

Outer-Product Matrix Representation of Optical Orthogonal Codes

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Abstract—In this letter, we introduce a whole new approach in defining and representing optical orthogonal codes (OOCs), namely, outer-product matrix representation. Instead of applying commonly used approaches based on inner product to construct OOC codes, we use the newly defined approach to obtain a more efficient algorithm in constructing and generating OOC codes. The outer-product matrix approach can obtain a family of OOC codes with a cardinality closer to the Johnson upper bound, when compared with the previously defined accelerated greedy algorithm using the inner-product approach. We believe the new look introduced in this letter on OOCs could help to devise new approaches in designing and generating OOC codes, using the rich literature in matrix algebra.

Index Terms—Design techniques, optical code-division multiple access (CDMA), optical networks, optical orthogonal codes (OOCs).

I. INTRODUCTION

EVER since the introduction of important optical orthogonal codes (OOCs) and their use in optical code-division multiple-access (CDMA) systems [1]–[7], a trend has been set to explore efficient algorithms in constructing and generating the above codes by mathematicians and sequence designers [8]–[12]. Up until now, most approaches were based on the initial inner-product definition of OOCs. Using the above definition, most algorithmic designs would follow the mathematical languages of combinatorics, graph theory, Steiner numbers, etc.

However, in this letter, we present a new approach based on outer-product matrices of OOCs. This approach, we believe, will give, for the first time, the mathematical language based on matrix algebra, with its tremendous body of literature, to possibly develop a more efficient algorithm in constructing and generating OOC codes.

A major weakness with most combinatorics or projective geometry-based algorithms is that they can only support certain code lengths, code weights, and auto- and cross-correlation integer values that have special properties. In other words, most algorithms developed so far can not be easily extended to generalize for all integers of choice on code length, weight, and auto- and cross-correlation constraints. We believe by shifting the fundamental definitions and properties of OOC from the inner-product to outer-product matrix may help to remove the above bottleneck in designing and generating OOCs.

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In Section II of this paper, we first present the basic definition of OOCs based on outer-product matrix representation, then discuss the fundamental properties of OOCs based on the above matrices. In Section III, we present the algebraic matrix properties of representing OOCs. In Section IV, we present the first algorithm in constructing and generating OOC codes based on their matrix representation, which proves to have larger cardinality when compared with its inner-product-based counterpart, namely, the accelerated greedy algorithm [3].

II. OUTER-PRODUCT MATRIX REPRESENTATION

Up until now [1]–[12], OOCs have been defined based on the inner product of two code vectors. According to [3], an $(n, w, \lambda_a, \lambda_c)$ OOC C is a collection of binary n -tuples, each of Hamming weight w . If we denote t to be the t th cyclic shift of code vector \mathbf{x} , i.e., $\mathbf{x}(t)$, then the two needed fundamental properties can be written as follows.

- 1) Auto-correlation: For any $\mathbf{x} = (x_0x_1 \cdots x_{n-1}) \in C$ and any integer $0 < \tau < n$

$$\mathbf{x}(t) \cdot \mathbf{x}(t + \tau) = \mathbf{x}(t)\mathbf{x}^T(t + \tau) = \sum_{t=0}^{n-1} \mathbf{x}_t\mathbf{x}_{t \oplus \tau} \leq \lambda_a. \quad (1)$$

- 2) Cross-correlation: For any $\mathbf{x} = (x_0x_1 \cdots x_{n-1}) \in C$ and $\mathbf{y} = (y_0y_1 \cdots y_{n-1}) \in C$, such that $\mathbf{x} \neq \mathbf{y}$ and any integer τ

$$\mathbf{x}(t) \cdot \mathbf{y}(t + \tau) = \mathbf{x}(t)\mathbf{y}^T(t + \tau) = \sum_{t=0}^{n-1} \mathbf{x}_t\mathbf{y}_{t \oplus \tau} \leq \lambda_c. \quad (2)$$

Since OOCs were defined in terms of periodic correlation, the addition in the subscripts above, denoted “ \oplus ,” is modulo- n , “ \cdot ” denotes the inner product of two code vectors, and superscript “ T ” implies transpose. In general, to investigate the above two properties using the inner-product approach, the inner products should be calculated for various τ .

However, the definition of OOCs can also be presented based on the outer product of two code vectors, which results in matrix representation and crystallizes many approaches based on matrix algebra to find new design techniques for OOCs.

