Invariant Subspace Approach to Boolean (Control) Networks

Daizhan Cheng *,†, Lijun Zhang †, Dongyao Bi ‡

* Center of STP Theory and Applications, LiaoCheng University, LiaoCheng 252000, P.R.China
† Institute of Systems Science, Chinese Academy of Sciences, Beijing 100190, P.R.China
‡ Northwestern Polytechnical University

E-mail: dcheng@iss.ac.cn

Abstract: A logical function can be used to characterizing a property of a state of Boolean network (BN), which is considered as an aggregation of states. To illustrate the dynamics of a set of logical functions, which characterize our concerned properties of a BN, the invariant subspace containing the set of logical functions is proposed, and its properties are investigated. Then the invariant subspace of Boolean control network (BCN) is also proposed. The dynamics of invariant subspace of BCN is also invariant. Finally, using outputs as the set of logical functions, the minimum realization of BCN is proposed, which provides a possible solution to overcome the computational complexity of large scale BNs/BCNs.

Keywords: Boolean (control) network, logical function, invariant subspace, minimum realization, semi-tensor product.

1 Introduction

The BN was firstly proposed by Kauffman in 1969 [7]. It has been proved to be a very efficient way for modeling and analyzing genetic regulatory network. Recently, motivated by the semi-tensor product (STP) of matrices, the investigation of BN and BCN becomes a heat research direction in control community. Nowadays, the STP approach becomes the mainstream in studying BNs and BCNs. We refer to some survey papers for its current development in theory and applications [5], [10], [11], [9].

The major obstacle in applications of STP approach to BNs and BCNs is the computational complexity. It is well known that BN structure analysis and BCN control design and many related problems are NP hard problem [15], while BNs from gene regulatory networks are usually of large scale. For a network with $n$ nodes, the state space of BN or BCN in STP model is of $2^n$ states. Hence, in general, the STP approach can only handle $n < 20$ cases or so.

A proper tool in dealing with large-scale BN (BCN) is aggregation [13, 14, 12]. To the authors’ best knowledge, the aggregation proposed so far is based on the structure of networks. This method has some weaknesses. First, it requires the knowledge on the structure of networks. It is not an easy job to get the structure of a large scale network. Second, such an aggregation does not represent certain properties of the nodes. Sometimes, classifying nodes according to their various properties is more important then their positions.

*Supported partly by NNSF 62073315 of China, and China Postdoctoral Science Foundation 2020TQ0184.
Support vector machine approach is a very powerful tool in aggregation, where it is called pattern recognition [1, 6]. In support vector machine a hyperplane $w^T x + b$, which separates points into two groups: $\{ x \mid w^T x + b > 0 \}$ and $\{ x \mid w^T x + b < 0, (\text{refer to Fig. 1}) \}$.

This paper uses the idea of support vector machine to aggregation of nodes in a large-scale Boolean network. A logical function $g(x)$ is considered as a support vector, which classifies nodes into two groups: $\{ x \mid g(x) = 1 \}$ and $\{ x \mid g(x) = 0, (\text{refer to Fig. 2}) \}$.

Several logical functions, which form a set of support vectors for various properties, become a subspace. Using this subspace, we may construct a logical dynamic system, which describes the dynamics of aggregated classis. Since this dynamic system might be much smaller than the original one, the computational complexity could be reduced a lot.

Roughly speaking, the idea for the approach in this paper is as follows: First, some logical functions
are chosen to characterize some properties of a BN/BCN, concerned by us. Then the smallest subspaces containing the set of logical functions, which is invariant under the dynamic evolution. The dynamic equation for the subspace is revealed, which completely described the evolution of the concerned logical variables, which correspond the set of logical functions. Finally, the outputs of a BCN are considered as the set of concerned logical functions, which lead to a minimum realization of the original BCN.

The rest of this paper is follows: Section 2 presents some preliminaries as follows: (i) STP of matrices, which is the fundamental tool for our approach; (ii) Matrix expression of BN and BCN, which is called the algebraic state space representation (ASSR). Section 3 presents the separating subspace approach for BNs. The separating logical functions and the invariant subspace containing the set of logical functions is constructed, and its properties are investigated. Finally, the dynamic equation is obtained for the invariant subspace. The invariant subspace and its dynamic equation of BCN are considered in Section 4. Section 5 considers the minimum realization of a BCN. Their dynamic equations are also revealed. Section 6 is a brief conclusion.

2 Preliminaries

2.1 STP of Matrices

Definition 2.1. [3, 4]: Let \( M \in \mathcal{M}_{m \times n} \), \( N \in \mathcal{M}_{p \times q} \), and \( t = \text{lcm}\{n, p\} \) be the least common multiple of \( n \) and \( p \). The semi-tensor product (STP) of \( M \) and \( N \), denoted by \( M \times N \), is defined as

\[
(M \otimes I_{t/n}) (N \otimes I_{t/p}) \in \mathcal{M}_{mt/n \times qt/p},
\]

where \( \otimes \) is the Kronecker product.

Note that when \( n = p \), \( M \times N = MN \). That is, the semi-tensor product is a generalization of conventional matrix product. Moreover, it keeps all the properties of conventional matrix product available [4]. Hence we can omit the symbol \( \times \). Throughout this paper the matrix product is assumed to be STP, and the symbol \( \times \) is mostly omitted.

The following are some basic properties:

Proposition 2.2. 1. (Associative Law)

\[
(F \times G) \times H = F \times (G \times H).
\]

2. (Distributive Law)

\[
\begin{aligned}
F \times (aG \pm bH) &= aF \times G \pm bF \times H, \\
(aF \pm bG) \times H &= aF \times H \pm bG \times H,
\end{aligned}
\]

\( a, b \in \mathbb{R} \).

Proposition 2.3. 1. Let \( X \in \mathbb{R}^m \), \( Y \in \mathbb{R}^n \) be two columns. Then

\[
X \times Y = X \otimes Y.
\]

2. Let \( \omega \in \mathbb{R}^m \), \( \sigma \in \mathbb{R}^n \) be two rows. Then

\[
\omega \times \sigma = \sigma \otimes \omega.
\]
About the transpose, we have

**Proposition 2.4.**

\[(A \times B)^T = B^T \times A^T.\]  

\[\tag{6}\]

About the inverse, we have

**Proposition 2.5.** Assume \(A\) and \(B\) are invertible, then

\[(A \times B)^{-1} = B^{-1} \times A^{-1}.\]  

\[\tag{7}\]

The following property is for STP only.

**Proposition 2.6.** Let \(X \in \mathbb{R}^m\) be a column and \(M\) a matrix. Then

\[X \times M = (I_m \otimes M) \times X.\]  

\[\tag{8}\]

**Definition 2.7.** A matrix \(W_{[m, n]} \in M_{mn \times mn}\), defined by

\[W_{[m, n]} := [I_n \otimes \delta^1_m, I_n \otimes \delta^2_m, \ldots, I_n \otimes \delta^m_m, \ldots]\]  

\[\tag{9}\]

is called the \((m, n)\)-th dimensional swap matrix, where \(\delta^j_m\) is the \(j\)-th column of \(I_m\).

