Semi-Tensor Product Approach to Boolean Functions^{*}

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<u>Abstract.</u> Boolean function can be expressed as matrix form using semi-tensor product of matrices. Using this approach, we give a neat proof of the conversion between the truth table and polynomial form of a Boolean function. The linear structure of Boolean functions is also investigated.

Key Words. Boolean function, semi-tensor product, truth table, polynomial form, linear structure.

1 Introduction

Boolean function (or logical function) is a basic concept in Boolean algebra, which plays a fundamental role in computer sciences, circuit design, cryptography, etc. [1, 2, 3, 4]. In investigation of cellular networks, Kauffman proposed the Boolean network, which is a dynamic system consisting of Boolean functions [5]. It has then become a very useful tool in modeling of cell regulation [6, 7]. Boolean function can also be used in game theory [8, 9, 10], it is called a simple game or voting game, by which the social choice (e.g. the election) can be analyzed.

Denoting $\mathcal{D} = \{0, 1\}$, an *n*-ary Boolean function is a function $f : \mathcal{D}^n \to \mathcal{D}$. There are many ways to express a Boolean function, among them the truth table is the most natural one. The polynomial expression and Walsh spectral expression are two of the most useful tools in the analysis of Boolean function. There are also some graphic expressions which are more efficient in computing, such as binary decision diagrams[11] and propositional directed acyclic graphs[12]. The conversion among different expressions is a fundamental and challenging topic.

Different applications invoke different properties of Boolean functions. Linearity is a basic property. In the design of circuits, linear Boolean functions are widely used, since they are easily realizable. But the linearity is a critical weakness in cryptography for the functions with linear structure are easily breakable [13], the ones with high nonlinearity are preferred. Thus, it is important to check whether a Boolean function has a linear structure.

Recently, using the semi-tensor product of matrices, a Boolean function can be expressed in a matrix form[14, 15]. In this paper, using the matrix form of Boolean function, we give a formula for the conversion between the truth table and the polynomial expression of Boolean functions and propose a way to find out the linear structure of a Boolean function.

The rest of the paper is organized as follows: Section 2 reviews the polynomial expression and the matrix form of Boolean function. Using the matrix form, a formula converting the truth table to the

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polynomial expression is obtained in Section 3, which is essentially the same as the known result [3], but neat proof is provided by using semi-tensor product. Section 4 investigates the linear structure of Boolean networks. Section 5 is a brief conclusion.

2 Expressions of Boolean Functions

To introduce the matrix expression of Boolean functions, we first briefly review the semi-tensor product (STP) of matrices.

Definition 2.1 [14] Let $A \in \mathcal{M}_{m \times n}$, $B \in \mathcal{M}_{p \times q}$, and c = lcm(n, p) (the least common multiple of n and p). Then the semi-tensor product (STP) of matrix A and B, denoted by $A \ltimes B$, is defined as

$$A \ltimes B = (A \otimes I_{\underline{c}})(B \otimes I_{\underline{c}}), \tag{1}$$

where " \otimes " is Kronecher product.

When n = p, it is easy to see that $A \ltimes B = AB$, and hence " \ltimes " will be omitted hereafter. For statement ease, we introduce some notations.

Notations:

- (i) Let δ_n^i be the *i*-th column of the identity matrix I_n , and $\Delta_n := \{\delta_n^1, \delta_n^2, \dots, \delta_n^n\}$. When n = 2 we simply use $\Delta := \Delta_2$.
- (ii) Denote by $\operatorname{Col}(A)$ (Row(A)) the set of columns (rows) of A, and $\operatorname{Col}_i(A)$ (Row_i(A)) the *i*-th column (row) of A.
- (iii) Assume a matrix $L = [\delta_n^{i_1} \delta_n^{i_2} \cdots \delta_n^{i_s}] \in \mathcal{M}_{n \times s}$, i.e., its columns, $\operatorname{Col}(L) \subset \Delta_n$. We call L a logical matrix, and simply denote it as

$$L = \delta_n [i_1 \ i_2 \ \cdots \ i_s].$$

The set of $n \times s$ logical matrices is denoted by $\mathcal{L}_{n \times s}$.

(iv) $W_{[n,m]}$ is a swap matrix which can swap two vectors in their "product". We refer to [14] for its definition and properties, and (2) for its basic functions.

One advantage of the STP is pseudo-commutativity. This property is important for obtaining the matrix form of Boolean functions. The following proposition about pseudo-commutativity and reducing matrix will be frequently used in this paper.

