

Invariant Subspace Approach to Boolean (Control) Networks ^{*}

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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 than their positions.

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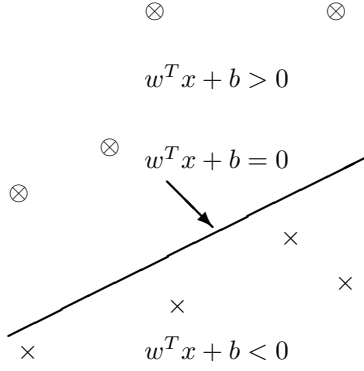


Figure 1: Hyperplane For Point Separation

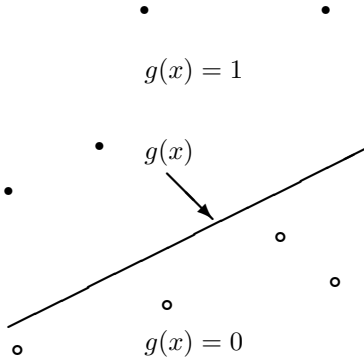


Figure 2: Logical Function For State Separation

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 $\{x \mid w^T x + b < 0\}$, (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 $\{x \mid g(x) = 0\}$, (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 $M \ltimes N$, is defined as

$$(M \otimes I_{t/n}) (N \otimes I_{t/p}) \in \mathcal{M}_{mt/n \times qt/p}, \quad (1)$$

where \otimes is the Kronecker product.

Note that when $n = p$, $M \ltimes N = MN$. 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 \ltimes . Throughout this paper the matrix product is assumed to be STP, and the symbol \ltimes is mostly omitted.

The following are some basic properties:

Proposition 2.2. 1. (Associative Law)

$$(F \ltimes G) \ltimes H = F \ltimes (G \ltimes H). \quad (2)$$

2. (Distributive Law)

$$\begin{cases} F \ltimes (aG \pm bH) = aF \ltimes G \pm bF \ltimes H, \\ (aF \pm bG) \ltimes H = aF \ltimes H \pm bG \ltimes H, \end{cases} \quad a, b \in \mathbb{R}. \quad (3)$$

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

$$X \ltimes Y = X \otimes Y. \quad (4)$$

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

$$\omega \ltimes \sigma = \sigma \otimes \omega. \quad (5)$$

About the transpose, we have

Proposition 2.4.

$$(A \ltimes B)^T = B^T \ltimes A^T. \quad (6)$$

About the inverse, we have

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

$$(A \ltimes B)^{-1} = B^{-1} \ltimes A^{-1}. \quad (7)$$

The following property is for STP only.

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

$$X \ltimes M = (I_m \otimes M) X. \quad (8)$$

Definition 2.7. [3] *A matrix $W_{[m,n]} \in \mathcal{M}_{mn \times mn}$, defined by*

$$W_{[m,n]} := [I_n \otimes \delta_m^1, I_n \otimes \delta_m^2, \dots, I_n \otimes \delta_m^m,] \quad (9)$$

is called the (m,n) -th dimensional swap matrix, where δ_m^i is the i -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*

$$W_{[m,n]} \ltimes X \ltimes Y = Y \ltimes X. \quad (10)$$

Definition 2.9. *Let $A \in \mathcal{M}_{p \times n}$ and $B \in \mathcal{M}_{q \times n}$. Then the Khatri-Rao Product of A and B is*

$$A * B = [\text{Col}_1(A) \ltimes \text{Col}_1(B), \dots, \text{Col}_n(A) \ltimes \text{Col}_n(B)] \in \mathcal{M}_{pq \times n}. \quad (11)$$

2.2 Matrix Expression of BN

Definition 2.10. *A BN is described by*

$$\begin{cases} x_1(t+1) = f_1(x_1(t), \dots, x_n(t)), \\ x_2(t+1) = f_2(x_1(t), \dots, x_n(t)), \\ \vdots \\ x_n(t+1) = f_n(x_1(t), \dots, x_n(t)), \end{cases} \quad (12)$$

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

Using vector form expression: $1 \sim \delta_2^1 = (1, 0)^T$, $0 \sim \delta_2^2 = (0, 1)^T$. Then $x(t)$ can be expressed as $x(t) \in \Delta_2$, where Δ_k is the set of columns of I_k .