The basic function of the swap matrix is to “swap” two vectors. That is,

**Proposition 2.8.** Let \(X \in \mathbb{R}^m\) and \(Y \in \mathbb{R}^n\) be two columns. Then

\[W_{[m, n]} \times X \times Y = Y \times X.\]  

\[\tag{10}\]

**Definition 2.9.** Let \(A \in M_{p \times n}\) and \(B \in M_{q \times n}\). Then the Khatri-Rao Product of \(A\) and \(B\) is

\[A \ast B = [\text{Col}_1(A) \times \text{Col}_1(B), \ldots, \text{Col}_n(A) \times \text{Col}_n(B)] \in M_{pq \times n}.\]  

\[\tag{11}\]

### 2.2 Matrix Expression of BN

**Definition 2.10.** A BN is described by

\[
\begin{align*}
  x_1(t + 1) &= f_1(x_1(t), \ldots, x_n(t)), \\
  x_2(t + 1) &= f_2(x_1(t), \ldots, x_n(t)), \\
  \vdots \\
  x_n(t + 1) &= f_n(x_1(t), \ldots, x_n(t)),
\end{align*}
\]

\[\tag{12}\]

where \(x_i(t) \in \mathcal{D} = \{0, 1\}, f_i : \mathcal{D}^n \rightarrow \mathcal{D}, i = 1, 2, \ldots, n\) are logical functions.

Using vector form expression: \(1 \sim \delta^1_2 = (1, 0)^T, 0 \sim \delta^2_2 = (0, 1)^T\). Then \(x(t)\) can be expressed as \(x(t) \in \Delta_2\), where \(\Delta_k\) is the set of columns of \(I_k\).

A matrix \(M \in M_{p \times q}\) is called a logical matrix, if \(\text{Col}(M) \subset \Delta_p\). The set of \(p \times q\) dimensional logical matrices is denoted by \(L_{p \times q}\).

Then a BN has its matrix form, called the ASSR of BN, as follows:
Proposition 2.11.  
(i) For a logical function \( f : \mathcal{D}^n \to \mathcal{D} \), there exists a unique logical matrix \( M_f \in \mathcal{L}_{2 \times 2^n} \) such that in vector form

\[
f(x_1, x_2, \cdots, x_n) = M_f \kappa_{i=1}^n x_i. \tag{13}
\]

(ii) Let \( M_i \) be the structure matrix of \( f_i, i = 1, 2, \cdots, n \). Then in vector form \((12)\) can be expressed into its componentwise ASSR as

\[
\begin{align*}
x_1(t+1) &= M_1 \kappa_{i=1}^n x_i(t), \\
x_2(t+1) &= M_2 \kappa_{i=1}^n x_i(t), \\
&\vdots \\
x_n(t+1) &= M_n \kappa_{i=1}^n x_i(t).
\end{align*} \tag{14}
\]

(iii) Setting \( x(t) = \kappa_{i=1}^n x_i(t) \), \((14)\) can further be expressed into its ASSR as

\[
x(t+1) = Mx(t), \tag{15}
\]

where

\[
M = M_1 \ast M_2 \ast \cdots \ast M_n \in \mathcal{L}_{2^n \times 2^n}
\]

is called the structure matrix of BN \((12)\).

Similarly, the BCN is described as follows:

\[
\begin{align*}
x_1(t+1) &= f_1(x_1(t), \cdots, x_n(t); u_1(t), \cdots, u_m(t)), \\
x_2(t+1) &= f_2(x_1(t), \cdots, x_n(t); u_1(t), \cdots, u_m(t)), \\
&\vdots \\
x_n(t+1) &= f_n(x_1(t), \cdots, x_n(t); u_1(t), \cdots, u_m(t)),
\end{align*} \tag{16}
\]

where \( u_j(t) \in \mathcal{D}, j = 1, \cdots, m \) are controls.

We also have similar algebraic expressions for BCN.

Proposition 2.12. Consider BCN \((10)\).

(i) Its componentwise ASSR is

\[
\begin{align*}
x_1(t+1) &= L_1 u(t)x(t), \\
x_2(t+1) &= L_2 u(t)x(t), \\
&\vdots \\
x_n(t+1) &= L_n u(t)x(t),
\end{align*} \tag{17}
\]

where \( u(t) = \kappa_{j=1}^m u_j(t) \), \( L_i \in \mathcal{L}_{2 \times 2^{m+n}} \) is the structure matrix of \( f_i, i = 1, \cdots, n \).

(ii) Its ASSR is

\[
x(t+1) = Lu(t)x(t), \tag{18}
\]

where \( L = L_1 \ast L_2 \ast \cdots \ast L_n \in \mathcal{L}_{2^n \times 2^{m+n}} \) is the structure matrix of BCN \((10)\).
Definition 2.13. [2] Consider BN (12) or BCN (13).

(i) Their state space, denoted by $X$, is defined as the set of all logical functions of $x_1, x_2, \cdots, x_n$, denoted by $\mathcal{F}_\ell\{x_1, x_2, \cdots, x_n\}$. That is,

$$X = \mathcal{F}_\ell\{x_1, x_2, \cdots, x_n\}. \quad (19)$$

(ii) Let $z_1, z_2, \cdots, z_r \in X$. Then the subspace generated by $z_1, z_2, \cdots, z_r$, is defined by

$$Z = \mathcal{F}_\ell\{z_1, z_2, \cdots, z_r\}. \quad (20)$$

Consider (20). Since $z_i \in X$, there is a structure matrix of $z_i$, denoted by $G_i$, such that in vector form we have

$$z_i = G_i x, \quad i = 1, \cdots, r. \quad (21)$$

where $x = \kappa_{i=1}^n x_i, G_i \in \mathcal{L}_{2 \times 2^n}$. Denote $z = \kappa_{i=1}^r z_i$, then we have

$$z = Tx, \quad (22)$$

where $T = G_1 \ast G_2 \ast \cdots \ast G_r \in \mathcal{L}_{2r \times 2^n}$.

Definition 2.14. Support $Z = \mathcal{F}_\ell\{z_1, z_2, \cdots, z_n\}$ has its algebraic expression (22) with non-singular $T$, then $X \to Z$ is called a coordinate change.

Remark 2.15. Since $T$ is a logical matrix, if $T$ is non-singular, then it is a permutation matrix. Hence $T^{-1} = T^T$. Now if $f \in X$, which can be expressed via its structure matrix $M_f$ as

$$f(x) = M_f x,$$

then it can also be expressed as

$$f(x) = \tilde{f}(z) = M_f T^T z.$$

3 Separating Subspace Approach to BN

3.1 Separating Function and Invariant Subspace

Note that a logical function $f(x_1, x_2, \cdots, x_n)$ can be considered as an index function of a subset of node set $N := \{x_1, x_2, \cdots, x_n\}$. Given an $S \subset N$, then its index function, denoted by $f_S$, can be defined as follows:

$$f_S(x) := \begin{cases} 
1, & x \in S, \\
0, & \text{Otherwise.}
\end{cases} \quad (23)$$

Let $\pi : 2^N \to \mathcal{F}_\ell\{x_1, x_2, \cdots, x_n\}$ determined by $\pi(S) = f_s$, which is defined by (23). Then it is obvious that $\pi$ is bijective. Based on this observation we can define separating logical function.

For a large-scale BN, if there are $n$ nodes, its number of states is $2^n$. Say, $n = 32$, then the states are $4.295E + 9$. So in its ASSR, the transition matrix, which has $2^n \times 2^n$ dimension, is not practically
computable. In fact, we may not be interested in its detailed state evolution. We are only interested in some particular properties of the BN. Using the idea of separating logical function approach, the set of separating logical functions of BN is proposed as follows: Assume we are interested in a property, say \( p \), for a BN. We may define a logical function \( z_p \) as follows:

\[
z_p(x) = \begin{cases} 
1, & \text{if } p \text{ is true at } x, \\
0, & \text{if } p \text{ is false at } x,
\end{cases}
\]

\( z_p \in X \). Such \( z_p \) is called a separating function, which classifies all states into two groups according to property \( p \).

