Orthogonal Discrete Fourier and Cosine Matrices for Signal Processing

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1. Introduction

The A DFT (Discrete Fourier Transform) has seen studied and applied to signal processing and communication theory. The relation between the Fourier matrix and the Hadamard transform was developed in [Ahmed & Rao, 1975; Whelchel & Guinn, 1968] for signal representation and classification and the Fast Fourier-Hadamard Transform (FFHT) was proposed. This idea was further investigated in [Lee & Lee, 1998] as an extension of the conventional Hadamard matrix. Lee et al [Lee & Lee, 1998] has proposed the Reverse Jacket Transform (RJT) based on the decomposition of the Hadamard matrix into the Hadamard matrix (unitary matrix) itself and a sparse matrix. Interestingly, the Reverse Jacket (RJ) matrix has a strong geometric structure that reveals a circulant expansion and contraction properties from a basic 2x2 sparse matrix.

The discrete Fourier transform (DFT) is an orthogonal matrix with highly practical value for representing signals and images [Ahmed & Rao, 1975; Lee, 1992; Lee, 2000]. Recently, the Jacket matrices which generalize the weighted Hadamard matrix were introduced in [Lee, 2000], [Lee & Kim, 1984, Lee, 1989, Lee & Yi, 2001; Fan & Yang, 1998]. The Jacket matrix is an abbreviated name of a reverse Jacket geometric structure. It includes the conventional Hadamard matrix [Lee, 1992; Lee, 2000; Lee et al., 2001; Hou et. al., 2003], but has the weights, ω, that are j or 2^k, where k is an integer, and j = √-1, located in the central part of Hadamard matrix. The weighted elements’ positions of the forward matrix can be replaced by the non-weighted elements of its inverse matrix and the signs of them do not change between the forward and inverse matrices, and they are only as element inverse and transpose. This reveals an interesting complementary matrix relation.

Definition 1: If a matrix \([J]_w\) of size \(m \times m\) has nonzero elements

\[
[J]_w = \begin{bmatrix}
  j_{0,0} & j_{0,1} & \cdots & j_{0,m-1} \\
  j_{1,0} & j_{1,1} & \cdots & j_{1,m-1} \\
  \vdots & \vdots & \ddots & \vdots \\
  j_{m-1,0} & j_{m-1,1} & \cdots & j_{m-1,m-1}
\end{bmatrix},
\]

(6-1)
Fourier Transforms - Approach to Scientific Principles

\[
\begin{bmatrix}
1 / j_{0,0} & 1 / j_{0,1} & \cdots & 1 / j_{0,m-1} \\
1 / j_{1,0} & 1 / j_{1,1} & \cdots & 1 / j_{1,m-1} \\
\vdots & \vdots & \ddots & \vdots \\
1 / j_{m-1,0} & 1 / j_{m-1,1} & \cdots & 1 / j_{m-1,m-1}
\end{bmatrix}
\]

(6-2)

where \( C \) is the normalizing constant, and \( T \) is of matrix transposition, then the matrix \([J]_m\) is called a Jacket matrix [Whelchel & Guinn, 1968],[Lee et al., 2001],[Lee et al, 2008; Chen et al., 2008]. Especially orthogonal matrices, such as Hadamard, DFT, DCT, Haar, and Slant matrices belong to the Jacket matrices family [Lee et al., 2001]. In addition, the Jacket matrices are associated with many kind of matrices, such as unitary matrices, and Hermitian matrices which are very important in communication (e.g., encoding), mathematics, and physics.

In section 2 DFT matrix is revisited in the sense of sparse matrix factorization. Section 3 presents recursive factorization algorithms of DFT and DCT matrix for fast computation. Section 4 proposes a hybrid architecture for implementation of algorithms simply by adding a switching device on a single chip module. Lastly, conclusions were drawn in section 4.

2. Preliminary of DFT presentation

The discrete Fourier transform (DFT) is a Fourier representation of a given sequence \( x(m) \), \( 0 \leq m \leq N - 1 \) and is defined as

\[
X(n) = \sum_{m=0}^{N-1} x(m)W^{nm}, \quad 0 \leq n \leq N - 1,
\]

(6-3)

where \( W = e^{-2\pi i / N} \). Let’s denote \( N \)-point DFT matrix as \( F_N = \left[ W^{nm} \right]_N \), \( n,m = \{0,1,2,...,N-1\} \), where \( W = e^{-2\pi i / N} \) (see about DFT in appendix), and the \( N \times N \) Sylvester Hadamard matrix as \( [H]_N \), respectively. The Sylvester Hadamard matrix is generated recursively by successive Kronecker products,

\[
[H]_N = [H]_{\frac{1}{2}} \otimes [H]_{\frac{1}{2}}
\]

(6-4)

for \( N=4 \), 8, 16, ... and \( [H]_{\frac{1}{2}} = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \). For the remainder of this chapter, analysis will be concerned only with \( N=2^k \), \( k=1,2,3, \ldots \) as the dimensionality of both the \( F \) and \( H \) matrices. 

**Definition 2**: A sparse matrix \( [S]_N \), which relates \( [F]_N \) and \( [H]_N \), can be computed from the factorization of \( F \) based on \( H \).

The structure of the \( S \) matrix is rather obscure. However, a much less complex and more appealing relationship will be identified for \( S \) [Park et al., 1999].

To illustrate the DFT using direct product we alter the denotation of \( W \) to lower case \( w = e^{-2\pi i} \), so that \( w^{n/N} \) becomes the n-th root of unit for \( N \)-point \( W \). For instance, the DFT matrix of dimension 2 is given by:

\[
\begin{bmatrix}
1 & 1 \\
1 & -1
\end{bmatrix}
\]
$$[F]_n = \begin{bmatrix} w^0 & w^0 & w^0 & w^0 \\ w^0 & w^1 & w^2 & w^3 \\ w^0 & w^2 & w^4 & w^6 \\ w^0 & w^3 & w^6 & w^9 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & e^{-j\frac{\pi}{2}} & e^{-j\frac{\pi}{2}} & e^{-j\frac{\pi}{2}} \\ 1 & e^{-j\frac{\pi}{2}} & e^{-j\frac{\pi}{4}} & e^{-j\frac{\pi}{4}} \\ 1 & e^{-j\frac{\pi}{2}} & e^{-j\frac{\pi}{4}} & e^{-j\frac{\pi}{4}} \end{bmatrix} \cdot [H]_n. \quad (6-5)$$