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

Then a BN has its matrix form, called the ASSR of BN, as follows:

Proposition 2.11. (i) For a logical function $f : \mathcal{D}^n \rightarrow \mathcal{D}$, there exists a unique logical matrix $M_f \in \mathcal{L}_{2 \times 2^n}$ such that in vector form

$$f(x_1, x_2, \dots, x_n) = M_f \bowtie_{i=1}^n x_i. \quad (13)$$

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

$$\begin{cases} x_1(t+1) = M_1 \bowtie_{i=1}^n x_i(t), \\ x_2(t+1) = M_2 \bowtie_{i=1}^n x_i(t), \\ \vdots, \\ x_n(t+1) = M_n \bowtie_{i=1}^n x_i(t). \end{cases} \quad (14)$$

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

$$x(t+1) = Mx(t), \quad (15)$$

where

$$M = M_1 * M_2 * \dots * M_n \in \mathcal{L}_{2^n \times 2^n}$$

is called the structure matrix of BN (12).

Similarly, the BCN is described as follows:

$$\begin{cases} x_1(t+1) = f_1(x_1(t), \dots, x_n(t); u_1(t), \dots, u_m(t)), \\ x_2(t+1) = f_2(x_1(t), \dots, x_n(t); u_1(t), \dots, u_m(t)), \\ \vdots, \\ x_n(t+1) = f_n(x_1(t), \dots, x_n(t); u_1(t), \dots, u_m(t)), \end{cases} \quad (16)$$

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

We also have similar algebraic expressions for BCN.

Proposition 2.12. Consider BCN (16).

(i) Its componentwise ASSR is

$$\begin{cases} x_1(t+1) = L_1 u(t) x(t), \\ x_2(t+1) = L_2 u(t) x(t), \\ \vdots, \\ x_n(t+1) = L_n u(t) x(t), \end{cases} \quad (17)$$

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

(ii) Its ASSR is

$$x(t+1) = Lu(t)x(t), \quad (18)$$

where $L = L_1 * L_2 * \dots * L_n \in \mathcal{L}_{2^n \times 2^{m+n}}$ is the structure matrix of BCN (16).

Definition 2.13. [2] Consider BN (12) or BCN (13).

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

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

- (ii) Let $z_1, z_2, \dots, z_r \in \mathcal{X}$. Then the subspace generated by z_1, z_2, \dots, z_r , is defined by

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

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

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

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

$$z = T x, \quad (22)$$

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

Definition 2.14. Support $\mathcal{Z} = \mathcal{F}_\ell\{z_1, z_2, \dots, z_r\}$ has its algebraic expression (22) with non-singular T , then $\mathcal{X} \rightarrow \mathcal{Z}$ 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 $f \in \mathcal{X}$, 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 $f(x_1, x_2, \dots, x_n)$ can be considered as an index function of a subset of node set $N := \{x_1, x_2, \dots, x_n\}$. 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 $\pi : 2^N \rightarrow \mathcal{F}_\ell\{x_1, x_2, \dots, x_n\}$ determined by $\pi(S) = f_s$, which is defined by (23). Then it is obvious that π 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:

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

$z_p \in \mathcal{X}$. 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 $z_i(x)$, $i = 1, 2, \dots, r$, where $r \ll n$. We then can aggregate x into 2^r groups as

$$x^k := \{x \mid z(x) = \delta_{2^r}^k\}, \quad k = 1, \dots, 2^r. \quad (24)$$

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

$$\mathcal{Z} := \mathcal{F}_\ell\{z_1, z_2, \dots, z_r\}$$

is called the $\{z_i \mid i = 1, 2, \dots, r\}$ aggregated subspace.

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

To find the dynamics of \mathcal{Z} , we need some new concepts.

Definition 3.2. Given BN (12).