In general, we are interested in a set of \( z_i(x), i = 1, 2, \ldots, r \), where \( r \ll n \). We then can aggregate \( x \) into \( 2^r \) groups as

\[
x^k := \{ x \| z(x) = \delta^k \}, \quad k = 1, \ldots, 2^r.
\]

**Definition 3.1.** Let \( z_i, i = 1, \ldots, r \) be a set of separating logical functions. Then \( \{ z_i \mid i = 1, 2, \ldots, r \} \) are called aggregating variables, and

\[
Z := F_\ell \{ z_1, z_2, \ldots, z_r \}
\]

is called the \( \{ z_i \mid i = 1, 2, \ldots, r \} \) aggregated subspace.

Support we are only interested the dynamics about \( Z \), which might be much more smaller than the original BN.

To find the dynamics of \( Z \), we need some new concepts.

**Definition 3.2.** Given BN (12).

(i) \( Z^1 = F_\ell \{ z^1 \} = F_\ell \{ z_1^1, z_2^1, \ldots, z_r^1 \} \) is called a regular subspace, if there exist \( z^2 = (z_1^2, z_2^2, \ldots, z_{n-r}^2) \), such that \( z = (z_1^1, \ldots, z_1^r, z_2^1, \ldots, z_{n-r}^2) \) is another coordinate frame. That is, \( X = Z = F_\ell \{ z^1, z^2 \} \).

(ii) Assume \( Z^1 \) is a regular subspace and \( z = (z^1, z^2) \) is a new coordinate frame. Moreover, under \( z \), (12) can be expressed as

\[
\begin{cases} 
z^1(t + 1) = \hat{F}^1(z^1(t)), & z^1(t) \in Z^1 \\
z^2(t + 1) = \hat{F}^2(z(t)), & z^2(t) \in Z^2, z(t) \in Z,
\end{cases}
\]

then \( Z^1 \) is called an \( M \)-invariant subspace.

Recall the ASSR (15) of (12). We have the following result:

**Theorem 3.3.** Consider BN (12) with its ASSR (17). Suppose \( Z^1 = F_\ell \{ z_1^1, z_2^1, \ldots, z_r^1 \} \) is a regular space with its ASSR as

\[
z^1 = Qx,
\]

where \( z^1 = \sum_{i=1}^r z_i^1, Q \in L_{2^r \times 2n} \). Then, \( Z^1 \) is an \( M \)-invariant subspace of (12), if and only if, there exists \( H \in L_{2^r \times 2^r} \) such that

\[
QM = HQ.
\]
Proof. (sufficiency) Since $Z^1$ is a regular subspace, there exists $z^2 = (z_1^2, z_2^2, \ldots, z_{n-r}^2)$, such that $z = (z^1, z^2)$ is a new coordinate frame. Hence,

$$z^1(t + 1) = Qx(t + 1) = QMx(t) = HQx(t) = Hz^1(t).$$  \hfill (28)

(28) shows that under coordinates $z$ the BN has the form of (25).

(necessity) Assume under coordinate frame $z$ the BN (12) has the form of (25). Moreover, assume the structure matrix of $\tilde{F}_1$ is $\tilde{M}_1 \in L_{2^k \times 2^k}$. Then

$$z^1(t + 1) = \tilde{M}_1 z^1(t) = \tilde{M}_1 Qx(t).$$

On the other hand,

$$z^1(t + 1) = Qx(t + 1) = QMx(t).$$

Since $x(t)$ is arbitrary, we have

$$QM = \tilde{M}_1 Q.$$

Set $H = \tilde{M}_1$, (28) follows. \hfill \Box

Remark 3.4. According to 3.3, to verify whether a regular subspace is an invariant subspace we have to check whether equation (28) has solution $H$. Since $Z^1$ is a regular subspace, its structure matrix $Q$ should have full row rank. Hence, if $H$ is the solution, then $H = H^*$, where

$$H^* := QMQ^T(QQ^T)^{-1}. \hfill (29)$$

Hence the see whether (28) has solution $H$ we have only to verify if $H^*$ is logical matrix and it satisfies (28).

We give an example.

Example 3.5. Consider the following BN:

$$\begin{align*}
    x_1(t + 1) &= (x_1(t) \land x_2(t) \land \neg x_4(t)) \lor (\neg x_1(t) \land x_2(t)) \\
    x_2(t + 1) &= x_2(t) \lor (x_3(t) \leftrightarrow x_4(t)) \\
    x_3(t + 1) &= (x_1(t) \land \neg x_4(t)) \lor (\neg x_1(t) \land x_2(t)) \lor (\neg x_1(t) \land \neg x_2(t) \land x_4(t)) \\
    x_4(t + 1) &= x_1(t) \land \neg x_2(t) \land x_4(t).
\end{align*}$$  \hfill (30)

Its ASSR is calculated as

$$x(t + 1) = Mx(t), \hfill (31)$$

where

$$M = \delta_{16}[11, 1, 11, 1, 11, 13, 15, 9, 1, 2, 1, 2, 9, 15, 13, 11].$$

Suppose $Z = F_1\{z_1, z_2, z_3\}$, where

$$\begin{align*}
    z_1 &= x_1 \lor x_4 \\
    z_2 &= \neg x_2 \\
    z_3 &= x_3 \leftrightarrow \neg x_4.
\end{align*}$$  \hfill (32)
Denote $x = \kappa_{i=1}^4 x_i$, $z = \kappa_{i=1}^3 z_i$, then
\[ z = Qx, \]
where $Q$ can be calculated as
\[ Q = \delta_8[8, 3, 7, 4, 6, 1, 5, 2, 4, 7, 3, 8, 2, 5, 1, 6]. \]

Using (29), we have
\[ H^* = \delta_8[2, 4, 8, 8, 1, 3, 3, 3]. \]

It is ready to verify (28). Hence $Z$ is an invariant subspace of (30).

### 3.2 Union of Invariant Subspaces

Assume $V_i, i = 1, 2$ are two $M$ invariant subspaces, where
\begin{align*}
V_1 &= F_\ell\{z_1^1, \ldots, z_p^1\}, \\
V_2 &= F_\ell\{z_1^2, \ldots, z_q^2\}. \\
\end{align*}

Then we have
\[ V_i = G_ix, \quad i = 1, 2, \]
where $x = \kappa_{i=1}^n x_i, G_1 \in L_{2p \times 2n}, G_2 \in L_{2q \times 2n}$.

**Theorem 3.6.** Assume $V_i, i = 1, 2$ are $M$ invariant subspaces. That is, there exist $H_1 \in L_{p \times p}$ and $H_2 \in L_{q \times q}$, such that
\begin{align*}
G_1M &= H_1G_1, \\
G_2M &= H_2G_2. \quad (35)
\end{align*}

Then
\[ V = V_1 \bigcup V_2 = F_\ell\{z_1^1, \ldots, z_p^1, z_1^2, \ldots, z_q^2\} \]
is also $M$-invariant. Moreover, the structure matrix of $V$, denoted by
\[ G = G_1 * G_2, \]
satisfies
\[ GM = HG, \]
where
\[ H = H_1 \otimes H_2. \]

To prove this theorem, we need the following lemma, which itself is useful.