Proposition 2.2 [15]

1. Let $x \in \mathbb{R}^m$ and $y \in \mathbb{R}^n$ be two column vectors. Then

$$W_{[m,n]}xy = yx. (2)$$

2. Let $x \in \mathbb{R}^t$ and A is a given matrix. Then

$$xA = (I_t \otimes A)x. \tag{3}$$

3. Let $x \in \Delta_n$. Then $x^2 = M_r^n x$, where

$$M_r^n := \operatorname{diag}[\delta_n^1 \ \delta_n^2 \ \cdots \ \delta_n^n]$$

is call the base-n order reducing matrix.

Then, we consider the expression of elements in \mathcal{D}^n . We propose the following three ways to express it.

(i) Component-wise (C-W) Form:

$$X = (x_1, x_2, \cdots, x_n), \quad x_i \in \mathcal{D}, \ i = 1, \cdots, n.$$

$$\tag{4}$$

(ii) Scalar Form: Consider $x_1 x_2 \cdots x_n$ as a binary number. Then in decimal form we have a number as

$$\chi = x_1 2^{n-1} + x_2 2^{n-2} + \dots + x_n, \tag{5}$$

where $0 \le \chi \le 2^n - 1$.

(iii) Vector form: Identify $1 \sim \delta_2^1$ and $0 \sim \delta_2^2$, then $x_i \in \Delta_2$ and we set

$$x := \ltimes_{i=1}^{n} x_i \in \Delta_{2^n}.$$
(6)

The component-wise form can be considered as the binary representation of the scalar form, thus they are equivalent. The vector form is obtained from the component-wise form. Then, the following conversion between scalar form and vector form ensures that the three expressions of elements in \mathcal{D}^s are equivalent. Thus, in the sequel we will not distinct them if there is no possible confusion.

Proposition 2.3 Let χ be a scalar form of $x \in \Delta_{2^n}$. Then

$$x = \delta_{2^n}^{2^n - \chi}.\tag{7}$$

Equivalently, let $x = \delta_{2^n}^t$. Then

$$\chi = 2^n - t. \tag{8}$$

Using the definitions and Proposition 2.3, it is easy to convert an element in \mathcal{D}^n from one form to another. We give an example for this.

Example 2.4 Let n = 8. Then

A natural way to express a Boolean function f is to range the values of all the elements in \mathcal{D}^n in a vector:

$$[f(\delta_{2^n}^1), f(\delta_{2^n}^2), \cdots, f(\delta_{2^n}^{2^n})]^T.$$
(9)

That is the truth table of f.

The addition \oplus and the multiplication \odot over \mathcal{D} is defined as

$$\begin{cases} a \oplus b := a + b \pmod{2} \\ a \odot b := ab \pmod{2}. \end{cases}$$
(10)

Then $GF(2) := \{\mathcal{D}, \oplus, \odot\}$ forms a field, call the Galois Field. In fact, \oplus and \odot are logical operators $\bar{\vee}$ and \wedge respectively. For the sake of compactness, we simply denote $a \oplus b = a + b$, and $a \odot b = ab$.

For $x \in \mathcal{D}$, denote $x^1 := x$, $x^0 = \neg x$. Then, it is obvious that for $c \in \mathcal{D}$,

$$x^{c} = \begin{cases} 1, & x = c \\ 0, & x \neq c. \end{cases}$$
(11)

For $X = (x_1, x_2, \cdots, x_n) \in \mathcal{D}^n$ and $C = (c_1, c_2, \cdots, c_n) \in \mathcal{D}^n$, define

$$X^{C} := \prod_{i=1}^{n} x_{i}^{c_{i}} = \begin{cases} 1, & X = C \\ 0, & X \neq C. \end{cases}$$
(12)

Hence, it is obvious that

$$f(X) = \sum_{C=0}^{2^n - 1} f(C) X^C.$$
(13)

Replacing x_i^1 by x_i , and x_i^0 by $x_i + 1$ in (13), the Boolean function f can be expressed as a polynomial in GF(2) as

$$f(x) = a_0 + a_1 x_1 + \dots + a_n x_n + a_{12} x_1 x_2 + \dots + a_{n-1 n} x_{n-1} x_n + \dots + a_{12 \dots n} x_1 x_2 \dots x_n$$

$$= a_0 + \sum_{k=1}^n \sum_{1 \le j_1 < \dots < j_k \le n} a_{j_1 \dots j_k} x_{j_1} \dots x_{j_k}.$$
(14)

The following theorem about the matrix form of Boolean functions was proved in [14].