Let’s define

$$[W]_n = \begin{bmatrix} w^{0/4} & 0 & 0 & 0 \\ 0 & w^{1/4} & 0 & 0 \\ 0 & 0 & w^{2/4} & 0 \\ 0 & 0 & 0 & w^{3/4} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad \text{and} \quad [E]_n = [F]_n[W]_n = \begin{bmatrix} 1 & -j \\ 1 & j \end{bmatrix}, \quad (6-6)$$

and in general, We define

$$[W]_N = \text{diag}(w^{0/2N}, w^{1/2N}, w^{2/2N}, \ldots, w^{(N-1)/2N})$$

and

$$[E]_N = [F]_n[W]_N = [P]_N[F]_N[W]_N. \quad (6-7)$$

where $[P]_N$ is a permutation matrix and $[F]_N$ is a permuted version of DFT matrix $[F]_n$.

3. A sparse matrix factorization of orthogonal transforms

3.1 A sparse matrix analysis of discrete Fourier transform

Now we will present the Jacket matrix from a direct product of a sparse matrix computation and representation given by [Lee, 1989], [Lee & Finlayson, 2007]

$$[J]_m = \frac{1}{m} [H]_m [S]_m, \quad (6-8)$$

where $m = 2^k+1$, $k \in \{1,2,3,4,\ldots\}$ and $[S]_m$ is sparse matrix of $[J]_m$. Thus the inverse of the Jacket matrix can be simply written as

$$([J]_m)^{-1} = ([S]_m)^{-1} [H]_m. \quad (6-9)$$

As mentioned previously, the DFT matrix is also a Jacket matrix. By considering the sparse matrix for the 4-piont DFT matrix $[F]_n$,

$$[F]_4 = \begin{bmatrix} W^0 & W^0 & W^0 & W^0 \\ W^0 & W^1 & W^2 & W^3 \\ W^0 & W^2 & W^4 & W^6 \\ W^0 & W^3 & W^6 & W^9 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & e^{-j\frac{\pi}{2}} & e^{-j\frac{\pi}{2}} & e^{-j\frac{\pi}{2}} \\ 1 & e^{-j\frac{\pi}{2}} & e^{-j\frac{\pi}{4}} & e^{-j\frac{\pi}{4}} \\ 1 & e^{-j\frac{\pi}{2}} & e^{-j\frac{\pi}{4}} & e^{-j\frac{\pi}{4}} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & -j & j \\ 1 & -j & -1 & -j \end{bmatrix}.$$
where $E_2 = \begin{bmatrix} 1 & -j \\ 1 & j \end{bmatrix}$, its inverse matrix is from element-inverse, such that

$$(E_2)^{-1} = \begin{bmatrix} 1 & 1 \\ j & -j \end{bmatrix} = \begin{bmatrix} 1/1 & -1/j \\ 1/1 & 1/j \end{bmatrix}^T. \quad (6-11)$$

In general, we can write that

$$\begin{bmatrix} \tilde{F} \end{bmatrix}_N = [\Pr]_N \begin{bmatrix} F \end{bmatrix}_N = \begin{bmatrix} \tilde{F} \end{bmatrix}_{N/2} - \begin{bmatrix} \tilde{E} \end{bmatrix}_{N/2}, \end{bmatrix}_{N/2} = \begin{bmatrix} I_{N/2} & I_{N/2} \\ I_{N/2} & -I_{N/2} \end{bmatrix} \begin{bmatrix} \tilde{F} \end{bmatrix}_{N/2} \begin{bmatrix} 0 \\ E_{N/2} \end{bmatrix}, \quad (6-12)$$

where $[F]_N = [\tilde{F}]_N$. And the submatrix $E_N$ could be written from (6-7) by

$$[E]_N = [F]_N [W]_N = [\Pr]_N [\tilde{F}]_N [W]_N, \quad (6-13)$$

where $[W]_N = \begin{bmatrix} W^0 & 0 & \cdots & 0 \\ 0 & W^1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \cdots & 0 & W^{N-1} \end{bmatrix}$, and $W$ is the complex unit for $2N$ point DFT matrix.

For example, $[E]_2 = \begin{bmatrix} 1 & -j \\ 1 & j \end{bmatrix}$ can be calculated by using

$$[E]_2 = [\Pr]_2 [\tilde{F}]_2 [W]_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} W^0 & 0 \\ 0 & W^1 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \end{bmatrix} = \begin{bmatrix} 1 & -j \\ 1 & j \end{bmatrix}. \quad (6-14)$$

By using the results from (6-12) and (6-13), we have a new DFT matrix decomposition as

$$\begin{bmatrix} \tilde{F} \end{bmatrix}_N = [\Pr]_N \begin{bmatrix} F \end{bmatrix}_N = \begin{bmatrix} I_{N/2} & I_{N/2} \\ I_{N/2} & -I_{N/2} \end{bmatrix} \begin{bmatrix} \tilde{F} \end{bmatrix}_{N/2} \begin{bmatrix} 0 \\ E_{N/2} \end{bmatrix} \end{bmatrix}_{N/2} = \begin{bmatrix} I_{N/2} & I_{N/2} \\ 0 & E_{N/2} \end{bmatrix} \begin{bmatrix} I_{N/2} & I_{N/2} \\ 0 & -I_{N/2} \end{bmatrix} \quad (6-15)$$
Finally, based on the recursive form we have

\[
\begin{bmatrix} \tilde{F}_N \\ \end{bmatrix} = \begin{bmatrix} I_{N/2} & 0 \\ 0 & \text{Pr}_{N/2} \\ \end{bmatrix} \begin{bmatrix} I_{N/2} & 0 \\ 0 & \text{Pr}_{N/2} \\ \end{bmatrix} \begin{bmatrix} I_{N/2} & 0 \\ 0 & \text{W}_{N/2} \\ \end{bmatrix} \begin{bmatrix} I_{N/2} & 0 \\ 0 & -I_{N/2} \\ \end{bmatrix}
\]

(6-15)

Using (6-16) butterfly data flow diagram for DFT transform is drawn from left to right to perform \( X = [F]_N x \).