**Lemma 3.7.** Let $A \in M_{p \times \ell}$, $B \in M_{q \times \ell}$, and $T \in L_{\ell \times r}$. Then
\[ (A * B)T = (AT) * (BT). \]

9
Proof. Denote
\[ A = [A^1, A^2, \ldots, A^t], \quad B = [B^1, B^2, \ldots, B^t], \]
where \( A^i = \text{Col}_i(A) \) \( (B^i = \text{Col}_i(B)) \) is the \( i \)-th column of \( A \) \( (B) \); and
\[ T = [\delta^1, \delta^2, \ldots, \delta^m]. \]

Then
\[
(A \ast B)T = ([A^1, A^2, \ldots, A^t] \ast [B^1, B^2, \ldots, B^t])T \\
= [A^1 \otimes B^1, A^2 \otimes B^2, \ldots, A^t \otimes B^t]T \\
= [A^{i_1} \otimes B^{i_1}, A^{i_2} \otimes B^{i_2}, \ldots, A^{i_m} \otimes B^{i_m}].
\]

(39) follows immediately. \( \square \)

Proof. (of Theorem 3.6) It is enough to prove (37) with (38). Denote \( G_1 = (G_1^1, \ldots, G_1^m) \), \( G_2 = (G_2^1, \ldots, G_2^m) \), where \( G_1^i = \text{Col}_i(G_1) \), \( G_2^i = \text{Col}_i(G_2) \), \( i = 1, 2, \ldots, 2^n \). Using Lemma 3.7
\[
GT = (G_1 \ast G_2)T = (G_1T) \ast (G_2T) \\
= (H_1G_1) \ast (H_2G_2) \\
= [(H_1G_1^1) \ast (H_2G_2^1), (H_1G_1^2) \ast (H_2G_2^2), \ldots, (H_1G_1^m) \ast (H_2G_2^m)] \\
= [(H_1G_1^1) \otimes (H_2G_2^1), (H_1G_1^2) \otimes (H_2G_2^2), \ldots, (H_1G_1^m) \otimes (H_2G_2^m)] \\
= [(H_1 \otimes H_2)(G_1^1 \otimes G_2^1), (H_1 \otimes H_2)(G_1^2 \otimes G_2^2), \ldots, (H_1 \otimes H_2)(G_1^m \otimes G_2^m)] \\
= (H_1 \otimes H_2)(G_1 \ast G_2) = (H_1 \otimes H_2)G.
\]
\( \square \)

### 3.3 Dynamics of Aggregated NB

Assume (12) is a large scale BN, and \( z_i, i = 1, \ldots, r \) are separating logical functions, which represent our interested properties. Denote by
\[ Z = F_{t \{ z_i \mid i = 1, \ldots, r \}} \]
We first try to find the smallest subspace \( Z \), which contains \( Z \) and is \( M \)-invariant.

**Algorithm 3.8.**

- **Step 1:** Set \( z^0 = \kappa_{i=1}^r z_i \), and assume
  \[ z^0 = G_0x. \]

  Calculate
  \[ z^1 = \{ G_0x \cup G_0Mx \} := G_1x. \]

- **Step \( k \):** Assume \( z^{k-1} = G_{k-1}x \) is known. Then
  \[ z^k = \{ G_{k-1}x \cup G_{k-1}Mx \} := G_kx. \]

- **Final Step:** Assume \( z^{k^*} = z^{k^*+1} \), then
  \[ Z := F_{t \{ z^{k^*} \}}. \] (40)
Remark 3.9. In Algorithm 3.8 at each step we assume in $z^i$ all the repeated functions have been deleted. Otherwise, $G_i$ maybe unnecessarily large.

By construction it is clear that the $\mathcal{Z}$ provided by (40) is the smallest subspace, containing $\mathcal{Z}$ and is $M$-invariant.

Definition 3.10. The dynamics of $\mathcal{Z}$ is called the $\{z_i \mid i = 1, \cdots, r\}$ aggregated BN.

Next, we try to find the dynamics of aggregated BN. Assume $\mathcal{Z} = \mathcal{F}(\bar{z})$ is a regular subspace, then

$$\bar{z} = \bar{G}x.$$  

Using Theorem 3.3 we have that

$$\bar{z}(t + 1) = \bar{G}x(t + 1) = \bar{G}Mx(t) = H\bar{G}x(t) = H\bar{z}(t).$$ (41)

Summarizing the above arguments, we have the following result.

Theorem 3.11. (41) is the dynamics of aggregated BN.

Remark 3.12. It is obvious that in Theorem 3.11 the regularity of $\mathcal{Z}$ has been ignored. From Algorithm 3.8 one sees easily that (27) is enough for (41). In fact, we do not care about if $\mathcal{Z}$ is regular or not. When it is not, we can not get the second part of equation (25), which is not interesting to us.

In the following an example is given to describe the technique for constructing aggregated BN.

Example 3.13. An opinion dynamic network is depicted in Fig. 3, where $x_i, i = 1, 2, \cdots, 9$ are players. Each player chooses his next opinion 1 (with white circle) for “agree” and 0 (with black circle) for “disagree” based on its neighborhood information. The boundary players A, B, C, D, E, F have invariant opinion 1, and U, V, W, X, Y, Z have invariant opinion 0.

Each player always follows the majority. Counting himself, a player has 5 neighbors. So the decision is unique. Note that they might have boundary neighbors, who have fixed attitude.

Using ASSR, we have

$$x(t + 1) = Mx(t),$$ (42)

where $x = \sum_{i=1}^{9} x_i$, and $M \in M_{256 \times 256}$ is in Appendix.

Now assume we are particularly interested in three situations: $S := \{x^1, x^2, x^3\}$, where

$$x^1 = \delta_{512}^{43} \sim \{1, 1, 1, 0, 1, 0, 1, 0, 0\},$$
$$x^2 = \delta_{512}^{143} \sim \{1, 0, 1, 1, 0, 1, 1, 1, 1\},$$
$$x^3 = \delta_{512}^{165} \sim \{1, 0, 1, 0, 1, 1, 0, 1, 1\}.$$

Then the index function for $S$ is defined as

$$g_1(x) = \begin{cases} 1, & x \in S, \\ 0, & Otherwise. \end{cases}$$
Correspondingly, we have its structure matrix

\[
\text{Col}_i(G_1) = \begin{cases} 
\delta_1^1, & \delta_{512} \in S, \\
\delta_2^1, & \text{Otherwise.}
\end{cases}
\]

Then \( G_2 = G_1 M \) can be expressed as

\[
\text{Col}_i(G_2) = \begin{cases} 
\delta_1^1, & i = 22, 89, 150, 278, \\
\delta_2^1, & \text{Otherwise.}
\end{cases}
\]

Furthermore,

\[ G_2 M = G_1. \]

Set \( z = z_1 z_2 \), where

\[ z_1 = G_1 x, \quad z_2 = G_3 x, \]

with \( x = \times_{i=1}^{9} x_i \). It follows that

\[
\begin{align*}
    z_1(t + 1) &= G_1 x(t + 1) \\
                &= G_1 M x(t) = G_2 x(t) \\
                &= z_2(t).
\end{align*}
\]

\[
\begin{align*}
    z_2(t + 1) &= G_2 x(t + 1) \\
                &= G_2 M x(t) = G_1 x(t) \\
                &= z_1(t).
\end{align*}
\]
Hence the smallest $M$ invariant subspace containing $g_1$ is
\[ G = F_t\{g_1, g_2\}. \]

The aggregated system becomes
\[ z_1(t+1) = z_2(t) = (J_T^2 \otimes I_2)z(t) \]
\[ z_2(t+1) = z_1(t) = (I_2 \otimes J_T^2)z(t), \]
where $z(t) = z_1(t)z_2(t)$. Hence, the ASSR of $z(t)$ is
\[ z(t+1) = [(J_T^2 \otimes I_2) * (I_2 \otimes J_T^2)]z(t) = \delta_4[1,3,2,4]z(t). \]

The aggregated BN (44) is much smaller than the original BN (41), but it is enough to describe the dynamics of the state $z^* = g_1(x)$, which is concerned by us.