Theorem 2.5 Let $f : \mathcal{D}^n \to \mathcal{D}$ be a Boolean function. Then there exists a unique logical matrix $M_f \in \mathcal{L}_{2 \times 2^n}$, called the structure matrix of f, such that in vector form f can be expressed as

$$f(x_1, \cdots, x_n) = M_f \ltimes_{i=1}^n x_i, \quad x_i \in \Delta.$$
(15)

The following table is the structure matrices of some logical operators.

Table	e 1: Structu	re Matrices of Operators
	Operator	Structure Matrix
-	-	$M_{\neg} = \delta_2[2\ 1]$
	\wedge	$M_{\wedge} = \delta_2 [1 \ 2 \ 2 \ 2]$
	\vee	$M_{\vee} = \delta_2 [1 \ 1 \ 1 \ 2]$
	\rightarrow	$M_{\rightarrow} = \delta_2 [1 \ 2 \ 1 \ 1]$
	\leftrightarrow	$M_{\leftrightarrow} = \delta_2 [1 \ 2 \ 2 \ 1]$
	$\bar{\lor}(\mathrm{or}\oplus)$	$M_{\oplus} = \delta_2 [2 \ 1 \ 1 \ 2]$

Remark 2.6 Since the structure matrix is a logical matrix, it can be totally determined by its first row:

$$m_f = \operatorname{Row}_1(M_f).$$

Actually, m_f is the truth table of Boolean function f.

We give an example to depict these expressions.

Example 2.7 Consider a Boolean function

$$f(x_1, x_2, x_3) = x_1 \land \neg (x_2 \to x_3).$$

In vector form, we have

$$\begin{aligned} f(x_1, x_2, x_3) = & M_{\wedge} x_1 M_{\neg} M_{\rightarrow} x_2 x_3 \\ = & M_{\wedge} (I_2 \otimes M_{\neg} M_{\rightarrow}) x_1 x_2 x_3 \\ := & M_f x_1 x_2 x_3 = \delta_2 [2 \ 1 \ 2 \ 2 \ 2 \ 2 \ 2 \ 2 \ 2 \] x_1 x_2 x_3. \end{aligned}$$

It is easy to check that

$$m_f = \operatorname{Row}_1(M_f) = [0\ 1\ 0\ 0\ 0\ 0\ 0]$$

is the truth table of f.

Then, by (13), the polynomial expression of f is

$$f(x_1, x_2, x_3) = x_1^1 x_2^1 x_3^0 = x_1 x_2 (x_3 + 1) = x_1 x_2 + x_1 x_2 x_3.$$

Walsh spectral expression is also a very useful expression of Boolean functions, and there are also some graph expressions. But they are beyond the scape of this paper, thus we do not discuss them here.

3 Conversion Between Truth Table and Polynomial Expression

Note that in vector form,

$$x_i \sim \begin{pmatrix} x_i^1 \\ x_i^0 \end{pmatrix} = \begin{pmatrix} x_i \\ x_i+1 \end{pmatrix},$$

and for an $n \times m$ matrix A, $A(I_m \otimes B) = A \otimes B$. Then by the matrix form (15) of Boolean function f, we have

$$f(x) = m_f \begin{pmatrix} x_1 \\ x_1 + 1 \end{pmatrix} \begin{pmatrix} x_2 \\ x_2 + 1 \end{pmatrix} \cdots \begin{pmatrix} x_n \\ x_n + 1 \end{pmatrix}$$
$$= m_f \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{pmatrix} 1 \\ x_1 \end{pmatrix} \cdots \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{pmatrix} 1 \\ x_n \end{pmatrix}$$
$$= m_f \left(\underbrace{\begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \otimes \cdots \otimes \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}}_{n} \right) \begin{pmatrix} 1 \\ x_1 \end{pmatrix} \begin{pmatrix} 1 \\ x_2 \end{pmatrix} \cdots \begin{pmatrix} 1 \\ x_n \end{pmatrix}$$
$$:= m_f P_n \xi_n,$$

where

$$P_n = \left(\underbrace{\begin{bmatrix} 0 & 1\\ 1 & 1 \end{bmatrix}}_{n} \otimes \dots \otimes \begin{bmatrix} 0 & 1\\ 1 & 1 \end{bmatrix}_{n}\right),\tag{16}$$

and

$$\xi_n = \begin{pmatrix} 1\\ x_1 \end{pmatrix} \begin{pmatrix} 1\\ x_2 \end{pmatrix} \cdots \begin{pmatrix} 1\\ x_n \end{pmatrix} = (1, x_n, x_{n-1}, x_{n-1}x_n, x_{n-2}, \cdots, x_1x_2 \cdots x_n)^T$$
(17)

is a basis of the polynomials on $GF^n(2)$. Then $m_f P_n \xi_n$ is already a polynomial.