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

$$\begin{cases} z^1(t+1) = \tilde{F}^1(z^1(t)), & z^1(t) \in \mathcal{Z}^1 \\ z^2(t+1) = \tilde{F}^2(z(t)), & z^2(t) \in \mathcal{Z}^2, z(t) \in \mathcal{Z}, \end{cases} \quad (25)$$

then \mathcal{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 (15). Suppose $\mathcal{Z}^1 = \mathcal{F}_\ell\{z_1^1, z_2^1, \dots, z_r^1\}$ is a regular space with its ASSR as

$$z^1 = Qx, \quad (26)$$

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

$$QM = HQ. \quad (27)$$

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

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

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

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

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 \mathcal{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

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

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:

$$\begin{cases} x_1(t+1) = (x_1(t) \wedge x_2(t) \wedge \neg x_4(t)) \vee (\neg x_1(t) \wedge x_2(t)) \\ x_2(t+1) = x_2(t) \vee (x_3(t) \leftrightarrow x_4(t)) \\ x_3(t+1) = (x_1(t) \wedge \neg x_4(t)) \vee (\neg x_1(t) \wedge x_2(t)) \vee (\neg x_1(t) \wedge \neg x_2(t) \wedge x_4(t)) \\ x_4(t+1) = x_1(t) \wedge \neg x_2(t) \wedge x_4(t). \end{cases} \quad (30)$$

Its ASSR is calculated as

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

where

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

Suppose $\mathcal{Z} = \mathcal{F}_\ell\{z_1, z_2, z_3\}$, where

$$\begin{cases} z_1 = x_1 \bar{\vee} x_4 \\ z_2 = \neg x_2 \\ z_3 = x_3 \leftrightarrow \neg x_4. \end{cases} \quad (32)$$

Denote $x = \bowtie_{i=1}^4 x_i$, $z = \bowtie_{i=1}^3 z_i$, 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 \mathcal{Z} is an invariant subspace of (30).

3.2 Union of Invariant Subspaces

Assume \mathcal{V}_i , $i = 1, 2$ are two M invariant subspaces, where

$$\begin{aligned}\mathcal{V}_1 &= \mathcal{F}_\ell\{z_1^1, \dots, z_p^1\}, \\ \mathcal{V}_2 &= \mathcal{F}_\ell\{z_1^2, \dots, z_q^2\}.\end{aligned}\tag{33}$$

Then we have

$$\mathcal{V}_i = G_i x, \quad i = 1, 2,\tag{34}$$

where $x = \bowtie_{i=1}^n x_i$, $G_1 \in \mathcal{L}_{2^p \times 2^n}$, $G_2 \in \mathcal{L}_{2^q \times 2^n}$.

Theorem 3.6. Assume \mathcal{V}_i , $i = 1, 2$ are M invariant subspaces. That is, there exist $H_1 \in \mathcal{L}_{p \times p}$ and $H_2 \in \mathcal{L}_{q \times q}$, such that

$$\begin{aligned}G_1 M &= H_1 G_1, \\ G_2 M &= H_2 G_2.\end{aligned}\tag{35}$$

Then

$$\mathcal{V} = \mathcal{V}_1 \bigcup \mathcal{V}_2 = \mathcal{F}_\ell\{z_1^1, \dots, z_p^1; z_1^2, \dots, z_q^2\}$$

is also M -invariant. Moreover, the structure matrix of \mathcal{V} , denoted by

$$G = G_1 * G_2,\tag{36}$$

satisfies

$$GM = HG,\tag{37}$$

where

$$H = H_1 \otimes H_2.\tag{38}$$

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

Lemma 3.7. Let $A \in \mathcal{M}_{p \times \ell}$, $B \in \mathcal{M}_{q \times \ell}$, and $T \in \mathcal{L}_{\ell \times r}$. Then

$$(A * B)T = (AT) * (BT).\tag{39}$$

Proof. Denote

$$A = [A^1, A^2, \dots, A^\ell], \quad B = [B^1, B^2, \dots, B^\ell],$$

where $A^i = \text{Col}_i(A)$ ($B^i = \text{Col}_i(B)$) is the i -th column of A (B); and

$$T = [\delta_\ell^{i_1}, \delta_\ell^{i_2}, \dots, \delta_\ell^{i_m}].$$

Then

$$\begin{aligned} (A * B)T &= ([A^1, A^2, \dots, A^\ell] * [B^1, B^2, \dots, B^\ell]) T \\ &= [A^1 \otimes B^1, A^2 \otimes B^2, \dots, A^\ell \otimes B^\ell] T \\ &= [A^{i_1} \otimes B^{i_1}, A^{i_2} \otimes B^{i_2}, \dots, A^{i_m} \otimes B^{i_m}]. \\ (AT) * (BT) &= [A^{i_1}, A^{i_2}, \dots, A^{i_m}] * [B^{i_1}, B^{i_2}, \dots, B^{i_m}] \\ &= [A^{i_1} \otimes B^{i_1}, A^{i_2} \otimes B^{i_2}, \dots, A^{i_m} \otimes B^{i_m}]. \end{aligned}$$