**Fig. 1.** Butterfly data flow diagram of proposed DFT matrix with order \( N \)

### 3.2 A Sparse matrix analysis of discrete cosine transform

Similar to the section 3.1, we will present the DCT matrices by using the element inverse or block inverse Jacket like sparse matrix [Lee, 2000; Park et al., 1999] decomposition. In this
section, the simple construction and fast computation for forward and inverse calculations and analysis of the sparse matrices, was very useful for developing the fast algorithms and orthogonal codes design.

Discrete Cosine Transform (DCT) is widely used in image processing, and orthogonal transform. There are four typical DCT matrices [Rao & Yip, 1990; Rao & Hwang, 1996],

\[
\begin{align*}
\text{DCT - I:} & \quad C_{(N+1)}^{n,m} = \sqrt{\frac{2}{N}} k_n k_m \cos \frac{m n \pi}{N}, \quad m, n = 0, 1, \ldots, N \; ; \\
\text{DCT - II:} & \quad C_{N}^{n,m} = \sqrt{\frac{2}{N}} k_m \cos \frac{m(n + \frac{1}{2}) \pi}{N}, \quad m, n = 0, 1, \ldots, N - 1 \; ; \\
\text{DCT - III:} & \quad C_{N}^{n,m} = \sqrt{\frac{2}{N}} k_m \cos \frac{(m + \frac{1}{2})m \pi}{N}, \quad m, n = 0, 1, \ldots, N - 1 \; ; \\
\text{DCT - IV:} & \quad C_{N}^{n,m} = \sqrt{\frac{2}{N}} \cos \frac{(m + \frac{1}{2})(n + \frac{1}{2}) \pi}{N}, \quad m, n = 0, 1, \ldots, N - 1 ,
\end{align*}
\]

where

\[
k_j = \begin{cases} 
1, & j = 1, 2, \ldots, N - 1 \\
\frac{1}{\sqrt{2}}, & j = 0, N 
\end{cases}
\]

To describe the computations of DCT, in this chapter, we will focus on the DCT - II algorithm, and introduce the sparse matrix decomposition and fast computations. The 2-by-2 DCT - II matrix can be simply written as

\[
\begin{bmatrix}
\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\
C_4^1 & C_4^3
\end{bmatrix}
\begin{bmatrix}
\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\
\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}}
\end{bmatrix}
= \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \frac{1}{\sqrt{2}},
\]

where \(1/\sqrt{2}\) can be seen as a special element inverse matrix of order 1, its inverse is \(\sqrt{2}\), and \(C_i = \cos(i\pi/4)\) is the cosine unit for DCT computations.

Furthermore, 4-by-4 DCT - II matrix is of the form

\[
\begin{bmatrix}
1 & 1 & 1 & 1 \\
\sqrt{2} & \sqrt{2} & \sqrt{2} & \sqrt{2} \\
C_8^1 & C_8^3 & C_8^7 & C_8^7 \\
C_8^2 & C_8^6 & C_8^6 & C_8^2 \\
C_8^4 & C_8^4 & C_8^2 & C_8^2 \\
C_8^8 & C_8^8 & C_8^8 & C_8^8
\end{bmatrix}.
\]
we can write

\[
[P][C] = \begin{bmatrix}
1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1
\end{bmatrix}
\begin{bmatrix}
\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\
C^1_8 & C^3_8 & C^5_8 & C^7_8 \\
C^2_8 & C^6_8 & C^2_8 & C^6_8 \\
C^3_8 & C^7_8 & C^1_8 & C^5_8
\end{bmatrix} = \begin{bmatrix}
\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\
C^1_8 & C^6_8 & C^6_8 & C^2_8 \\
C^2_8 & C^3_8 & C^5_8 & C^7_8 \\
C^3_8 & C^7_8 & C^1_8 & C^5_8
\end{bmatrix},
\]  

(6-23)

where \([P] = \begin{bmatrix} 1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1 \end{bmatrix}\) is permutation matrix. \([P]_N\) permutation matrix is a special case which has the form

\[
[P]_N = \begin{bmatrix}
1 & 0 & 0 & \ldots & 0 & 0 & \ldots & 0 \\
0 & 0 & 0 & \ldots & 1 & 0 & \ldots & 0 \\
0 & 1 & 0 & \ldots & 0 & 0 & \ldots & 0 \\
0 & 0 & 0 & \ldots & 0 & 0 & \ldots & 0 \\
0 & 0 & 1 & \ldots & 0 & 0 & \ldots & 0 \\
0 & 0 & 0 & \ldots & \ldots & \ldots & \ldots & \ldots \\
0 & 0 & \ldots & \ldots & 0 & 0 & \ldots & 0 \\
0 & 0 & \ldots & \ldots & \ldots & \ldots & \ldots & 1
\end{bmatrix}, \quad N \geq 4,
\]

where

\[
[P]_N = \begin{bmatrix}
pr_{i,j} \end{bmatrix}_N,
\]

with

\[
pr_{i,j} = \begin{cases}
1, & \text{if} \quad i = 2j, \quad 0 \leq j \leq \frac{N}{2} - 1, \\
1, & \text{if} \quad i = (2j + 1) \mod N, \quad \frac{N}{2} \leq j \leq N - 1, \\
0, & \text{others}
\end{cases}
\]

where \(i,j \in \{0, 1, \ldots, N - 1\}\).

Since \(C^1_8 = -C^7_8, C^2_8 = -C^6_8, C^3_8 = -C^5_8\), we rewrite (6-23) as

\[
[P]_4[C] = \begin{bmatrix}
\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\
C^1_8 & C^3_8 & C^5_8 & C^7_8 \\
C^2_8 & C^6_8 & C^2_8 & C^6_8 \\
C^3_8 & C^7_8 & C^1_8 & C^5_8
\end{bmatrix} = \begin{bmatrix}
\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\
C^1_8 & -C^2_8 & -C^2_8 & C^1_8 \\
C^2_8 & C^3_8 & -C^3_8 & -C^1_8 \\
C^3_8 & -C^4_8 & C^4_8 & -C^3_8
\end{bmatrix},
\]  

(6-24)
and let us define a column permutation matrix \( [P_c] = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{bmatrix} \), and \( [P_c]_N \) is a reversible permutation matrix which is defined by

\[
[P_c] = [I], \quad \text{and} \\
[P_c]_N = \begin{bmatrix} I_{N/4} & 0 & 0 & 0 \\ 0 & I_{N/4} & 0 & 0 \\ 0 & 0 & I_{N/4} & 0 \\ 0 & 0 & 0 & I_{N/4} \end{bmatrix}, \quad N \geq 4.
\]