The first matrix we consider is obtained by the outer product of a vector with itself. For this purpose, we consider each code-word \mathbf{x} as a vector of length n (or a $1 \times n$ matrix), for which the related $n \times n$ matrix will be

$$\mathbf{D}_{\mathbf{xx}} = [d_{ij}]_{n \times n} = \mathbf{x} \times \mathbf{x} = \mathbf{x}^T \mathbf{x} \quad (3)$$

where “ \times ” means the outer product of two vectors, and

$$d_{ij} = x_i x_j = \begin{cases} 1, & x_i = x_j = 1 \\ 0, & \text{otherwise.} \end{cases}$$

Example 1: Let $n = 5$, $w = 2$, $\lambda_a = \lambda_c = 1$, and $\mathbf{x} = (10100)$; then

$$\mathbf{D}_{\mathbf{xx}} = \mathbf{x}^T \mathbf{x} = \begin{pmatrix} 1 \\ 0 \\ 1 \\ 0 \\ 0 \end{pmatrix} (1 \ 0 \ 1 \ 0 \ 0) = \begin{pmatrix} 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}.$$

The second matrix considered is the outer product of two different code vectors. For any two code vectors $\mathbf{x} = (x_0 x_1 \cdots x_{n-1})$ and $\mathbf{y} = (y_0 y_1 \cdots y_{n-1})$, $x_i, y_i \in \{0, 1\}$, we associate an $n \times n$ matrix

$$\mathbf{D}_{\mathbf{xy}} = [d_{ij}]_{n \times n} = \mathbf{x} \times \mathbf{y} = \mathbf{x}^T \mathbf{y} \quad (4)$$

where

$$d_{ij} = x_i y_j = \begin{cases} 1, & x_i = y_j = 1 \\ 0, & \text{otherwise.} \end{cases}$$

Example 2: Assume a $(5, 2, 1, 1)$ OOC with two code vectors $\mathbf{x} = (10100)$ and $\mathbf{y} = (11000)$; then

$$\mathbf{D}_{\mathbf{xy}} = \mathbf{x}^T \mathbf{y} = \begin{pmatrix} 1 \\ 0 \\ 1 \\ 0 \\ 0 \end{pmatrix} (1 \ 1 \ 0 \ 0 \ 0) = \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}.$$

III. PROPERTIES OF OUTER-PRODUCT MATRIX REPRESENTATION

With the above definitions the two matrices $\mathbf{D}_{\mathbf{xx}}$ and $\mathbf{D}_{\mathbf{xy}}$ will have the following three properties:

- 1) the matrices $\mathbf{D}_{\mathbf{xx}}$ and $\mathbf{D}_{\mathbf{xy}}$ are $(0, 1)$ matrices;
- 2) the matrix $\mathbf{D}_{\mathbf{xx}}$ is symmetric;
- 3) in the matrices $\mathbf{D}_{\mathbf{xx}}$ and $\mathbf{D}_{\mathbf{xy}}$, the nonzero columns (rows) are equal to each other.

In the following, we show how the two needed fundamental properties of OOC codes can be captured and represented at once for various τ using outer-product matrices.

Definition: Assume a matrix $\mathbf{A} = [a_{ij}]_{n \times n}$. In this matrix, the elements a_{ij} , where $i = j$, are located on the main diagonal, and the elements which have equal $(j - i)_{\text{mod}n}$ are defined as the “ $(j - i)_{\text{mod}n}$ th diagonal.” In Fig. 1, five elements ($n = 5$) located on the second diagonal of a 5×5 array are shown.

Lemma 1: The value of the auto-correlation function of code vector \mathbf{x} for the τ th cyclic shift is equal to the summation of the elements of the τ th diagonal in outer-product matrix $\mathbf{D}_{\mathbf{xx}}$.

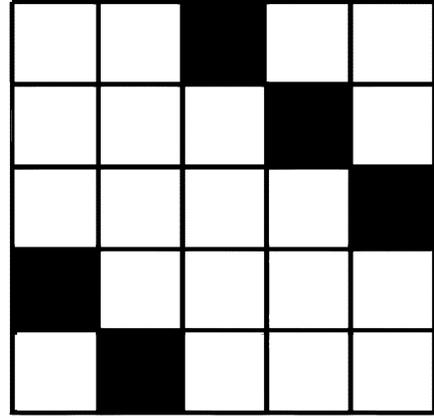


Fig. 1. Example of a 5×5 matrix with its second diagonal shown by black squares.