(39) follows immediately. \square

Proof. (of Theorem 3.6) It is enough to prove (37) with (38). Denote $G_1 = (G_1^1, \dots, G_1^{2^n})$, $G_2 = (G_2^1, \dots, G_2^{2^n})$, where $G_1^i = \text{Col}_i(G_1)$, $G_2^i = \text{Col}_i(G_2)$, $i = 1, 2, \dots, 2^n$. Using Lemma 3.7,

$$\begin{aligned} GT &= (G_1 * G_2)T = (G_1T) * (G_2T) \\ &= (H_1G_1) * (H_2G_2) \\ &= [(H_1G_1^1) * (H_2G_2^1), (H_1G_1^2) * (H_2G_2^2), \dots, (H_1G_1^{2^n}) * (H_2G_2^{2^n})] \\ &= [(H_1G_1^1) \otimes (H_2G_2^1), (H_1G_1^2) \otimes (H_2G_2^2), \dots, (H_1G_1^{2^n}) \otimes (H_2G_2^{2^n})] \\ &= [(H_1 \otimes H_2)(G_1^1 \otimes G_2^1), (H_1 \otimes H_2)(G_1^2 \otimes G_2^2), \dots, (H_1 \otimes H_2)(G_1^{2^n} \otimes G_2^{2^n})] \\ &= (H_1 \otimes H_2)(G_1 * G_2) = (H_1 \otimes H_2)G. \end{aligned}$$

\square

3.3 Dynamics of Aggregated NB

Assume (12) is a large scale BN, and z_i , $i = 1, \dots, r$ are separating logical functions, which represent our interested properties. Denote by

$$\mathcal{Z} = \mathcal{F}_\ell\{z_i \mid i = 1, \dots, r\}$$

We first try to find the smallest subspace $\overline{\mathcal{Z}}$, which contains \mathcal{Z} and is M -invariant.

Algorithm 3.8. • *Step 1:* Set $z^0 = \times_{i=1}^r z_i$, 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

$$\overline{\mathcal{Z}} := \mathcal{F}_\ell\{z^{k^*}\}. \quad (40)$$

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 $\overline{\mathcal{Z}}$ provided by (40) is the smallest subspace, containing \mathcal{Z} and is M -invariant.

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

Next, we try to find the dynamics of aggregated BN.

Assume $\overline{\mathcal{Z}} = \mathcal{F}_\ell(\bar{z})$ is a regular subspace, then

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

Using Theorem 3.3, we have that

$$\begin{aligned} \bar{z}(t+1) &= \bar{G}x(t+1) = \bar{G}Mx(t) \\ &= H\bar{G}x(t) = H\bar{z}(t). \end{aligned} \tag{41}$$

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 $\overline{\mathcal{Z}}$ has been ignored. From Algorithm 3.8 one sees easily that (27) is enough for (41). In fact, we do not care about if $\overline{\mathcal{Z}}$ 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 , $i = 1, 2, \dots, 9$ are players. Each player chooses his next opinion 1 (with whit circle) for “agree” and 0 (with block 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

$$x(t+1) = Mx(t), \tag{42}$$

where $x = \times_{i=1}^9 x_i$ and $M \in \mathcal{M}_{256 \times 256}$ is in Appendix.

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

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

Then the index function for S is defined as

$$g_1(x) = \begin{cases} 1, & x \in S, \\ 0, & \text{Otherwise.} \end{cases}$$

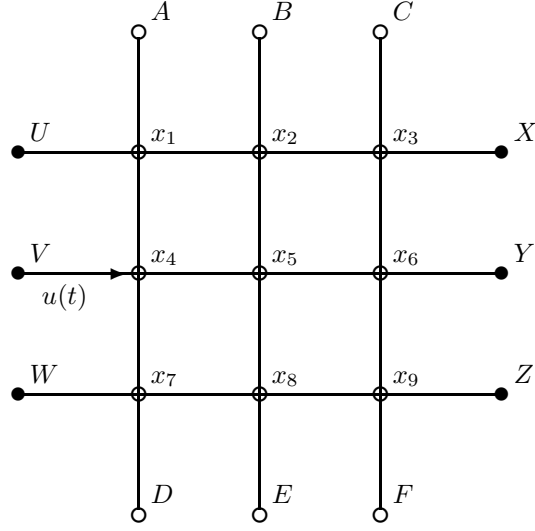


Figure 3: Social Network

Correspondingly, we have its structure matrix

$$\text{Col}_i(G_1) = \begin{cases} \delta_2^1, & \delta_{512}^i \in S, \\ \delta_2^2, & \text{Otherwise.} \end{cases}$$