**Remark 3.14.** From Example 3.13 one sees easily that as the related attractor of a BN is of small size, then the aggregated BN might reduce the size of the original BN tremendously. Now one may ask if the related attractor is of large size, then what can we do? Of course, if the $M$ invariant subspace containing the separating logical functions, which represent the properties interesting to us, involves large size attractors, then the aggregated BN may still have large scale. Fortunately, as pointed by Kauffman [8]: The “vast order” of a large scale cellular network is decided by “tiny attractors”. This fact makes the aggregation technique more useful.

## 4 Invariant Subspace of BCN

Consider BCN (16) with its ASSR (18). Splitting $L$ into $2^m$ blocks as
\[ L = [M_1, M_2, \ldots, M_{2^m}], \]
where
\[ M_r = L\delta^r_{2^m} \in \mathcal{L}_{2^n \times 2^n}, \quad r = 1, 2, \ldots, 2^m. \]

**Definition 4.1.** (i) $Z$ is said to be $L$ invariant, if $Z$ is $M_i$ invariant for all $i = 1, 2, \ldots, 2^m$.

(ii) $Z$ is said to be partly $L$ invariant with respect to $U \subset \delta_{2^m} \{1, 2, \ldots, 2^m\}$, if $Z$ is $M_i$ invariant for all $u \in U$.

**Definition 4.2.** (i) $V$ is called a control invariant subspace containing $Z$, if it contains $Z$, and for any control $u$ it is $Lu$ invariant.

(ii) The intersection of all control invariant subspaces containing $Z$ is called the smallest control invariant subspace containing $Z$, and denoted by $\overline{Z}$.

(iii) $V$ is called a partly control invariant subspace containing $Z$ with respect to $U$, if it contains $Z$, and for any control $u \in U$ it is $Lu$ invariant.

(ii) The intersection of all partly control invariant subspaces containing $Z$ with respect to $U$ is called the smallest partly control invariant subspace containing $Z$ with respect to $U$, and denoted by $\overline{Z}^U$. 

13
Assume \( Z = \mathcal{F}_\ell \{ z_1, z_2, \ldots, z_r \} \) and \( \overline{Z} = \mathcal{F}_\ell \{ z_1, \cdots, z_r, z_{r+1}, \cdots, z_s \} \). Denote \( z = \kappa^n_i z_i \), then there exists a \( G \in \mathcal{L}_{2^n \times 2^n} \), such that
\[
z = G \kappa^n_i x_i := Gx.
\]
Since \( Z \) is control invariant subspaces, for \( u = \delta_{2m} \) we have
\[
GM_i = H_i G, \quad i = 1, 2, \cdots, 2m.
\]

It follows that
\[
\begin{align*}
z(t + 1) &= Gx(t + 1) = GLu(t)x(t) \\
&= [H_1, H_2, \cdots, H_{2m}] Gx(t) \\
&= [H_1, H_2, \cdots, H_{2m}] u(t) z(t)
\end{align*}
\]
Define \( H := [H_1, H_2, \cdots, H_{2m}] \), then we have the aggregated BCN as
\[
z(t + 1) = Hu(t) z(t). \tag{47}
\]

Next, we consider the case when there is a constrain on control, as \( u(t) \in U \subset \Delta_{2m} \). Assume \( U \) is state-depending. That is,
\[
U = \{ u \neq \delta_{2m}^\alpha \text{ if } z \in X_\alpha \subset X = \Delta_{2^n} \mid \alpha \in \Xi \subset \Delta_{2m} \}.
\]
We need the following notation: \( A \in \mathcal{M}_{p \times q} \) is called a zero-extended logical matrix if
\[
\text{Col}(A) \subset \Delta_p \cup 0_p,
\]
That is \( A \) may contain some zero columns.

Now consider partly control invariant subspaces containing \( Z \). Assume when \( z = \delta_{2m}^k \), \( u = \delta_{2m}^\alpha \) is forbidden. Then in equation (47) we set
\[
\text{Col}_k(H_\alpha) = 0_{2^n}.
\]
Finally, we can construct the modified \( H \), denoted by \( H^U \), to describe the partly control invariant aggregated BCN, which has its dynamic equation as
\[
z(t + 1) = H^U u(t) z(t). \tag{48}
\]

We use an example to depict it.

**Example 4.3.** Recall Example 3.13. Assume the boundary player \( V \) is replaced by a control \( u(t) \), ( refer to Fig. 3).
Then it is a normal routine to figure out the dynamics of this BCN as
\[
x(t + 1) = [N, M] u(t) x(t), \tag{49}
\]
where \( M \) is the same as in Example 3.13, \( N \) is also in Appendix.
Assume we are still particularly interested in the \( S \) as in Example 3.13 i.e., \( S := \{ x^1, x^2, x^3 \} \), where \( x^1 = \delta_{512}^{143}, x^2 = \delta_{512}^{143} \sim \{ 1, 0, 1, 1, 0, 1, 1, 1 \}, x^3 = \delta_{512}^{165} \).
Then it is easy to calculate that \( G_1 N = G_3, \ G_3 N = G_4, \ G_4 N = G_5, \ G_5 N = G_7; \ G_2 N = G_6, \ G_6 N = G_5; \ G_7 M = G_7, \ G_7 N = G_7 \), where

\[
\text{Col}_i(G_3) = \begin{cases} \delta^1_2, & i = 43, 47, 143, 164, 229, 420, \\ \delta^2_2, & \text{Otherwise}. \end{cases}
\]

\[
\text{Col}_i(G_4) = \begin{cases} \delta^1_2, & i = 59, 118, 278, \\ \delta^2_2, & \text{Otherwise}. \end{cases}
\]

\[
\text{Col}_i(G_5) = \begin{cases} \delta^1_2, & i = 164, 299, 420, \\ \delta^2_2, & \text{Otherwise}. \end{cases}
\]

\[
\text{Col}_i(G_6) = \begin{cases} \delta^1_2, & i = 278, \\ \delta^2_2, & \text{Otherwise}. \end{cases}
\]

\[
\text{Col}_i(G_7) = \delta^2_2, \quad i = 1, 2, \cdots, 512.
\]

Set

\[
z_i = G_i x, \quad i = 1, 2, \cdots, 7,
\]

\( W = \{3, 4, 5, 6\} \) and we assume the feasible control set

\[
U = \{u(t) \neq \delta_2^2 | x(t) \in W\}
\]

Finally, the partly control invariant aggregation BCN, is obtained as follows.

\[
z(t + 1) = H^U u(t) z(t),
\]

where \( z(t) = (z_1(t), z_2(t), z_3(t), z_4(t), z_5(t), z_6(t), z_7(t))^T \), and

\[
H^U = \delta_7 [6, 3, 4, 5, 7, 5, 7, 2, 1, 0, 0, 0, 7].
\]

The state-transition graph is depicted in Fig. 4.

