RELATIONSHIP BETWEEN DCT-II, DCT-VI, AND DST-VII TRANSFORMS

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ABSTRACT
Discrete Sine Transforms of type VII (DST-VII) have recently received considerable interest in video coding. In this paper, we show that there exists a direct connection between DST-VII and DCT-II transforms, allowing their joint computation for certain transform sizes. This connection also yields fast algorithms for constructing DCT-VI and DCT-VII.

Index Terms— KLT, DCT-II, DCT-VI, DST-VII, factorizations, video coding.

1. INTRODUCTION
The Discrete Cosine Transforms of types II and IV are among most fundamental, well understood, and much appreciated tools in data compression. The DCT-II is used at the core of standards for image and video compression, such as JPEG, ITU-T H.26x-series, and MPEG 1-4 standards [1]. The DCT-IV is used in audio coding algorithms, such as ITU-T Rec. G.722.1, MPEG-4 AAC, and others [2]. Such transforms are very well studied, and a number of efficient techniques exist for their computation [1, 3, 4, 5, 6, 7].

Much less known are so-called "odd" sinusoidal transforms: Discrete Cosine and Sine Transforms of types V, VI, VII, and VIII. Existence of some of such transforms was discovered by A. Jain in 1979 [8]. A complete tabulation was developed in 1985 by Wang and Hunt [9]. However, not much work has followed. Surveys of related results can be found in [10, 3].

Recently, DST of types VI and VII have surfaced as useful tools in image and video coding. In 2010, Han, Saxena, and Rose have shown that DST-VII produce good approximations of Karhunen-Loeve Transform (KLT) for model of residual signals after Intra-prediction [11]. This was subsequently validated in the course of experimental work on ISO/IEC/ITU-T High Efficiency Video Coding (HEVC) standard [12, 13].

The adoption of DST-VII in HEVC has prompted a discussion on the existence of fast algorithms for computing such transforms [12, 14]. This question was addressed in 2011 by Chivukula and Reznik[15], who have established a connection between DST-VII and DCT.

This paper offers an alternative solution by establishing a mapping between DST-VII, DCT-VI, and DCT-II. This mapping yields fast algorithms not only for DST-VI-VII, but also for DCT-VI-VII, as well as possible joint factorizations of such transforms. The obtained mapping may also be of interest from methodological standpoint, as it suggests additional connections between DST-VII, DCT-VI and KLT of residual and mixed signals.

The rest of this paper is organized as follows. Section 2 provides definitions. Section 3 establishes mapping between DCT-II, DCT-VI and DST-VII transforms. Section 4 explains how this mapping can be used to construct fast algorithms. Discussion and concluding remarks are offered in Section 5.

2. DEFINITIONS
Let $N$ be the length of data sequence. The matrices of Discrete Fourier Transform (DFT) and Discrete Cosine and Sine transforms of types II, III, IV, VI, and VII will be defined as follows:

\begin{align*}
\text{DFT:} & \quad [F_N]_{mn} = e^{-j2\pi mn/N}, \quad m, n \in [0, N-1] \\
\text{DCT-II:} & \quad [C_{II}]_{mn} = \cos \frac{m(2n+1)\pi}{2N}, \quad m, n \in [0, N-1] \\
\text{DCT-III:} & \quad [C_{III}]_{mn} = \cos \frac{(m+n+1)\pi}{2N}, \quad m, n \in [0, N-1] \\
\text{DCT-IV:} & \quad [C_{IV}]_{mn} = \cos \frac{m(2n+1)\pi}{4N}, \quad m, n \in [0, N-1] \\
\text{DCT-VI:} & \quad [C_{VI}]_{mn} = \cos \frac{m(2n+1)\pi}{2N+1}, \quad m, n \in [0, N] \\
\text{DCT-VII:} & \quad [C_{VII}]_{mn} = \cos \frac{(m+n+1)(2n+1)\pi}{2N+1}, \quad m, n \in [0, N] \\
\text{DST-VI:} & \quad [S_{VI}]_{mn} = \sin \frac{(m+n+1)(2n+1)\pi}{2N+1}, \quad m, n \in [0, N-1] \\
\text{DST-VII:} & \quad [S_{VII}]_{mn} = \sin \frac{(m+n+1)(2n+1)\pi}{2N+1}, \quad m, n \in [0, N] \\
\end{align*}