Proof: If d_{ij} is located on the τ th diagonal, then $(j - i)_{\text{mod}n} = \tau$, or $j = i \oplus \tau$. Therefore

$$\mathbf{S}_{\mathbf{xx}}(\tau) = \sum_{d_{ij} \in \tau \text{th diagonal of } \mathbf{D}_{\mathbf{xx}}} d_{ij} = \sum_{i=0}^{n-1} x_i x_j = \sum_{i=0}^{n-1} x_i x_{i \oplus \tau}$$

which is equal to the summation in (1).

Lemma 2: The amount of cross-correlation function of two code vectors \mathbf{x} and \mathbf{y} for the τ th cyclic shift is equal to the summation of the elements of the τ th diagonal in outer-product matrix $\mathbf{D}_{\mathbf{xy}}$.

Proof: Similar to *Lemma 1*

$$\mathbf{S}_{\mathbf{xy}}(\tau) = \sum_{d_{ij} \in \tau \text{th diagonal of } \mathbf{D}_{\mathbf{xy}}} d_{ij} = \sum_{i=0}^{n-1} x_i y_j = \sum_{i=0}^{n-1} x_i y_{i \oplus \tau}$$

which is equal to the summation in (2).

Example 3: Consider the two code vectors of *Example 2*. Fig. 2 shows the auto- and cross-correlation functions of \mathbf{x} and \mathbf{y} . We have also three matrices

$$\mathbf{D}_{\mathbf{xx}} = \begin{pmatrix} 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$$\mathbf{D}_{\mathbf{yy}} = \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$$\mathbf{D}_{\mathbf{xy}} = \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}.$$

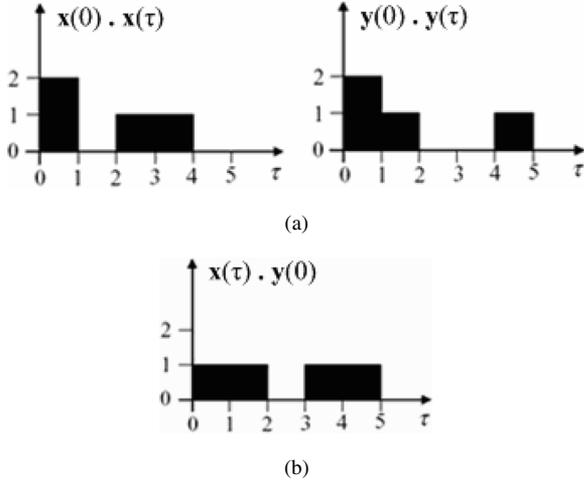


Fig. 2. (a) Auto-correlation functions of Example 3. (b) Cross-correlation function of Example 3.

Summations of elements in various diagonals of matrices are

$$\begin{aligned} \{S_{\mathbf{x}\mathbf{x}}(0), S_{\mathbf{x}\mathbf{x}}(1), \dots, S_{\mathbf{x}\mathbf{x}}(4)\} &= \{2, 0, 1, 1, 0\} \\ \{S_{\mathbf{y}\mathbf{y}}(0), S_{\mathbf{y}\mathbf{y}}(1), \dots, S_{\mathbf{y}\mathbf{y}}(4)\} &= \{2, 1, 0, 0, 1\} \\ \{S_{\mathbf{x}\mathbf{y}}(0), S_{\mathbf{x}\mathbf{y}}(1), \dots, S_{\mathbf{x}\mathbf{y}}(4)\} &= \{1, 1, 0, 1, 1\}. \end{aligned}$$

We see that the elements of these sets are equal to the correlation functions shown in Fig. 2.