Then $G_2 = G_1M$ can be expressed as

$$\text{Col}_i(G_2) = \begin{cases} \delta_2^1, & i = 22, 89, 150, 278, \\ \delta_2^2, & \text{Otherwise.} \end{cases}$$

Furthermore,

$$G_2M = G_1.$$

Set $z = z_1z_2$, where

$$z_1 = G_1x, \quad z_2 = G_3x,$$

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

$$\begin{aligned} z_1(t+1) &= G_1x(t+1) \\ &= G_1Mx(t) = G_2x(t) \\ &= z_2(t). \end{aligned}$$

$$\begin{aligned} z_2(t+1) &= G_2x(t+1) \\ &= G_2Mx(t) = G_1x(t) \\ &= z_1(t). \end{aligned}$$

Hence the smallest M invariant subspace containing g_1 is

$$G = \mathcal{F}_\ell\{g_1, g_2\}.$$

The aggregated system becomes

$$\begin{aligned} z_1(t+1) &= z_2(t) = (J_2^T \otimes I_2)z(t) \\ z_2(t+1) &= z_1(t) = (I_2 \otimes J_2^T)z(t), \end{aligned} \quad (43)$$

where $z(t) = z_1(t)z_2(t)$. Hence, the ASSR of $z(t)$ is

$$z(t+1) = [(J_2^T \otimes I_2) * (I_2 \otimes J_2^T)] z(t) = \delta_4[1, 3, 2, 4]z(t). \quad (44)$$

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

$$L = [M_1, M_2, \dots, M_{2^m}], \quad (45)$$

where

$$M_r = L\delta_{2^m}^r \in \mathcal{L}_{2^n \times 2^n}, \quad r = 1, 2, \dots, 2^m.$$

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

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

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

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

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

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

Assume $\mathcal{Z} = \mathcal{F}_\ell\{z_1, z_2, \dots, z_r\}$ and $\overline{\mathcal{Z}} = \mathcal{F}_\ell\{z_1, \dots, z_r, z_{r+1}, \dots, z_s\}$. Denote $z = \times_{i=1}^s z_i$, then there exists a $G \in \mathcal{L}_{2^s \times 2^n}$, such that

$$z = G \times_{i=1}^n x_i := Gx.$$

Since $\overline{\mathcal{Z}}$ is control invariant subspaces, for $u = \delta_{2^m}^i$ we have

$$GM_i = H_i G, \quad i = 1, 2, \dots, 2^m. \quad (46)$$

It follows that

$$\begin{aligned} z(t+1) &= Gx(t+1) = GLu(t)x(t) \\ &= [H_1, H_2, \dots, H_{2^m}]Gx(t) \\ &= [H_1, H_2, \dots, H_{2^m}]u(t)z(t) \end{aligned}$$

Define $H := [H_1, H_2, \dots, H_{2^m}]$, then we have the aggregated BCN as

$$z(t+1) = Hu(t)z(t). \quad (47)$$

Next, we consider the case when there is a constrain on control, as $u(t) \in U \subset \Delta_{2^m}$. Assume U is state-depending. That is,

$$U = \{u \neq \delta_{2^m}^\alpha \text{ if } z \in X_\alpha \subset \mathcal{X} = \Delta_{2^n} \mid \alpha \in \Xi \subset \Delta_{2^m}\}.$$

We need the following notation: $A \in \mathcal{M}_{p \times q}$ is called a zero-extended logical matrix if

$$\text{Col}(A) \subset \Delta_p \cup \mathbf{0}_p.$$

That is A may contain some zero columns.