5 Minimum Realization of BCN

Consider a BCN \([10]\) with outputs (observers)

\[
\begin{align*}
y_1(t) &= \xi_1(x_1(t), \cdots, x_n(t)), \\
y_2(t) &= \xi_2(x_1(t), \cdots, x_n(t)), \\
&\vdots \\
y_r(t) &= \xi_r(x_1(t), \cdots, x_n(t)).
\end{align*}
\]

Then the input-output BCN \([10]-[51]\) has ASSR as

\[
\begin{align*}
x(t + 1) &= Lu(t)x(t), \\
y(t) &= H x(t).
\end{align*}
\]

15
After a coordinate change $T: x \rightarrow z$, expressed by $z = Tx$, where $T \in L_{2^n \times 2^n}$, (18) becomes

$$
\begin{align*}
\begin{cases}
  z(t+1) = \tilde{L}u(t)z(t), \\
y(t) = \tilde{H}z(t),
\end{cases}
\end{align*}
$$

(53)

where

$$
\begin{align*}
\tilde{L} &= TL(I_m \otimes T), \\
\tilde{H} &= HT^T.
\end{align*}
$$

If under the coordinate frame $z$ (51), expressed as (53), has the form of (25), then it is clear that $Z = F_\ell\{z^1\}$ is a control invariant subspace, containing $\mathcal{Y}$. In fact, we can ignore $z^2$ and give the following definition.

**Definition 5.1.** Consider BCN (51), if there exists a subspace $Z = F_\ell\{z^1_1, x^1_2, \ldots, z^1_r\}$ such that

$$
\begin{align*}
\begin{cases}
  z^1(t+1) = F^1(z^1, u), \\
y(t) = \xi(z^1(t)),
\end{cases}
\end{align*}
$$

(54)

then (54) is called a realization of (16)-(51).

**Remark 5.2.**

(i) From Definition 5.1, $Z$ is a control invariant subspace, containing $\mathcal{Y}$.

(ii) In Definition 5.1, $Z$ is not required to be a regular subspace.

(iii) It is obvious that (54) and (16)-(51) have the same input-output mapping.

**Definition 5.3.** Consider BCN (51), if $Z = F_\ell\{z^1_1, x^1_2, \ldots, z^1_r\}$ is the smallest control invariant subspace containing $\mathcal{Y}$, then the corresponding BN (54) is called the minimum realization of (16)-(51).
Proposition 5.4. Assume \( Z = \mathcal{F}_\ell \{ z_1^1, x_2^1, \cdots, z_r^1 \} \) is the smallest control invariant subspace containing \( \mathcal{Y} \) and \( Z = Gx \). then

(i) there exists a set of logical matrix \( H_i \in L_{r \times r}, i = 1, 2, \cdots, 2^m \) such that

\[
GM_i = H_i G, \quad i = 1, 2, \cdots, 2^m; \tag{55}
\]

(ii) the minimum realization of (16)-(51) has its ASSR as

\[
\begin{cases}
    z^1(t+1) = Hu(t)z^1(t), \\
y(t) = \Xi z^1(t),
\end{cases}
\]

where \( \Xi \) is the structure matrix of \( \xi \), and

\[
H = [H_1, H_2, \cdots, H_{2^m}].
\]

The following algorithm provides a way to construct the minimum realization of a BCN.

Algorithm 5.5.  

• **Step 1:**

  Set

  \( O_0 = \{ y_1, y_2, \cdots, y_p \} \).

  Calculate

  \( O_1 = \{ yM_1, yM_2, \cdots, YM_{2^m} \mid y \in O_0 \} \setminus \{ O_0 \} \).

• **Step s:** (\( s > 0 \))

  Calculate

  \( O_{s+1} = \{ yM_1, yM_2, \cdots, yM_{2^m} \mid y \in O_s \} \setminus \{ O_r \mid r = 0, 1, \cdots, s \} \).

• **Last Step.** If

  \( O_{s^*+1} = \emptyset \).

  then

  \( \mathcal{Z}^* := \mathcal{F}_\ell \{ O_r \mid r = 0, 1, \cdots, s^* \} \)

  is the smallest control invariant subspace containing \( \mathcal{Y} \).

Assume \( \mathcal{Z}^* = \mathcal{F}_\ell \{ z_1, z_2, \cdots, z_r \} \), set \( z = \kappa \sum_{i=1}^r z_i \), then

\[
z(t+1) = [H_1, H_2, \cdots, H_{2^m}] u(t)z(t),
\]

\[
y(t) = \Xi z(t) \tag{57}
\]

is the minimum realization of BCN (16)-(51).

Next, we consider an example.
Example 5.6. Consider a BCN, with its ASSR as

\begin{equation}
\begin{cases}
x(t + 1) = Lu(t)x(t), \\
y(t) = \Xi x(t),
\end{cases}
\end{equation}

(58)

where \( x(t) = \bigwedge_{i=1}^{n} x_i(t) \), \( u(t) = u_1(t)u_2(t) \), and

\[
L = [M_1, M_2, M_3, M_4],
\]

with

\[
M_1 = \begin{pmatrix}
0 & 0 & 1 \\
1 & 0 & 0 \\
0 & 1 & 0 \\
0 & 0 & X
\end{pmatrix},
M_2 = \begin{pmatrix}
0 & 1 & 0 \\
1 & 0 & 0 \\
0 & 0 & 1 \\
0 & 0 & X
\end{pmatrix},
\]

\[
M_3 = \begin{pmatrix}
1 & 0 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1 \\
0 & 0 & X
\end{pmatrix},
M_4 = \begin{pmatrix}
0 & 0 & 1 \\
0 & 1 & 0 \\
1 & 0 & 0 \\
0 & 0 & X
\end{pmatrix},
\]

where \( X \in \mathcal{L}_{(2^{n-3}) \times (2^{n-3})} \) is an uncertain logical matrix.

\[
\Xi = \delta_2[1, 2, 1, 2, 2, \ldots, 2].
\]

Denote \( y_1 = y \), then it is easy to calculate that

\[
y_1M_1 = \delta_2[2, 1, 12, 2, \ldots, 2] := y_2,
y_1M_2 = y_2,
y_1M_3 = y_1,
y_1M_4 = y_1,
y_2M_1 = \delta_2[1, 2, 2, \ldots, 2] := y_3,
y_2M_2 = y_1,
y_2M_3 = y_2,
y_2M_4 = y_3,
y_3M_1 = y_1,
y_3M_2 = y_3,
y_3M_3 = y_3,
y_3M_4 = y_2.
\]

Let

\[
z_1 = y_1, \quad z_2 = y_2, \quad z_3 = y_3.
\]

Hence we have

\[
z_1(t + 1) = [Z_2, Z_2, Z_1]u(t)z(t),
z_2(t + 1) = [Z_3, Z_1, Z_2]u(t)z(t),
z_3(t + 1) = [Z_1, Z_3, Z_2]u(t)z(t).
\]
where \( u(t) = u_1(t)u_2(t), z(t) = z_1(t)z_2(t)z_3(t), \) and

\[
Z_1 = I_2 \otimes J_T^T = \delta_2[1, 1, 1, 1, 2, 2, 2],
Z_2 = J_2 \otimes I_2 \otimes J_2 = \delta_2[1, 1, 2, 1, 1, 2, 2],
Z_3 = J_4 \otimes I_2 = \delta_2[1, 2, 1, 2, 1, 1, 2].
\]

Finally, the minimum realization of (58) is obtained as

\[
\begin{cases}
    z(t + 1) = L^* u(t)z(t), \\
y(t) = Z_1 z(t),
\end{cases}
\]  
(59)

where

\[
L^* = \delta_8[1, 2, 3, 4, 5, 6, 7, 8, 1, 2, 3, 4, 5, 6, 7, 8].
\]

The state-transition graph of this minimum realization is depicted in Fig. 5.