We represent two other useful matrices $\mathbf{C}_{\mathbf{x}\mathbf{x}}$ and $\mathbf{C}_{\mathbf{x}\mathbf{y}}$. For a single code vector \mathbf{x} , we consider $\mathbf{C}_{\mathbf{x}\mathbf{x}}$ as

$$\mathbf{C}_{\mathbf{x}\mathbf{x}} = [c_{ij}]_{n \times n} = \sum_{t=0}^{n-1} \mathbf{x}(t) \times \mathbf{x}(t) = \sum_{t=0}^{n-1} \mathbf{x}^T(t) \mathbf{x}(t) \quad (5)$$

and for two code vectors \mathbf{x} and \mathbf{y} , we consider $\mathbf{C}_{\mathbf{x}\mathbf{y}}$ as

$$\mathbf{C}_{\mathbf{x}\mathbf{y}} = [c_{ij}]_{n \times n} = \sum_{t=0}^{n-1} \mathbf{x}(t) \times \mathbf{y}(t) = \sum_{t=0}^{n-1} \mathbf{x}^T(t) \mathbf{y}(t). \quad (6)$$

Lemma 3: The elements of $\mathbf{C}_{\mathbf{x}\mathbf{x}}$ located on τ th diagonal are equal to the auto-correlation function of vector \mathbf{x} for the τ th cyclic shift.

Proof: If c_{ij} is on the τ th diagonal of $\mathbf{C}_{\mathbf{x}\mathbf{x}}$ as we saw before, then $j = i \oplus \tau$. On the other hand, from (5), we have

$$c_{ij} = \sum_{t=0}^{n-1} x_{i \oplus t} x_{j \oplus t}$$

therefore

$$c_{ij} = \sum_{t=0}^{n-1} x_{i \oplus t} x_{i \oplus \tau \oplus t} = \sum_{k=0}^{n-1} x_k x_{k \oplus \tau}. \quad (7)$$

Lemma 4: The elements of $\mathbf{C}_{\mathbf{x}\mathbf{y}}$ located on the τ th diagonal are equal to the cross-correlation function of vectors \mathbf{x} and \mathbf{y} for the τ th cyclic shift.

Proof: Similar to Lemma 3, from (6), we can write

$$c_{ij} = \sum_{t=0}^{n-1} x_{i \oplus t} y_{j \oplus t} = \sum_{t=0}^{n-1} x_{i \oplus t} y_{i \oplus \tau \oplus t} = \sum_{k=1}^n x_k y_{k \oplus \tau}. \quad (8)$$

Example 4: Consider the two code vectors of Example 2. Then

$$\begin{aligned} \mathbf{C}_{\mathbf{x}\mathbf{x}} &= \begin{pmatrix} 2 & 0 & 1 & 1 & 0 \\ 0 & 2 & 0 & 1 & 1 \\ 1 & 0 & 2 & 0 & 1 \\ 1 & 1 & 0 & 2 & 0 \\ 0 & 1 & 1 & 0 & 2 \end{pmatrix} \\ \mathbf{C}_{\mathbf{y}\mathbf{y}} &= \begin{pmatrix} 2 & 1 & 0 & 0 & 1 \\ 1 & 2 & 1 & 0 & 0 \\ 0 & 1 & 2 & 1 & 0 \\ 0 & 0 & 1 & 2 & 1 \\ 1 & 0 & 0 & 1 & 2 \end{pmatrix} \\ \mathbf{C}_{\mathbf{x}\mathbf{y}} &= \begin{pmatrix} 1 & 1 & 0 & 1 & 1 \\ 1 & 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 & 0 \\ 0 & 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 \end{pmatrix}. \end{aligned}$$

Using Lemmas 3 and 4, we can find a way to verify which code vectors satisfy the correlation properties.

Lemma 5: Any code vector \mathbf{x} satisfies the auto-correlation constraint if and only if (iff) all the elements of $\mathbf{C}_{\mathbf{x}\mathbf{x}}$ not located on the main diagonal are less than or equal to λ_a .

Lemma 6: Any two code vectors \mathbf{x} and \mathbf{y} satisfy the cross-correlation constraint iff all the elements of $\mathbf{C}_{\mathbf{x}\mathbf{y}}$ are less than or equal to λ_c .

Now we can present the following theorem for the special case of $\lambda_c = 1$.

Theorem 1: In an $(n, w, \lambda_a, 1)$ OOC, with m code vectors \mathbf{x}_i ($i = 1, \dots, m$), if the τ th diagonal of $\mathbf{C}_{\mathbf{x}\mathbf{x}}^i$ has a nonzero value, then this diagonal in all other matrices $\mathbf{C}_{\mathbf{x}\mathbf{x}}^j$ ($i \neq j$) is zero.