In the above definitions, we have intentionally omitted normalization constants (such as $\sqrt{2/N}$ and $\lambda_i = \begin{cases} 1/\sqrt{2}, & i=0, 1, \ldots, N-1 \\ 1, & i \neq 0 \end{cases}$), conventionally used in definition of DCT-II) as they don't affect factorization structures of the transforms. Sub-indices $N$ and $N+1$ indicate lengths of the transforms. We follow Wang and Hunt’s convention of coupling $N$-point DST-VI/VII with $N+1$-point DCT-VI/VII [9].

As easily noticed, transforms of types II and III, as well as VI and VII are closely related:

$$
(C_{II})^T_N = C_{III}^N; \quad (C_{VI})^T_{N+1} = C_{VII}^N; \quad (S_{VI})^T_N = S_{VII}^N.
$$

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In this section we prove the following statement.

Theorem 1. The following holds:

\[
C_{2N+1}^{\text{II}} = Q_{2N+1} \left( \begin{array}{cc}
C_N^{\text{VI}} \\
S_{2N}^{\text{VII}}
\end{array} \right) \left( \begin{array}{cc}
I_N & J_N \\
-J_N & I_N
\end{array} \right)
\]

where \( Q_{2N+1} \) is a matrix, such that when applied to a vector \( x \), it produces the following sign alterations and reordering:

\[
\hat{x}_{2i} = x_i, \quad i = 0, \ldots, N,
\]
\[
\hat{x}_{2i+1} = (-1)^i x_{N+1+i}, \quad i = 0, \ldots, N-1,
\]

and \( I_N \) and \( J_N \) are \( N \times N \) identity and order-reversal matrices respectively.

Proof. Let us consider a 2N+1-long input sequence \( x = x_0, \ldots, x_{2N} \), and apply DCT-II over it:

\[
X_k^{\text{II}} = \sum_{n=0}^{2N} x_n \cos \frac{\pi(2n+1)k}{2(2N+1)}, \quad k = 0, \ldots, 2N.
\]

We first look at even output values (\( k = 2i, i = 0, \ldots, N \)):

\[
X_{2i}^{\text{II}} = \sum_{n=0}^{2N} x_n \cos \frac{\pi(2n+1)2i}{2(2N+1)}
= \sum_{n=0}^{2N} x_n \cos \frac{\pi(2n+1)i}{2N+1}.
\]

We split this sum as follows:

\[
X_{2i}^{\text{II}} = \sum_{n=0}^{N} x_n \cos \frac{\pi(2n+1)i}{2N+1} + \sum_{n=N+1}^{2N} x_n \cos \frac{\pi(2n+1)i}{2N+1}
= \sum_{n=0}^{N} x_n \cos \frac{\pi(2(2N-n)+1)i}{2N+1} + \sum_{n=0}^{N-1} x_{2N-n} \cos \frac{\pi(2n+1)i}{2N+1},
\]

which implies that

\[
X_{2i}^{\text{II}} = X_i^{\text{VI}} + X_i^{\text{VII}}.
\]
where $X_{C^{DFT}}$ is a DCT-VI transform over the first $N+1$ elements of input sequence $x$, and $X_{mC^{DFT}}$ is a DCT-VI transform over the following input:

$$x'_n = \begin{cases} x_{2N-n}, & \text{if } n = 0, \ldots, N-1, \\ 0, & \text{if } n = N. \end{cases}$$