Thus we have

\[
[P][C][P_c] = \begin{bmatrix} 1 & 1 & 1 & 1 \\ \sqrt{2} & \sqrt{2} & \sqrt{2} & \sqrt{2} \\ C_8 & C_8 & C_8 & C_8 \\ C_8^2 & C_8^2 & C_8^2 & C_8^2 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{bmatrix}, \quad \text{where} \quad \begin{bmatrix} C_8 & C_8 \\ C_8^2 & C_8^2 \\ C_8^3 & C_8^3 \\ C_8^4 & C_8^4 \end{bmatrix}
\]

\[
= \begin{bmatrix} 1 & 1 & 1 & 1 \\ \sqrt{2} & \sqrt{2} & \sqrt{2} & \sqrt{2} \\ C_8 & -C_8 & C_8 & -C_8 \\ C_8^2 & C_8^2 & C_8^2 & C_8^2 \end{bmatrix} = \begin{bmatrix} C_2 & C_2 \\ B_2 & -B_2 \end{bmatrix}, \quad (6-25)
\]

where \( [C] = \begin{bmatrix} 1 & 1 \\ \sqrt{2} & \sqrt{2} \\ 1 & 1 \\ \sqrt{2} & \sqrt{2} \end{bmatrix} \), the 2-by-2 DCT - II matrix and \( [B] = \begin{bmatrix} C_8 & C_8 \\ C_8^3 & C_8^3 \end{bmatrix} \). Thus we can write that

\[
[P][C][P_c] = \begin{bmatrix} C_2 & 0 \\ B_2 & -B_2 \end{bmatrix} = \begin{bmatrix} I_2 & I_2 \\ I_2 & -I_2 \end{bmatrix} \begin{bmatrix} C_2 & 0 \\ 0 & B_2 \end{bmatrix}^T, \quad (6-26)
\]

it is clear that \( \begin{bmatrix} C_2 & 0 \\ 0 & B_2 \end{bmatrix} \) is a block inverse matrix, which has

\[
\begin{bmatrix} C_2 & 0 \\ 0 & B_2 \end{bmatrix}^{-1} = \begin{bmatrix} (C_2)^{-1} & 0 \\ 0 & (B_2)^{-1} \end{bmatrix}. \quad (6-27)
\]
The (6-27) is jacket-like sparse matrix with block inverse.

In general the permuted DCT-II matrix $\tilde{C}_N$ can be constructed recursively by using
\[
\begin{align*}
\tilde{C}_N &= \left[ P \right]_N \left[ C \right]_N \left[ P c \right]_N = \\
&= \begin{bmatrix}
C_{N/2} & C_{N/2} \\
B_{N/2} & -B_{N/2}
\end{bmatrix}
\begin{bmatrix}
I_{N/2} & I_{N/2} \\
I_{N/2} & -I_{N/2}
\end{bmatrix}
\begin{bmatrix}
C_{N/2} & 0 \\
0 & B_{N/2}
\end{bmatrix}^T.
\end{align*}
\]
where $\left[ C \right]_{N/2}$ denotes the $\frac{N}{2} \times \frac{N}{2}$ DCT-II matrix, and $\left[ B \right]_{N/2}$ can be calculated by using
\[
\left[ B \right]_{N/2} = \left[ (C_{2N}^{(m,n)})_{m,n} \right]_{N/2},
\]
where
\[
\begin{align*}
f(m,1) &= 2m - 1, \\
f(m,n+1) &= f(m,n) + f(m,1) \times 2,
\end{align*}
\]
\[m,n \in \{1,2,...,N/2\}.\]

For example, in the 4-by-4 permuted DCT-II matrix $\tilde{C}_4$, $B_2$ could be calculated by using $f(1,1) = 1$, $f(2,1) = 3$, $f(1,2) = f(1,1) + f(1,1) \times 2 = 3$, and $f(2,2) = f(2,1) + f(2,1) \times 2 = 9$,
\[
\left[ B \right]_2 = \left[ (C_{8}^{(m,n)})_{m,n} \right]_4 = \\
\begin{bmatrix}
(C_{6}^{(1,1)})_{1,1} & (C_{6}^{(1,2)})_{1,1} \\
(C_{6}^{(2,1)})_{2,1} & (C_{6}^{(2,2)})_{2,2}
\end{bmatrix}.
\]

and its inverse is of $\tilde{C}_N$ can be simply computed from the block inverse
\[
\left( \left[ P \right]_N \left[ C \right]_N \left[ P c \right]_N \right)^{-1} = \\
\begin{bmatrix}
I_{N/2} & I_{N/2} \\
I_{N/2} & -I_{N/2}
\end{bmatrix}
\begin{bmatrix}
C_{N/2} & 0 \\
0 & B_{N/2}
\end{bmatrix}^{-1} = \\
\begin{bmatrix}
(C_{N/2})^{-1} & 0 \\
0 & (B_{N/2})^{-1}
\end{bmatrix}
\begin{bmatrix}
\frac{2}{N}I_{N/2} & \frac{2}{N}I_{N/2} \\
\frac{2}{N}I_{N/2} & -\frac{2}{N}I_{N/2}
\end{bmatrix}^T.
\]

For example, the $8 \times 8$ DCT-II matrix has
Additionally, it is clearly that the function (6-28) also can be recursively constructed by
using different permutations matrices $[\tilde{P}]_N$ and $[\tilde{P}c]_N$, as

$$
[\tilde{P}]_N [\tilde{C}]_N [\tilde{P}c]_N = \begin{bmatrix}
\sqrt{2} & \sqrt{2} & \sqrt{2} & \sqrt{2} & \sqrt{2} & \sqrt{2} \\
C_{16} & C_{3} & C_{5} & C_{7} & C_{9} & C_{11} \\
C_{16} & C_{16} & C_{16} & C_{16} & C_{16} & C_{16} \\
C_{16} & C_{16} & C_{16} & C_{16} & C_{16} & C_{16} \\
C_{16} & C_{16} & C_{16} & C_{16} & C_{16} & C_{16} \\
C_{16} & C_{16} & C_{16} & C_{16} & C_{16} & C_{16} \\
\end{bmatrix}
$$

and it can be represented by

$$
\begin{bmatrix}
\sqrt{2} & \sqrt{2} & \sqrt{2} & \sqrt{2} & \sqrt{2} & \sqrt{2} \\
C_{16} & C_{16} & C_{16} & C_{16} & C_{16} & C_{16} \\
C_{16} & C_{16} & C_{16} & C_{16} & C_{16} & C_{16} \\
C_{16} & C_{16} & C_{16} & C_{16} & C_{16} & C_{16} \\
C_{16} & C_{16} & C_{16} & C_{16} & C_{16} & C_{16} \\
C_{16} & C_{16} & C_{16} & C_{16} & C_{16} & C_{16} \\
\end{bmatrix}
$$