Motivated by Example 5.6, the following result is easily verifiable.

**Proposition 5.7.** Consider BCN (16)-(51). If there exists a coordinate change \( z = Tx, \)
such that

\[
TM_i T^T = \begin{bmatrix}
J^1_i & 0 & \cdots & 0 \\
0 & J^2_i & \cdots & 0 \\
0 & 0 & \ddots & \cdots \\
0 & 0 & \cdots & J^s_i
\end{bmatrix}, \quad i = 1, 2, \ldots, 2^m,
\]

where \( z = (z^1, z^2, \ldots, z^s) \) and \( z^k \) corresponds to \( J^k_i \). Moreover, if \( y \in \mathcal{F}_k(z^k) \), then there exists a realization

\[
\begin{align*}
\{ & z(t+1) = [J^1_k, J^2_k, \ldots, J^{2^m}_k] u(t) z(t), \\
y(t) = \Xi_k z(t) \}
\end{align*}
\]

Moreover, if \( J^k_i, \quad i = 1, 2, \ldots, 2^m \) can not be further diagonized simultaneously for any \( 1 \leq k \leq s \), then (60) is a minimum realization.

**Remark 5.8.**
(i) Minimum realization can also be considered as a kind of aggregations, which separate states into two categories: related states and unrelated states. Then only related states are modeled.

(ii) For a large scale BN, we can inject controls on different nodes and observe some other nodes (which are considered as outputs). Then observe the input-output relations to investigate the minimum realization, which reveals part structure of the BN. By changing input nodes and output nodes, another part structure may be revealed. The minimum realizations might be of much smaller sizes, which makes the investigations easier. This method may provide a way to solve the problem of computational complexity.

An alternative way to deal with a large-scale BN is to observe some special interested states. Then the observers, as a set of logical functions, can be considered as separating functions. Then we may define the follows:

**Definition 5.9.** A BN with some outputs is called an observe-based BN. Using observers as a set of logical functions, the dynamic equations of the minimum-invariant subspace containing observers are called the observe-based minimum realization of the observe-based BN.

In fact, through observed data, we may construct the dynamic equations of the observe-based minimum realization. In this way, part of the structure of overall BN can be constructed. Using different observes, the interested parts of structure of overall BN might be construct.

### 6 Conclusion

In this paper a logical function is considered as an index function of a subset of nodes of a BN or BCN. Using this idea, a set of logical functions are used as separating functions to aggregate nodes. Then the (minimum) invariant subspace containing the preassigned set of logical functions, is constructed. Furthermore, the dynamic equations for the invariant subspace, which represents the aggregated nodes, are obtained. Then the (minimum) invariant subspace of BNC is also defined and the corresponding dynamic equations are also constructed. Finally, as the outputs of a BC/BNC are considered as the set
of separating functions, the minimum realization of a BCN (or the observe-based minimum realization for BC) is defined, and their properties are investigated.

When a BN/BCN is of large scale, the structure matrix of overall BN might be huge and practically uncomputable. Using input-output realization and observe-based minimum realization, the interested parts of structure of the BN could be obtained. These might be much smaller sub-BN may dominate the behaviors of whole BN. Hence, this technique may provide an efficient way to solve the computational complexity of large scale BN/BCN.

Acknowledgment This work was completed when the second and third authors visiting the Center of STP Theory and Applications.

References

[1] C.J. Burges, A tutorial on support vector machines for pattern recognition, *Data Mining Knowl. Disc.*, Vol. 2, 121-167, 1998.

[2] D. Cheng, H. Qi, State-space analysis of Boolean networks, *IEEE Trans. Neur. Netw.*, Vol. 21, No. 4, 584-594, 2010.

[3] D. Cheng, H. Qi, Z. Li, *Analysis and Control of Boolean Networks: A Semi-tensor Product Approach*, Springer, London, 2011.

[4] D. Cheng, H. Qi, Y. Zhao, *An Introduction to Semi-tensor Product of Matrices and Its Applications*, World Scientific, Singapore, 2012.

[5] E. Fornasini, M.E. Valcher, Recent developments in Boolean networks control, *J. Contr. Dec.*, Vol. 3, No. 1, 1-18, 2016.

[6] S. Haykin, *Neural Networks and Learning Machines*, 3rd Ed., Prentice Hall, New York, 2009.

[7] S.A. Kauffman, Metabolic stability and epigenesis in randomly constructed genetic nets, *J. Theor. Biol.*, Vol. 22, 437-467, 1969.

[8] S.A. Kauffman, *At Home in the Universe*, Oxford Univ. Press, London, 1995.

[9] H. Li, G. Zhao, M. Meng, J. Feng, A survey on applications of semi-tensor product method in engineering, *Science China*, Vol. 61, 010202:1-010202:17, 2018.

[10] J. Lu, H. Li, Y. Liu, F. Li, Survey on semi-tensor product method with its applications in logical networks and other finite-valued systems, *IET Contr. Thm& Appl.*, Vol. 11, No. 13, 2040-2047, 2017.

[11] A. Muhammad, A. Rushdi, F.A. M. Ghaleb, A tutorial exposition of semi-tensor products of matrices with a stress on their representation of Boolean function, *JKAU Comp. Sci.*, Vol. 5, 3-30, 2016.

[12] Optimal one-bit perturbation in Boolean networks based on cascading aggregation, *Front Inform. Technol. Electron. Eng.*, Vol. 21, No. 1, 294-303, 2020.
[13] Y. Zhao, J. Kim, M. Filippone, Aggregation algorithm towards large-scale Boolean network analysis, *IEEE Trans. Aut. Contr.*, Vol. 58, No. 8, 1976-1985, 2013.

[14] Y. Zhao, B. Ghosh, D. Cheng, Control of large-scale Boolean networks via network aggregation, *IEEE Trans. Neur. Netw. Learn Sys.*, Vol. 27, No. 7, 1527-1536, 2015.

[15] Q. Zhao, A remark on 'Scalar equations for synchronous Boolean networks with biologic applications', by C. Farrow, J. Heidel, J. Maloney, and J. Rogers, *IEEE Trans. Neural Netw.*, Vol. 16, No. 6, 1715-17176, 2005.