Proof: According to Lemma 3, if the τ th diagonal of $\mathbf{C}_{\mathbf{x}\mathbf{x}}^i$ is nonzero, \mathbf{x}_i has at least two chip pulses with relative distance τ . Since $\lambda_c = 1$, no other code vector \mathbf{x}_j ($i \neq j$) can have two or more chip pulses with relative distance τ . So the τ th diagonal of all other matrices should be zero (see Example 4).

As the result of Theorem 1, we conclude with the two following lemmas.

Lemma 7: In a family of OOCs with $\lambda_c = 1$, if $i \neq j$, then $\mathbf{C}_{\mathbf{x}\mathbf{x}}^i \neq \mathbf{C}_{\mathbf{x}\mathbf{x}}^j$.

Lemma 8: In a family of OOCs with $\lambda_c = 1$, matrix $\mathbf{D}_{\mathbf{x}\mathbf{x}}^i$ has at most λ_a "1" elements in its τ th off-principal diagonal. On the other hand, if the τ th off-principal diagonal of matrix $\mathbf{D}_{\mathbf{x}\mathbf{x}}^i$ has some "1" elements, then this diagonal in all other matrices is all zero.

Now by relating a matrix $\mathbf{D}_{\mathbf{x}\mathbf{x}}^i$ to any codeword \mathbf{x}_i , we can introduce the new definition of OOCs with $\lambda_c = 1$ based on

tigated some of their properties and discussed a new definition of OOCs based on these properties. Furthermore, we presented a new search algorithm based on outer-product matrices, and we showed that it is more efficient when compared with the accelerated greedy algorithm in terms of number of obtained codewords, especially when the cardinality of the OOC is large.

APPENDIX

In this appendix, we give pseudocode that just clarifies all the steps of outer-product matrix-based search algorithm.

```
// md: total number of desired codewords
// wd: code weight
// n: code length
// Forbidden positions because of cross-correlation
  constraints are filled with  $md + 1$ 
// Forbidden positions because of auto-correlation
  constraints are filled with  $md + 2$ 
reset all variables with zero
for  $i = 0$  to  $n$ ,  $D(i, i) = md + 1$ 
if  $\text{mod}(n, 2) = 0$  then for  $i = 0$  to  $n - 1$ ,
 $D(\text{mod}(n/2 + i, n), i) = md + 1$ 
 $D1(s, :, :) = D$ 
while( $m! = md$ )
   $m = m + 1$ 
   $w = 1$ 
  replace  $md + 2$  with 0 in  $D$ 
  while( $w! = wd$ )
     $D1(s, :, :) = D$ 
     $s = s + 1$ 
    cnt=number of empty positions in  $D(0, :)$ 
    if cnt <  $wd - w$  then
      while(cnt <  $wd - w$ )
         $s = s - 1$ 
        if  $w > 1$  then  $w = w - 1$  else
          ( $w = wd - 1$  and  $m = m - 1$ )
           $D1(s - 1, 0, x(m, w)) = md + 2$ 
          cnt=number of empty positions in
           $D1(s - 1, 0, x(m, w))$ 
         $D = D1(s - 1, :, :)$ 
   $t$ =random number between 0 and  $n - 1$  where
   $D(0, t) = 0$ 
   $D(0, t) = m$ 
   $x(m, w) = t$ 
   $w = w + 1$ 
  fill all other places in  $t$ th diagonal with  $md + 1$ 
```

for $j = 1$ to $w - 2$

$$y = x(m, j)$$

$$D(y, t) = m \text{ and } D(t, y) = m$$

Fill all other places in $|t - y|$ th and $n - |t - y|$ th diagonal with $md + 1$

for $j = 0$ to $w - 2$

$$y = x(m, j)$$

if $\text{mod}(t - y, 2) = D(0, (t + y)/2) = 0$ then
 $D(0, \text{mod}((y + t)/2, n)) = md + 2$

if $\text{mod}(n + y - t, 2) = D(0, \text{mod}((n + y + t)/2, n)) = 0$ then $D(0, \text{mod}((n + y + t)/2, n)) = md + 2$

$$y = x(m, j + 1)$$

for $i = 1$ to $n - 1$

if $D(y, i) = m + 1$ and $D(0, i) = 0$
 then $D(0, i) = m + 2$

// At this time the numbers in x represent chip positions
 of md codewords

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