Additionally, it is clearly that the function (6-28) also can be recursively constructed by
using different permutations matrices $[\tilde{P}]_N$ and $[\tilde{P}c]_N$, as

$$
[\tilde{P}]_N [\tilde{C}]_N [\tilde{P}c]_N = \begin{bmatrix}
\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\
C_{16}^4 & -C_{16}^4 & C_{16}^4 & -C_{16}^4 & C_{16}^4 & -C_{16}^4 \\
C_{16}^6 & -C_{16}^6 & C_{16}^6 & -C_{16}^6 & C_{16}^6 & -C_{16}^6 \\
C_{16}^8 & -C_{16}^8 & C_{16}^8 & -C_{16}^8 & C_{16}^8 & -C_{16}^8 \\
C_{16}^{10} & -C_{16}^{10} & C_{16}^{10} & -C_{16}^{10} & C_{16}^{10} & -C_{16}^{10} \\
C_{16}^{12} & -C_{16}^{12} & C_{16}^{12} & -C_{16}^{12} & C_{16}^{12} & -C_{16}^{12} \\
\end{bmatrix}
$$
\[
\begin{bmatrix}
\tilde{P}_{N}\ 
\tilde{C}_{N}
\end{bmatrix}_N =
\begin{bmatrix}
P_{N/2}C_{N/2}PC_{N/2} & P_{N/2}C_{N/2}PC_{N/2} \\
P_{N/2}C_{N/2}PC_{N/2} & -I_{N/2}B_{N/2}BP_{N/2}
\end{bmatrix}
\begin{bmatrix}
\tilde{C}_{N/2}
\tilde{C}_{N/2}
\end{bmatrix}_N,
\]
(6-35)

where \( \tilde{B}_{N/2} = [B]_{N/2}[Pc]_{N/2} \).

For example, the 4-by-4 DCT - II case has
\[
\begin{bmatrix}
\tilde{B}_{4/2}
\end{bmatrix}_{4/2} = \begin{bmatrix}
C_{4/2}^1 & C_{4/2}^3 \\
C_{4/2}^3 & -C_{4/2}^1
\end{bmatrix}
\begin{bmatrix}
1 & 0 \\
0 & 1
\end{bmatrix} = \begin{bmatrix}
C_{4/2}^1 & C_{4/2}^3 \\
C_{4/2}^3 & -C_{4/2}^1
\end{bmatrix}.
\]
(6-36)

Moreover, the matrix \( [B]_2 = \begin{bmatrix}
C_8^1 & C_8^3 \\
C_8^3 & -C_8^1
\end{bmatrix} \) can be decomposed by using the 2-by-2 DCT - II matrix as
\[
\begin{bmatrix}
B_2
\end{bmatrix}_2 = \begin{bmatrix}
\sqrt{2} & 0 \\
-\sqrt{2} & 2
\end{bmatrix} \begin{bmatrix}
\sqrt{2} & 0 \\
\sqrt{2} & 2
\end{bmatrix}
\begin{bmatrix}
C_8^1 & 0 \\
0 & C_8^3
\end{bmatrix},
\]
(6-37)

where \( \begin{bmatrix}
\sqrt{2} & 0 \\
-\sqrt{2} & 2
\end{bmatrix} \) is a upper triangular matrix, \( \begin{bmatrix}
\sqrt{2} & 0 \\
\sqrt{2} & 2
\end{bmatrix} \) is a diagonal matrix, and we use the cosine related function
\[
\cos(2k + 1)\phi_m = 2\cos(2k\phi_m)\cos\phi_m - \cos(2k - 1)\phi_m ,
\]
(6-38)

where \( \phi_m \) is \( m \)-th angle.

In a general case, we have
\[
\begin{bmatrix}
B_2
\end{bmatrix}_N = \begin{bmatrix}
K_2
\end{bmatrix}_N \begin{bmatrix}
C_2
\end{bmatrix}_N \begin{bmatrix}
D_2
\end{bmatrix}_N,
\]
(6-39)

where
\[
\begin{bmatrix}
K_2
\end{bmatrix}_N = \begin{bmatrix}
\sqrt{2} & 0 & 0 & \cdots \\
-\sqrt{2} & 2 & 0 & \cdots \\
\sqrt{2} & -2 & 2 & \cdots \\
\vdots & \vdots & \vdots & \ddots
\end{bmatrix},
\begin{bmatrix}
D_2
\end{bmatrix}_N = \begin{bmatrix}
\Phi_{4N} & 0 & \cdots & 0 \\
0 & \Phi_{4N} & \cdots & \vdots \\
0 & \cdots & \cdots & 0 \\
0 & \cdots & 0 & \Phi_{4N-1}
\end{bmatrix},
\]
(6-40)

and \( \Phi_i = 2i + 1, \ i \in \{0, 1, 2, \ldots, N - 1\} \). See appendix 2 for proof of (6-39).