Alternatively, the standard polynomial expression (14) of a Boolean function f can be written as

$$f(x) = \beta \eta_n \tag{18}$$

where

$$\eta_n = (1, x_1, \cdots, x_n, x_1 x_2, \cdots, x_{n-1} x_n, \cdots, x_1 x_2 \cdots x_n)^T$$

arranged as a natural alphabetic and power increasing order. To convert $m_f P_n \xi_n$ into the standard polynomial expression, we just need to reorder the entries in ξ_n .

Lemma 3.1 Let μ_{i_1,i_2,\cdots,i_r} , $i_1 < i_2 \cdots < i_t$ be the positions where $x_{i_1}x_{i_2}\cdots x_{i_t}$ appear in ξ_n . Then

$$\mu_{i_1, i_2, \cdots, i_r} = \sum_{j=1}^r 2^{n-i_j} + 1.$$
(19)

Proof. By direct computation, we know that for any j, the position where x_{n-j} appears for the first time is $\mu_{n-j}=2^j+1$. And then $x_{n-j}x_{n-k}$ appears at 2^k after x_{n-k} , thus its position is 2^j+2^k+1 .

Then for any $x_{n-i_1}x_{n-i_2}\cdots x_{n-i_t}$, we can locate $x_{n-i_1}, x_{n-i_1}x_{n-i_2}, \cdots, x_{n-i_1}x_{n-i_2}\cdots x_{n-i_t}$ one after another in a sequence, and in that way the conclusion follows.

Using Lemma 3.1, we construct Φ_n as follows

$$\Phi_n = \delta_{2^n} [1, \phi_1, \phi_2, \cdots, \phi_n], \tag{20}$$

where $\phi_r = (\mu_{1,2,\dots,r}, \mu_{2,\dots,r+1}, \dots, \mu_{n-r+1,n-r+2,\dots,n}), \quad r = 1, 2, \dots, n$. Then one can check easily that

$$\xi_n = \Phi_n \eta_n.$$

Note that $\Phi_n^{-1} = \Phi_n^T$, the following theorem is obvious.

Theorem 3.2 Assume m_f is the truth table of a Boolean function f, and the corresponding standard polynomial expression is (18). Then we have

$$\beta = m_f P_n \Phi_n$$

$$m_f = \beta \Phi_n^T P_n^{-1}.$$
(21)

We give an example to depict this.

Example 3.3 Recall Example 2.7. The truth table of f is

$$m_f = [0\ 1\ 0\ 0\ 0\ 0\ 0]$$

By straightforward computation we have

$$P_{3} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \otimes \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \otimes \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 1 & 0 & 0 & 0 & 1 & 1 \\ 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}.$$

Then

$$\begin{aligned} f(x) &= m_f P_3 \begin{pmatrix} 1\\ x_1 \end{pmatrix} \begin{pmatrix} 1\\ x_2 \end{pmatrix} \begin{pmatrix} 1\\ x_3 \end{pmatrix} \\ &= [0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 1 \ 1] (1, x_3, x_2, x_2 x_3, x_1, x_1 x_3, x_1 x_2, x_1 x_2 x_3)^T \\ &= x_1 x_2 + x_1 x_2 x_3. \end{aligned}$$

It is the same as the polynomial expression obtained in Example 2.7.

4 Linearity of Boolean Functions

In this section, we consider the linearity of Boolean functions. Since Boolean functions can be expressed in polynomial form, the Boolean functions with only linear terms are called the linear Boolean functions. The linear structure is a generalized linearity. It was pointed out in introduction that the existence of linear structures is a weakness in cryptography. We first give the rigorous definition of linear structure.

Definition 4.1 Let $f : \mathcal{D}^n \to \mathcal{D}$ be a Boolean function.

- 1. $a \in \mathcal{D}^n$ is called an invariant linear structure (ILS) of f, if f(x+a) + f(x) = 0.
- 2. $a \in \mathcal{D}^n$ is called a variant linear structure (VLS) of f, if f(x+a) + f(x) = 1.
- 3. Denote by

$$E_{0} := \{a \in \mathcal{D}^{n} \mid f(x+a) + f(x) = 0\}$$

$$E_{1} := \{a \in \mathcal{D}^{n} \mid f(x+a) + f(x) = 1\}$$

$$E := E_{0} \cup E_{1}.$$
(22)

Then E is called the linear structure subspace of f.

