2)^T$, and $(1385/921, 1102/921, 457/1842, 185/921)^T$ are locally stable. Figures 7, 8, 9, 10, and 11 confirm that the others are unstable.

**Example 2.** For system (100), take $d_1 = 1$, $a_{11} = -1$, $a_{12} = -0.2$, $B_1 = 0.2$, $d_2 = a_{22} = 2$, $a_{21} = 1/3$, $B_2 = -0.2$, $I_1 = 9$, $I_2 = 0$, $\alpha_1 = \alpha_2 = 3$, $\beta_1 = \beta_2 = 1$, and

\[
\begin{align*}
f_i(x) &= \begin{cases} 
-3, & -\infty < x < -2, \\
2x + 1, & -2 \leq x \leq 2, \\
-x + 7, & 2 < x \leq 10, \\
-3, & 10 < x < +\infty.
\end{cases} \quad (i = 1, 2).
\end{align*}
\]
It is easy to see that \( N_2 = \{2\}, N_4 = \{1\} \). Herein, the parameters satisfy conditions in Remark 9 and Theorem 10:

\[
|a_{21}| \max \{[l_{1,1}], [l_{1,2}]\} = \frac{2}{3} < \frac{28}{15} = \min \left\{ \left( a_{22} + a_2^{-1} B_1 \beta_2 \right) l_{2,1} - d_2, \right. \\
\left. d_2 - \left( a_{22} + a_2^{-1} B_2 \beta_2 \right) l_{2,2} \right\}, \quad (104) \\
\begin{align*}
&d_2 - a_{22} l_{2,2} - |a_{21}| \max \{[l_{1,1}], [l_{1,2}]\} = \frac{10}{3} > 0, \\
&-d_2 + a_{22} l_{2,1} - |a_{21}| \max \{[l_{1,1}], [l_{1,2}]\} = \frac{4}{3} > 0.
\end{align*}
\]

From Remark 9 and Theorem 10, it follows that the system has exactly three equilibria \((62/5, -17/5, -1, -1)^T\), \((59/5, -1/2, -1, 0)^T\), and \((3256/295, 188/59, -1, 75/59)^T\); the first and the third equilibria are locally stable, while the second equilibrium point is unstable. From Figures 12, 13, 14, 15, and 16, it can be seen that the two equilibria \((62/5, -17/5, -1, -1)^T\) and \((3256/295, 188/59, -1, 75/59)^T\) are locally stable. Figure 17 confirms that the equilibrium point \((59/5, -1/2, -1, 0)^T\) is unstable.

4. Conclusions

In this paper, the multistability and instability issues have been studied for CNNs with Mexican-hat-type activation functions. We showed that under some conditions, the system has \(3^{N_2}\) equilibria, \(2^{N_4}\) of them are locally stable and others are unstable. Two examples with their computer simulations were given to illustrate the effectiveness of the obtained results. Some thorough analyses are needed further. Here, we only treated neural networks without time delay, how about when time delays are presented? This needs to be investigated in the future.

Conflict of Interests

The authors declare that they have no conflict of interests regarding the publication of this paper.

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