Now consider partly control invariant subspaces containing \mathcal{Z} . Assume when $z = \delta_{2^s}^k$, $u = \delta_{2^m}^\alpha$ is forbidden. Then in equation (47) we set

$$\text{Col}_k(H_\alpha) = \mathbf{0}_{2^s}.$$

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

$$z(t+1) = H^U u(t)z(t). \quad (48)$$

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

$$x(t+1) = [N, M]u(t)x(t), \quad (49)$$

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

Then it is easy to calculate that $G_1N = G_3$, $G_3N = G_4$, $G_4N = G_5$, $G_5N = G_7$; $G_2N = G_6$, $G_6N = G_5$; $G_7M = G_7$, $G_7N = G_7$, where

$$\text{Col}_i(G_3) = \begin{cases} \delta_2^1, & i = 43, 47, 143, 164, 229, 420, \\ \delta_2^2, & \text{Otherwise.} \end{cases}$$

$$\text{Col}_i(G_4) = \begin{cases} \delta_2^1, & i = 59, 118, 278, \\ \delta_2^2, & \text{Otherwise.} \end{cases}$$

$$\text{Col}_i(G_5) = \begin{cases} \delta_2^1, & i = 164, 299, 420, \\ \delta_2^2, & \text{Otherwise.} \end{cases}$$

$$\text{Col}_i(G_6) = \begin{cases} \delta_2^1, & i = 278, \\ \delta_2^2, & \text{Otherwise.} \end{cases}$$

$$\text{Col}_i(G_7) = \delta_2^2, \quad i = 1, 2, \dots, 512.$$

Set

$$z_i = G_i x, \quad i = 1, 2, \dots, 7,$$

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

$$U = \{u(t) \neq \delta_2^2 \mid x(t) \in W\}$$

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

$$z(t+1) = H^U u(t) z(t), \tag{50}$$

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

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

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

5 Minimum Realization of BCN

Consider a BNC (16) with outputs (observers)

$$\begin{aligned} y_1(t) &= \xi_1(x_1(t), \dots, x_n(t)), \\ y_2(t) &= \xi_2(x_1(t), \dots, x_n(t)), \\ &\vdots \\ y_r(t) &= \xi_r(x_1(t), \dots, x_n(t)). \end{aligned} \tag{51}$$

Then the input-output BCN (16)-(51) has ASSR as

$$\begin{cases} x(t+1) = Lu(t)x(t), \\ y(t) = Hx(t). \end{cases} \tag{52}$$

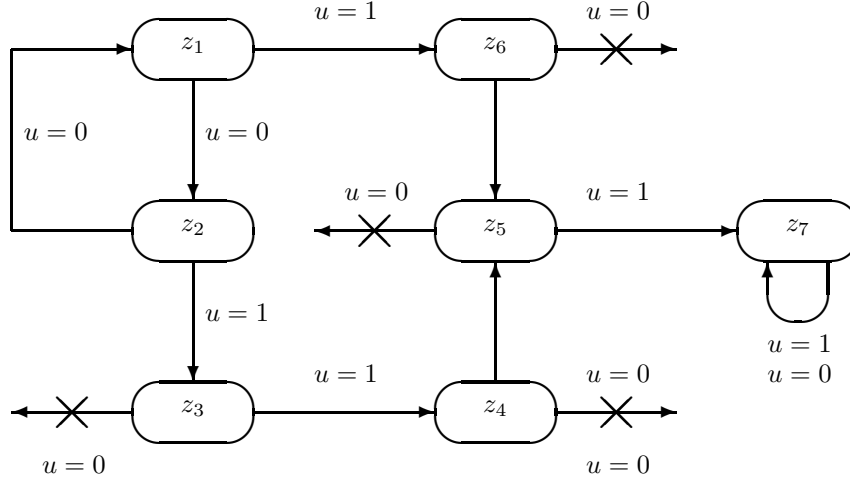


Figure 4: State Transition Graph of aggregated BCN (50))

After a coordinate change $T : x \rightarrow z$, expressed by $z = Tx$, where $T \in \mathcal{L}_{2^n \times 2^n}$, (18) becomes [3]

$$\begin{cases} z(t+1) = \tilde{L}u(t)z(t), \\ y(t) = \tilde{H}z(t), \end{cases} \quad (53)$$

where

$$\begin{aligned} \tilde{L} &= TL (I_{2^m} \otimes T^T), \\ \tilde{H} &= HT^T. \end{aligned}$$

If under the coordinate frame z (51), expressed as (53), has the form of (25), then it is clear that $\mathcal{Z} = \mathcal{F}_\ell\{z^1\}$ 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 $\mathcal{Z} = \mathcal{F}_\ell\{z_1^1, x_2^1, \dots, z_r^1\}$ such that

$$\begin{cases} z^1(t+1) = F^1(z^1, u), \\ y(t) = \xi(z^1(t)), \end{cases} \quad (54)$$

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

Remark 5.2. (i) From Definition 5.1, \mathcal{Z} is a control invariant subspace, containing \mathcal{Y} .