By using the results from (6-28) and (6-39), we have a new form for DCT - II matrix
\[
\begin{bmatrix}
\tilde{C}
\end{bmatrix}_N = \begin{bmatrix}
[P\tilde{C}]
\end{bmatrix}_N \begin{bmatrix}
C
\end{bmatrix}_N \begin{bmatrix}
Pc
\end{bmatrix}_N = \begin{bmatrix}
I_{N/2} & I_{N/2} \\
I_{N/2} & -I_{N/2}
\end{bmatrix} \begin{bmatrix}
C_{N/2} & 0 \\
0 & B_{N/2}
\end{bmatrix}^T = \begin{bmatrix}
I_{N/2} & 0 \\
0 & B_{N/2}
\end{bmatrix} \begin{bmatrix}
I_{N/2} & I_{N/2} \\
I_{N/2} & -I_{N/2}
\end{bmatrix}
\]

By using the results from (6-28) and (6-39), we have a new form for DCT - II matrix
\[
\begin{bmatrix}
\tilde{C}
\end{bmatrix}_N = \begin{bmatrix}
[P\tilde{C}]
\end{bmatrix}_N \begin{bmatrix}
C
\end{bmatrix}_N \begin{bmatrix}
Pc
\end{bmatrix}_N = \begin{bmatrix}
I_{N/2} & I_{N/2} \\
I_{N/2} & -I_{N/2}
\end{bmatrix} \begin{bmatrix}
C_{N/2} & 0 \\
0 & B_{N/2}
\end{bmatrix}^T = \begin{bmatrix}
I_{N/2} & 0 \\
0 & B_{N/2}
\end{bmatrix} \begin{bmatrix}
I_{N/2} & I_{N/2} \\
I_{N/2} & -I_{N/2}
\end{bmatrix}
\]

By using the results from (6-28) and (6-39), we have a new form for DCT - II matrix
\[
\begin{bmatrix}
\tilde{C}
\end{bmatrix}_N = \begin{bmatrix}
[P\tilde{C}]
\end{bmatrix}_N \begin{bmatrix}
C
\end{bmatrix}_N \begin{bmatrix}
Pc
\end{bmatrix}_N = \begin{bmatrix}
I_{N/2} & I_{N/2} \\
I_{N/2} & -I_{N/2}
\end{bmatrix} \begin{bmatrix}
C_{N/2} & 0 \\
0 & B_{N/2}
\end{bmatrix}^T = \begin{bmatrix}
I_{N/2} & 0 \\
0 & B_{N/2}
\end{bmatrix} \begin{bmatrix}
I_{N/2} & I_{N/2} \\
I_{N/2} & -I_{N/2}
\end{bmatrix}
\]

By using the results from (6-28) and (6-39), we have a new form for DCT - II matrix
\[
\begin{bmatrix}
\tilde{C}
\end{bmatrix}_N = \begin{bmatrix}
[P\tilde{C}]
\end{bmatrix}_N \begin{bmatrix}
C
\end{bmatrix}_N \begin{bmatrix}
Pc
\end{bmatrix}_N = \begin{bmatrix}
I_{N/2} & I_{N/2} \\
I_{N/2} & -I_{N/2}
\end{bmatrix} \begin{bmatrix}
C_{N/2} & 0 \\
0 & B_{N/2}
\end{bmatrix}^T = \begin{bmatrix}
I_{N/2} & 0 \\
0 & B_{N/2}
\end{bmatrix} \begin{bmatrix}
I_{N/2} & I_{N/2} \\
I_{N/2} & -I_{N/2}
\end{bmatrix}
\]
Given the recursive form of (6-41), we can write

\[
\begin{pmatrix}
C_{N/2} & 0 \\
0 & K_{N/2} C_{N/2} D_{N/2}
\end{pmatrix} \begin{pmatrix}
I_{N/2} & I_{N/2} \\
I_{N/2} & -I_{N/2}
\end{pmatrix}
\]

This can be seen as

\[
\begin{pmatrix}
I_{N/2} & 0 \\
0 & K_{N/2}
\end{pmatrix} \begin{pmatrix}
C_{N/2} & 0 \\
0 & I_{N/2}
\end{pmatrix} \begin{pmatrix}
I_{N/2} & 0 \\
0 & D_{N/2}
\end{pmatrix} \begin{pmatrix}
I_{N/2} & I_{N/2} \\
I_{N/2} & -I_{N/2}
\end{pmatrix}
\]

By taking all permutation matrices outside, we can rewrite (6-42) as

\[
\begin{pmatrix}
I_{N/4} & 0 \\
0 & I_{N/4}
\end{pmatrix} \begin{pmatrix}
I_{2} & 0 \\
0 & K_{2}
\end{pmatrix} \begin{pmatrix}
I_{N/2} & I_{N/2} \\
I_{N/2} & -I_{N/2}
\end{pmatrix} (\mathcal{P}_{\mathcal{C}})_{N}^{-1}
\]

Using (6-43) butterfly data flow diagram for DCT-II transform is drawn as Fig. 2 from left to right to perform \(X = [C]_{N}X\).
3.3 Hybrid DCT/DFT architecture on element inverse matrices

It is clear that the form of (6-43) is the same as that of (6-16), where we only need change $K_l$ to $Pr_l$ and $D_l$ to $W_l$, with $l \in \{2,4,8,...,N/2\}$. Consequently, the results show that the DCT-II and DFT can be unified by using same algorithm and architecture within some characters changed. As illustrated in Fig.1, and Fig.2, we find that the DFT calculations can be obtained from the architecture of DCT by replacing the $[D]_N$ to $[W]_N$, and a permutation matrix $[Pr]_N$ to $[K]_N$. Hence a unified function block diagram for DCT/DFT hybrid architecture algorithm can be drawn as Fig.3. In this figure, we can joint DCT and DFT in one chip or one processing architecture, and use one switching box to control the output data flow. It will be useful to developing the unified chip or generalized form for DCT and DFT together.

![Hybrid Architecture for DCT/DFT Processor](image)

**Fig. 3.** A unified function block diagram for proposed DCT/DFT hybrid architecture algorithm

4. Conclusion

We propose a new representation of DCT/DFT matrices via the Jacket transform based on the block-inverse processing. Following on the factorization method of the Jacket transform, we show that the inverse cases of DCT/DFT matrices are related to their block inverse sparse matrices and the permutations. Generally, DCT/DFT can be represented by using the same architecture based on element inverse or block inverse decomposition. Linking between two transforms was derived based on matrix recursion formula. Discrete Cosine Transform (DCT) has applications in signal classification and representation, image coding, and synthesis of video signals. The DCT-II is a popular structure and it is usually accepted as the best suboptimal transformation that its
performance is very close to that of the statistically optimal Karhunen-Loeve transform for picture coding. Further, the discrete Fourier transform (DFT) is also a popular algorithm for signal processing and communications, such as OFDM transmission and orthogonal code designs. Being combined these two different transforms, a unified fast processing module to implement DCT/DFT hybrid architecture algorithm can be designed by adding switching device to control either DCT or DFT processing depending on mode of operation. Further investigation is needed for unified treatment of recursive decomposition of orthogonal transform matrices exploiting the properties of Jacket-like sparse matrix architecture for fast trigonometric transform computation.