4. If $E \neq \{0\}$, f is called a Boolean function with linear structure (BFLS). For a BFLS, if $E_0 \neq \{0\}$, it is said to be of type I, otherwise, it is said to be of type II.

Though there have already been many efficient algorithms to check wether a Boolean function have linear structure[13], we can give a precise formula to calculate E_0 and E_1 . Let $f : \mathcal{D}^n \to \mathcal{D}$ be a Boolean function with its structure matrix $M_f \in \mathcal{L}_{2 \times 2^n}$. Denote by $a = \ltimes_{i=1}^n a_i, x = \ltimes_{i=1}^n x_i$. Then it is easy to see that $(a_1, \dots, a_n) \in E_0$, if and only if

$$M_f M_{\oplus} a_1 x_1 M_{\oplus} a_2 x_2 \cdots M_{\oplus} a_n x_n = M_f x_1 x_2 \cdots x_n.$$

$$\tag{23}$$

A straightforward computation shows that (23) is equivalent to

$$M_f M_{\oplus} \ltimes_{i=1}^{n-1} (I_{2^{2i}} \otimes M_{\oplus}) \ltimes_{i=1}^{n-1} (I_{2^i} \otimes W_{[2,2^i]}) ax = M_f x.$$
(24)

Define

$$\Psi_f := M_f M_{\oplus} \ltimes_{i=1}^{n-1} (I_{2^{2i}} \otimes M_{\oplus}) \ltimes_{i=1}^{n-1} (I_{2^i} \otimes W_{[2,2^i]}),$$

where $M_{\oplus} \ltimes_{i=1}^{n-1} (I_{2^{2i}} \otimes M_{\oplus}) \ltimes_{i=1}^{n-1} (I_{2^i} \otimes W_{[2,2^i]})$ is a constant matrix depends only on n. Split Ψ_f into 2^n blocks as

$$\Psi_f = [\psi_1 \ \psi_2 \ \cdots \ \psi_{2^n}],$$

where $\psi_k = \text{Blk}_k(\Psi_f), k = 1, 2, \dots, 2^n$. Then the following result is straightforward verifiable.

Proposition 4.2 Let $\alpha = \ltimes_{i=1}^{n} a_i = \delta_{2^n}^i$. Then $(a_1, \dots, a_n) \in E_0$, iff $\psi_i = M_f$. $(a_1, \dots, a_n) \in E_1$, if and only if $\psi_i = M_{\neg}M_f$.

Example 4.3 Recall Example 2.7. From the polynomial form

$$f(x) = x_1 x_2 + x_1 x_2 x_3,$$

one sees easily that it is not linear Boolean function. Next, we will check whether it has a linear structure. Since

$$M_f = \delta_2 [2 \ 1 \ 2 \ 2 \ 2 \ 2 \ 2],$$

and

$$M_{\neg}M_f = \delta_2[1\ 2\ 1\ 1\ 1\ 1\ 1],$$

then

$\Psi_f = \delta_2[2$	2	2	2	2	2	1	2	2	2	2	2	2	2	2	1
2	2	2	2	1	2	2	2	2	2	2	2	2	1	2	2
2	2	1	2	2	2	2	2	2	2	2	1	2	2	2	2
1	2	2	2	2	2	2	2	2	1	2	2	2	2	2	2].

Since only $\psi_8 = M_f$, we know that $E_0 = \{\mathbf{0}\}$ and $E_1 = \emptyset$, f has no linear structure.

5 Conclusion

Using semi-tensor product of matrices, Boolean functions can be expressed in matrix form. This paper showed that it is more convenient to convert a Boolean functions truth table into its polynomial expression using the semi-tensor product approach. The formula to find the linear structure is also given in the matrix form. The semi-tensor product approach is a new method to represent Boolean functions, this paper only investigates a few usages of this approach. We believe that more properties of Boolean functions can be revealed using this approach in the future.

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布尔函数的半张量积方法

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摘要: 利用矩阵的半张量积,布尔函数可以被表示为矩阵形式.通过这个方法,我们给出了布尔函数 从真值表到多项式形式转换的一个简洁的证明.并且研究了布尔函数的线性结构. **关键词:** 布尔函数,半张量积,真值表,多项式表示,线性结构.