(ii) In Definition 5.1 \mathcal{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 $\mathcal{Z} = \mathcal{F}_\ell\{z_1^1, x_2^1, \dots, z_r^1\}$ 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 $\mathcal{Z} = \mathcal{F}_\ell\{z_1^1, x_2^1, \dots, z_r^1\}$ is the smallest control invariant subspace containing \mathcal{Y} and $\mathcal{Z} = Gx$. then

(i) there exists a set of logical matrix $H_i \in \mathcal{L}_{r \times r}$, $i = 1, 2, \dots, 2^m$ such that

$$GM_i = H_i G, \quad i = 1, 2, \dots, 2^m; \quad (55)$$

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

$$\begin{cases} z^1(t+1) = Hu(t)z^1(t), \\ y(t) = \Xi z^1(t), \end{cases} \quad (56)$$

where Ξ is the structure matrix of ξ , and

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

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

Algorithm 5.5. • *Step 1:*

Set

$$\mathcal{O}_0 = \{y_1, y_2, \dots, y_p\}.$$

Calculate

$$\mathcal{O}_1 = \{yM_1, yM_2, \dots, yM_{2^m} \mid y \in \mathcal{O}_0\} \setminus \{\mathcal{O}_0\}.$$

• *Step s: (s > 0)*

Calculate

$$\mathcal{O}_{s+1} = \{yM_1, yM_2, \dots, yM_{2^m} \mid y \in \mathcal{O}_s\} \setminus \{\mathcal{O}_r \mid r = 0, 1, \dots, s\}.$$

• *Last Step. If*

$$\mathcal{O}_{s^*+1} = \emptyset.$$

then

$$\mathcal{Z}^* := \mathcal{F}_\ell\{\mathcal{O}_r \mid r = 0, 1, \dots, s^*\}$$

is the smallest control invariant subspace containing y .

Assume $\mathcal{Z}^* = \mathcal{F}_\ell\{z_1, z_2, \dots, z_r\}$, set $z = \bowtie_{i=1}^r z_i$, then

$$\begin{aligned} z(t+1) &= [H_1, H_2, \dots, H_{2^m}] u(t) z(t), \\ y(t) &= \Xi z(t) \end{aligned} \quad (57)$$

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

Next, we consider an example.

Example 5.6. Consider a BCN, with its ASSR as

$$\begin{cases} x(t+1) = Lu(t)x(t), \\ y(t) = \Xi x(t), \end{cases} \quad (58)$$

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

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

with

$$\begin{aligned} M_1 &= \begin{pmatrix} \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} & \mathbf{0} \\ \mathbf{0} & \mathbf{X} \end{pmatrix}, & M_2 &= \begin{pmatrix} \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} & \mathbf{0} \\ \mathbf{0} & \mathbf{X} \end{pmatrix}, \\ M_3 &= \begin{pmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} & \mathbf{0} \\ \mathbf{0} & \mathbf{X} \end{pmatrix}, & M_4 &= \begin{pmatrix} \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix} & \mathbf{0} \\ \mathbf{0} & \mathbf{X} \end{pmatrix}, \end{aligned}$$

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

$$\Xi = \delta_2[1, 2, 1, \underbrace{2, 2, \dots, 2}_{2^n-3}].$$

Denote $y_1 = y$, then it is easy to calculate that

$$\begin{aligned} y_1 M_1 &= \delta_2[2, 1, 1, 2, \dots, 2] := y_2, \\ y_1 M_2 &= y_2, \\ y_1 M_3 &= y_1, \\ y_1 M_4 &= y_1, \\ y_2 M_1 &= \delta_2[1, 1, 2, 2, \dots, 2] := y_3, \\ y_2 M_2 &= y_1, \\ y_2 M_3 &= y_2, \\ y_2 M_4 &= y_3, \\ y_3 M_1 &= y_1, \\ y_3 M_2 &= y_3, \\ y_3 M_3 &= y_3, \\ y_3 M_4 &= y_2. \end{aligned}$$