7 Appendix

(i) The structure matrix of BN (41):

\[
M = 5_{122} = \begin{bmatrix}
1 & 1 & 1 & 2 & 1 & 1 & 5 & 8 & 1 & 10 & 2 & 10 & 1 & 10 & 6 & 16 \\
1 & 9 & 1 & 12 & 33 & 43 & 39 & 48 & 9 & 10 & 26 & 28 & 41 & 44 & 64 & 64 \\
1 & 1 & 5 & 6 & 37 & 37 & 37 & 40 & 1 & 10 & 22 & 30 & 37 & 46 & 54 & 64 \\
33 & 41 & 53 & 64 & 37 & 47 & 55 & 64 & 57 & 58 & 62 & 64 & 61 & 64 & 64 & 64 \\
1 & 9 & 1 & 10 & 1 & 9 & 5 & 16 & 73 & 74 & 74 & 74 & 74 & 74 & 78 & 80 \\
9 & 9 & 9 & 12 & 41 & 43 & 47 & 48 & 73 & 74 & 90 & 92 & 105 & 108 & 128 & 128 \\
1 & 9 & 5 & 14 & 37 & 45 & 37 & 48 & 73 & 74 & 94 & 94 & 109 & 110 & 126 & 128 \\
41 & 41 & 61 & 64 & 45 & 47 & 63 & 64 & 121 & 122 & 126 & 128 & 125 & 128 & 128 & 128 \\
1 & 1 & 1 & 2 & 1 & 1 & 1 & 5 & 8 & 65 & 74 & 82 & 90 & 95 & 74 & 86 & 96 \\
1 & 9 & 17 & 28 & 33 & 43 & 55 & 64 & 89 & 90 & 90 & 92 & 121 & 124 & 128 & 128 \\
257 & 257 & 277 & 278 & 293 & 293 & 309 & 312 & 337 & 346 & 342 & 350 & 373 & 382 & 374 & 384 \\
305 & 313 & 309 & 320 & 309 & 319 & 311 & 320 & 377 & 378 & 382 & 384 & 381 & 384 & 384 & 384 \\
65 & 73 & 65 & 74 & 65 & 73 & 69 & 80 & 73 & 74 & 90 & 90 & 73 & 74 & 94 & 96 \\
201 & 201 & 217 & 220 & 233 & 235 & 255 & 256 & 217 & 218 & 210 & 210 & 249 & 252 & 256 & 256 \\
321 & 329 & 341 & 350 & 357 & 365 & 373 & 384 & 345 & 346 & 350 & 350 & 381 & 382 & 382 & 384 \\
505 & 505 & 509 & 512 & 509 & 511 & 511 & 512 & 505 & 506 & 510 & 512 & 509 & 512 & 512 & 512 \\
1 & 1 & 1 & 2 & 33 & 33 & 37 & 40 & 1 & 10 & 2 & 10 & 33 & 42 & 38 & 48 \\
33 & 41 & 33 & 44 & 33 & 43 & 39 & 48 & 41 & 42 & 58 & 60 & 41 & 44 & 64 & 64 \\
289 & 289 & 293 & 294 & 293 & 293 & 293 & 296 & 289 & 298 & 310 & 318 & 293 & 302 & 310 & 320 \\
289 & 297 & 309 & 320 & 293 & 303 & 311 & 320 & 313 & 314 & 318 & 320 & 317 & 320 & 320 & 320 \\
1 & 9 & 1 & 10 & 33 & 41 & 37 & 48 & 73 & 74 & 74 & 74 & 105 & 106 & 110 & 112 \\
169 & 169 & 169 & 172 & 169 & 171 & 175 & 176 & 233 & 234 & 250 & 252 & 233 & 236 & 256 & 256 \\
289 & 297 & 293 & 302 & 293 & 301 & 293 & 304 & 361 & 362 & 382 & 382 & 365 & 366 & 382 & 384 \\
425 & 425 & 445 & 448 & 429 & 431 & 447 & 448 & 505 & 506 & 510 & 512 & 509 & 512 & 512 & 512 \\
257 & 257 & 257 & 258 & 289 & 289 & 293 & 296 & 321 & 330 & 338 & 346 & 353 & 362 & 374 & 384 \\
417 & 425 & 433 & 444 & 417 & 427 & 439 & 448 & 505 & 506 & 508 & 508 & 505 & 508 & 512 & 512 \\
289 & 289 & 309 & 310 & 293 & 293 & 309 & 312 & 369 & 378 & 374 & 382 & 373 & 382 & 374 & 384 \\
433 & 441 & 437 & 448 & 437 & 447 & 439 & 448 & 505 & 506 & 510 & 512 & 509 & 512 & 512 & 512 \\
449 & 457 & 449 & 458 & 481 & 489 & 485 & 496 & 457 & 458 & 474 & 474 & 489 & 490 & 510 & 512 \\
489 & 489 & 508 & 508 & 489 & 491 & 511 & 512 & 505 & 506 & 506 & 508 & 505 & 508 & 512 & 512 \\
481 & 489 & 501 & 510 & 485 & 493 & 501 & 512 & 505 & 506 & 510 & 510 & 509 & 510 & 510 & 512 \\
505 & 505 & 509 & 512 & 509 & 511 & 511 & 512 & 505 & 506 & 510 & 512 & 509 & 512 & 512 & 512 \\
\end{bmatrix}
\]

(ii) Structure matrix for Example 4.3.
\[ N = 6512 \]

```
1 4 1 2 1 1 5 8 1 10 2 10 1 10 6 16
1 9 1 12 1 11 7 16 9 10 26 28 9 12 32 32
1 1 5 6 5 5 8 1 10 22 30 5 14 22 32
1 9 21 32 37 47 55 64 25 26 30 32 61 64 64 64
1 9 1 10 1 9 5 16 73 74 90 92 73 76 96 96
1 9 5 14 5 13 5 16 73 74 94 94 77 78 94 96
1 9 29 32 45 47 63 64 89 90 94 96 77 78 94 96
1 1 1 2 1 1 5 8 65 74 82 90 65 74 86 96
1 9 17 28 1 11 23 32 89 90 90 92 89 92 96 96
257 257 277 278 261 261 277 280 337 346 342 350 341 350 342 352
273 281 277 288 309 319 311 320 345 346 350 352 381 384 384 384
65 73 65 74 65 73 69 80 73 74 90 90 73 74 94 96
291 291 217 220 201 203 223 224 217 218 218 220 217 220 224 224
321 329 341 350 325 333 341 352 345 346 350 350 349 350 350 352
473 473 477 480 509 511 511 512 473 474 478 480 509 512 512 512
1 1 1 2 1 1 1 5 8 1 1 2 10 2 10 1 10 6 16
1 9 1 12 33 43 39 48 9 10 26 28 41 44 64 64
257 257 261 262 293 293 293 296 257 266 278 286 293 302 310 320
249 297 309 320 293 303 311 320 313 314 318 320 317 320 320 320
1 9 1 10 1 9 5 16 73 74 74 74 73 74 78 80
137 137 137 140 169 171 175 176 201 202 218 220 233 236 256 256
257 265 261 270 293 301 293 304 329 330 350 350 365 366 382 384
425 425 445 448 449 431 447 448 505 506 510 512 509 512 512 512
257 257 257 258 257 257 261 264 321 330 338 346 321 330 342 352
395 393 401 412 417 427 439 448 473 474 474 474 476 505 508 512 512
257 257 277 278 293 293 309 312 337 346 342 350 373 382 374 384
433 441 437 448 437 447 439 448 505 506 510 512 509 512 512 512
449 457 449 458 449 457 453 464 457 458 474 474 457 458 478 480
457 457 473 476 489 491 511 512 473 474 474 474 476 505 508 512 512
449 457 460 478 485 493 501 512 473 474 478 478 480 509 510 510 512
505 505 509 512 509 511 511 512 505 506 510 512 509 512 512 512
```