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6. Appendix

6.1 Appendix 1

The DFT matrix brings higher powers of \( w \), and the problem turns out to be

\[
\begin{bmatrix}
1 & 1 & 1 & \cdots & 1 \\
1 & w & w^2 & \cdots & w^{n-1} \\
1 & w^2 & w^4 & \cdots & w^{2(n-1)} \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
1 & w^{n-1} & w^{2(n-1)} & \cdots & w^{n(n-1)}
\end{bmatrix}
\begin{bmatrix}
c_0 \\
c_1 \\
c_2 \\
\vdots \\
c_{n-1}
\end{bmatrix}
= \begin{bmatrix}
y_0 \\
y_1 \\
y_2 \\
\vdots \\
y_{n-1}
\end{bmatrix}. \tag{A-1}
\]

and the inverse form

\[
F^{-1} = \frac{1}{n} \begin{bmatrix}
1 & 1 & 1 & \cdots & 1 \\
1 & w & w^2 & \cdots & w^{n-1} \\
1 & w^2 & w^4 & \cdots & w^{2(n-1)} \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
1 & w^{n-1} & w^{2(n-1)} & \cdots & w^{n(n-1)}
\end{bmatrix}. \tag{A-2}
\]

Then, we can define the Fourier matrix as follows.

**Definition A.1:** An \( n \times n \) matrix \( F = \begin{bmatrix} a_{ij} \end{bmatrix} \) is a Fourier matrix if

\[
a_{ij} = w^{(i-1)(j-1)}, \quad w = e^{2\pi i/n}, \quad \text{and} \ i, j \in \{1, 2, \ldots, n\}. \tag{A-3}
\]

and the inverse form \( F^{-1} = \frac{1}{n} \left( \begin{bmatrix} a_{ij} \end{bmatrix} \right)^{-1} = \frac{1}{n} \left( \begin{bmatrix} w^{(i-1)(j-1)} \end{bmatrix} \right) \).

For example, in the cases \( n = 2 \) and \( n = 3 \), and inverse is an element-wise inverse like Jacket matrix, then, we have

\[
F_2 = \begin{bmatrix}
w^0 & w^0 \\
w^0 & w^1
\end{bmatrix}, \quad F_2^{-1} = \frac{1}{2} \begin{bmatrix}
w^0 & w^0 \\
w^0 & w^1
\end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 & 1 \\
1 & -1
\end{bmatrix}
\]

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and

\[
F_3 = \begin{bmatrix}
1 & 1 & 1 \\
1 & e^{i2\pi/3} & e^{i4\pi/3} \\
1 & e^{i4\pi/3} & e^{i8\pi/3}
\end{bmatrix}, \quad F_3^{-1} = \frac{1}{3} \begin{bmatrix}
1 & 1 & 1 \\
1 & e^{-i2\pi/3} & e^{-i4\pi/3} \\
1 & e^{-i4\pi/3} & e^{-i8\pi/3}
\end{bmatrix}.
\]

We need to confirm that \( F^{-1} \) equals the identity matrix. On the main diagonal that is clear. Row \( j \) of \( F \) times column \( j \) of \( F^{-1} \) is \( (1/\text{N})(1+1+...+1) \), which is 1. The harder part is off the diagonal, to show that row \( j \) of \( F \) times column \( k \) of \( F^{-1} \) gives zero:

\[
\text{if } j \neq k.
\]

The key is to notice that those terms are the powers of :  

\[
1 + W + W^2 + ... + W^{\text{N}-1} = 0.
\]

### 6.2 Appendix 2

In a general case, we have

\[
[B]_N = [K]_N [C]_N [D]_N,
\]

where

\[
[K]_N = \begin{bmatrix}
\sqrt{2} & 0 & 0 & \cdots \\
-\sqrt{2} & 2 & 0 & \cdots \\
\sqrt{2} & -2 & 2 & \cdots \\
\vdots & \vdots & \ddots & \ddots
\end{bmatrix}, \quad [D]_N = \begin{bmatrix}
C_{4N}^{\phi_0} & 0 & \cdots & 0 \\
0 & C_{4N}^{\phi_1} & \ddots & \vdots \\
\vdots & \vdots & \ddots & 0 \\
0 & \cdots & 0 & C_{4N}^{\phi_{N-1}}
\end{bmatrix},
\]

and \( \Phi_i = 2i+1, \ i \in \{0,1,2,...,N-1\} \).

**Proof:** In case of \( N \times N \) DCT - II matrix, \([C]_N\), it can be represented by using the form as

\[
[C]_N = \frac{1}{\sqrt{2}} \begin{bmatrix}
C_{4N}^{2k_0\phi_0} & C_{4N}^{2k_0\phi_1} & \cdots & C_{4N}^{2k_0\phi_{N-2}} & C_{4N}^{2k_0\phi_{N-1}} \\
C_{4N}^{2k_1\phi_0} & C_{4N}^{2k_1\phi_1} & \cdots & C_{4N}^{2k_1\phi_{N-2}} & C_{4N}^{2k_1\phi_{N-1}} \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
C_{4N}^{2k_{N-2}\phi_0} & C_{4N}^{2k_{N-2}\phi_1} & \cdots & C_{4N}^{2k_{N-2}\phi_{N-2}} & C_{4N}^{2k_{N-2}\phi_{N-1}}
\end{bmatrix},
\]

where \( k_i = i + 1, \ i \in \{0,1,2,...\} \).

According to (6-37), a \( N \times N \) matrix \([B]_N\) from \([C]_N\) can be simply presented by
And based on (6-78), we have the formula

\[ C_{4N}^{(2k+1)\Phi_w} = 2C_{4N}^{2k\Phi_w} C_{4N}^{\Phi_w} - C_{4N}^{(2k+1)\Phi_w} = -C_{4N}^{(2k+1)\Phi_w} + 2C_{4N}^{2k\Phi_w} C_{4N}^{\Phi_w}, \]

where \( m \in \{0, 1, 2, \ldots\} \).