Let

$$z_1 = y_1, \quad z_2 = y_2, \quad z_3 = y_3.$$

Hence we have

$$\begin{aligned} z_1(t+1) &= [Z_2, Z_2, Z_1, Z_1]u(t)z(t), \\ z_2(t+1) &= [Z_3, Z_1, Z_2, Z_3]u(t)z(t), \\ z_3(t+1) &= [Z_1, Z_3, Z_3, Z_2]u(t)z(t), \end{aligned}$$

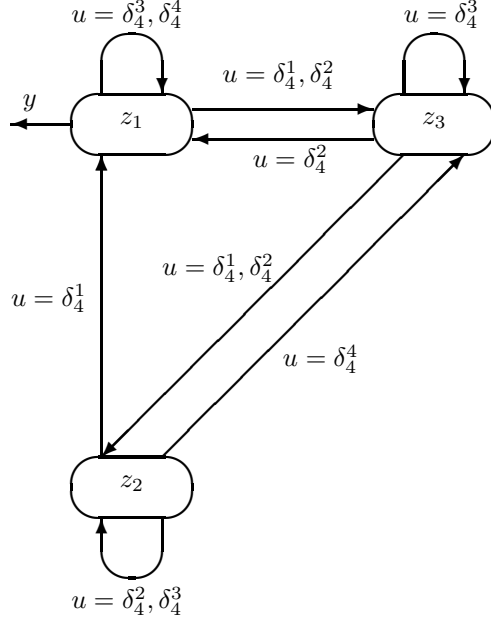


Figure 5: State Transition Graph of aggregated BCN (59))

where $u(t) = u_1(t)u_2(t)$, $z(t) = z_1(t)z_2(t)z_3(t)$, and

$$\begin{aligned} Z_1 &= I_2 \otimes J_4^T = \delta_2[1, 1, 1, 1, 2, 2, 2, 2], \\ Z_2 &= J_2 \otimes I_2 \otimes J_2 = \delta_2[1, 1, 2, 2, 1, 1, 2, 2], \\ Z_3 &= J_4 \otimes I_2 = \delta_2[1, 2, 1, 2, 1, 2, 1, 2]. \end{aligned}$$

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

$$\begin{cases} z(t+1) = L^*u(t)z(t), \\ y(t) = Z_1z(t), \end{cases} \quad (59)$$

where

$$L^* = \delta_8 \begin{bmatrix} 1, 3, 5, 7, 2, 4, 6, 8, 1, 2, 5, 6, 3, 4, 7, 8, \\ 1, 2, 3, 4, 5, 6, 7, 8, 1, 3, 2, 4, 5, 7, 6, 8 \end{bmatrix}.$$

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

$$TM_iT^T = \begin{bmatrix} J_1^i & \mathbf{0} & \cdots & \mathbf{0} \\ \mathbf{0} & J_2^i & \cdots & \mathbf{0} \\ & & \ddots & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \cdots & J_s^i \end{bmatrix}, \quad i = 1, 2, \dots, 2^m,$$

where $z = (z^1, z^2, \dots, z^s)$ and z^k corresponds to J_k^i . Moreover, if $y \in \mathcal{F}_\ell\{z^k\}$, then there exists a realization

$$\begin{cases} z(t+1) = [J_k^1, J_k^2, \dots, J_k^{2^m}] u(t)z(t), \\ y(t) = \Xi_k z(t). \end{cases} \quad (60)$$

Moreover, if J_k^i , $i = 1, 2, \dots, 2^m$ can not be further diagonalized 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.

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7 Appendix

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

$$M = \delta_{512} [$$

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	73	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	5	8	65	74	82	90	65	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	218	220	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	506	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	505	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]

(ii) Structure matrix for Example 4.3.

$N = \delta_{512}[$	1	1	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	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	74	74	73	74	78	80
	9	9	9	12	9	11	15	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
	9	9	29	32	45	47	63	64	89	90	94	96	125	128	128	128
	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
	201	201	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	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
	257	257	261	262	293	293	293	296	257	266	278	286	293	302	310	320
	289	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	429	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
	385	393	401	412	417	427	439	448	473	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	476	505	508	512	512
	449	457	469	478	485	493	501	512	473	474	478	478	509	510	510	512
	505	505	509	512	509	511	511	512	505	506	510	512	509	512	512	512].