Thus we can calculate that

\[ \begin{bmatrix} K_{4N} \end{bmatrix} [C_{4N}] [D_{4N}] = \begin{bmatrix} \sqrt{2} & 0 & 0 & 0 & \cdots & 0 \\ -\sqrt{2} & 2 & 0 & 0 & 0 \\ \sqrt{2} & -2 & 2 & 0 & 0 \\ -\sqrt{2} & 2 & -2 & 2 & \cdots \\ \sqrt{2} & -2 & 2 & -2 & 2 \\ \vdots & \vdots & \vdots & \vdots & \ddots \\ C_{4N}^{\Phi_0} & 0 & \cdots & 0 \\ 0 & C_{4N}^{\Phi_0} & \cdots & \vdots \\ \vdots & \vdots & \ddots & 0 \\ 0 & \cdots & 0 & C_{4N}^{\Phi_{N-1}} \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & \cdots & 1 & 1 \\ 1 & 1 & 1 & \cdots & 1 & 1 \\ 1 & 1 & 1 & \cdots & 1 & 1 \\ 1 & 1 & 1 & \cdots & 1 & 1 \\ 1 & 1 & 1 & \cdots & 1 & 1 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 1 & 1 & 1 & \cdots & 1 & 1 \end{bmatrix} \]

\[ \begin{bmatrix} C_{4N}^{\Phi_0} & 0 & \cdots & 0 \\ 0 & C_{4N}^{\Phi_0} & \cdots & \vdots \\ \vdots & \vdots & \ddots & 0 \\ 0 & \cdots & 0 & C_{4N}^{\Phi_{N-1}} \end{bmatrix} = \begin{bmatrix} 1 & -1 + 2C_{4N}^{2k\Phi_0} & \cdots & 1 \\ -1 + 2C_{4N}^{2k\Phi_0} & 1 - 2C_{4N}^{2k\Phi_0} + 2C_{4N}^{2k\Phi_1} & \cdots & -1 + 2C_{4N}^{2k\Phi_{N-1}} \\ \vdots & \vdots & \ddots & \vdots \\ C_{4N}^{\Phi_0} & 0 & \cdots & 0 \\ 0 & C_{4N}^{\Phi_0} & \cdots & \vdots \\ \vdots & \vdots & \ddots & 0 \\ 0 & \cdots & 0 & C_{4N}^{\Phi_{N-1}} \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & \cdots & 1 & 1 \\ 1 & 1 & 1 & \cdots & 1 & 1 \\ 1 & 1 & 1 & \cdots & 1 & 1 \\ 1 & 1 & 1 & \cdots & 1 & 1 \\ 1 & 1 & 1 & \cdots & 1 & 1 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 1 & 1 & 1 & \cdots & 1 & 1 \end{bmatrix} \]

\[ \begin{bmatrix} C_{4N}^{\Phi_0} & 0 & \cdots & 0 \\ 0 & C_{4N}^{\Phi_0} & \cdots & \vdots \\ \vdots & \vdots & \ddots & 0 \\ 0 & \cdots & 0 & C_{4N}^{\Phi_{N-1}} \end{bmatrix} = \begin{bmatrix} C_{4N}^{\Phi_0} - 2C_{4N}^{2k\Phi_0} C_{4N}^{\Phi_0} & \cdots & C_{4N}^{\Phi_0} - 2C_{4N}^{2k\Phi_0} C_{4N}^{\Phi_0} \\ \vdots & \ddots & \vdots \\ C_{4N}^{\Phi_0} - 2C_{4N}^{2k\Phi_0} C_{4N}^{\Phi_0} & \cdots & C_{4N}^{\Phi_0} - 2C_{4N}^{2k\Phi_0} C_{4N}^{\Phi_0} \end{bmatrix}, \]
Since \( k_0 = 1 \), we get
\[
-C_{4N}^{\phi_0} + 2C_{4N}^{2k_0\phi_0}C_{4N}^{\phi_0} = -C_{4N}^{2(2k_0-1)\phi_0} + 2C_{4N}^{2k_0\phi_0}C_{4N}^{\phi_0} = C_{4N}^{(2k_0+1)\phi_0}, \quad (A-10)
\]
and
\[
C_{4N}^{\phi_0} - 2C_{4N}^{2k_0\phi_0}C_{4N}^{\phi_0} = -(-C_{4N}^{2k_0\phi_0} + 2C_{4N}^{2k_0\phi_0}C_{4N}^{\phi_0}) = -C_{4N}^{2k_0+1\phi_0}. \quad (A-11)
\]
In case of \( k_i = i+1 \), we have
\[
(C_{4N}^{\phi_i})_{m+1} - (2k_i - 1)\phi_m = (2(k_i - 1) + 1)\phi_m = (2k_i - 1)\phi_m, \text{ then we get}
\]
\[
C_{4N}^{\phi_i} - 2C_{4N}^{2k_i\phi_i}C_{4N}^{\phi_i} = -C_{4N}^{(2k_i-1)\phi_i} + 2C_{4N}^{2k_i\phi_i}C_{4N}^{\phi_i} = -C_{4N}^{(2k_i+1)\phi_i} + 2C_{4N}^{2k_i\phi_i}C_{4N}^{\phi_i} \quad (A-12)
\]
Taking the (A-10)-(A-12) to (A-9), we can rewrite that
\[
[K]_N [C]_N [D]_N = \begin{bmatrix}
-C_{4N}^{\phi_0} + 2C_{4N}^{2k_0\phi_0}C_{4N}^{\phi_0} & C_{4N}^{\phi_0} & \cdots & C_{4N}^{\phi_{N-1}} \\
-C_{4N}^{\phi_0} + 2C_{4N}^{2k_0\phi_0}C_{4N}^{\phi_0} & C_{4N}^{\phi_0} & \cdots & C_{4N}^{\phi_{N-1}} \\
\vdots & \vdots & \ddots & \vdots \\
-C_{4N}^{\phi_0} + 2C_{4N}^{2k_0\phi_0}C_{4N}^{\phi_0} & C_{4N}^{\phi_0} & \cdots & C_{4N}^{\phi_{N-1}} \\
C_{4N}^{(2k_0+1)\phi_0} & C_{4N}^{(2k_0+1)\phi_0} & \cdots & C_{4N}^{(2k_0+1)\phi_0}
\end{bmatrix} = [B]_N. \quad (A-13)
\]
The proof is completed.

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