We now turn our attention to the odd output values ($k = 2i + 1, i = 0, \ldots, N-1$):

$$X_{C^{DFT}}^{2i+1} = \sum_{n=0}^{2N} x_n \cos \frac{(2n+1)(2i+1)}{2(2N+1)}$$

$$= (-1)^{i+1} \sum_{n=0}^{2N} x_{2N-n} \sin \frac{\pi(N-n)(2i+1)}{2N+1}$$

We split this sum as follows:

$$X_{C^{DFT}}^{2i+1} + (-1)^i \sum_{n=0}^{N-1} x_{2N-n} \sin \frac{\pi(N-n)(2i+1)}{2N+1}$$

$$= (-1)^{i+1} \sum_{n=N+1}^{2N} x_{2N-n} \sin \frac{\pi(N-n)(2i+1)}{2N+1}$$

$$= (-1)^i \sum_{n=0}^{N-1} x_{2N-n} \sin \frac{\pi(n+1)(2i+1)}{2N+1},$$

which implies that

$$X_{C^{DFT}}^{2i+1} = (-1)^{i+1} \tilde{X}_{D^{VII}} + (i)^i \tilde{X}_{D^{VII}}$$

where $\tilde{X}_{D^{VII}}$ is an $N$-point DST-VII transform over a sequence

$$\tilde{x}_n = x_{N+1-n}, \ n = 0, \ldots, N-1,$$

and $\tilde{X}_{D^{VII}}$ is an $N$-point DST-VII transform over a sequence

$$\tilde{x}_n = x_{N+1-n}, \ n = 0, \ldots, N-1.$$

By combining all these mappings we arrive at expression (1).

We present a flowgraph of the resulting mapping between DCT-II, DCT-VI, and DST-VII transforms in Figure 1.a. Only $2N$ additions, permutations and sign changes are needed to convert output of DCT-VI, and DST-VII into DCT-II.


#### 4.1. Connection to DFT

From (1) it follows that fast computation of DCT-VI and DST-VII can be reduced to computing subsets of DCT-II. According to Heideman [4] it is also known that computing of DCT-II of odd numbers is equivalent to computing same-length DFT. Considering $2N+1$-point transforms, we can summarize Heideman’s result as follows:

$$C_{2N+1}^{DFT} = H_1 \begin{pmatrix} [R(F_{2N+1})]_{rows \ 0, \ldots, N} \\ [3(F_{2N+1})]_{rows \ N+1, \ldots, 2N} \end{pmatrix} H_2,$$

where $R(F_{2N+1})$ and $3(F_{2N+1})$ denote real and imaginary parts of the DFT transform matrix of size $2N+1$, and $H_1$ and $H_2$ are some permutation and sign-inversion matrices [4].

In combination with (1) this formula shows that an $N+1$-point DCT-VI, an $N$-point DST-VII and an $2N+1$-point DCT-II can be computed by mapping to a $2N+1$-point DFT. Since many algorithms for computing of DFT are readily available (see e.g. [16]), this automatically leads to fast algorithms for computing DCT-VI and DST-VII.
4.2. Examples of fast algorithms for \( N = 4 \)

We use Winograd DFT module of length 9 shown in Figure 2.a. This particular factorization comes from [16]. By using this flowgraph and mappings (3) and (1) we easily obtain 9-point DCT-II, 5-point DCT-VI, and 4-point DST-VII. This is shown in Figure 2.b.

We note that all these algorithms are very efficient in terms of multiplicative complexity. Thus, obtained 9-point DCT-II requires only 8 non-trivial multiplications. In contrast, the least complex algorithms for computing DCT-II of size 8 (nearest dyadic-size) requires 11 multiplications [7].

The obtained 4-point DST-VII is also very efficient: it uses only 5 multiplications. This factorization is immediately suitable for implementing an integer approximation of DST-VII transform defined in HEVC standard [13].

Finally, factorization of a 5-point DCT-VI shown in Figure 2.b needs only 3 real multiplications and 2 shifts (multiplications by factors 1/2).

4.3. Fast computing of transforms of length \( 2^k N(2N + 1) \)

It is known that a transform of a composite length \( N = pq \) where \( p \) and \( q \) are co-prime, can be decomposed into a cascade of \( p \) \( q \)-point transforms and \( q \) \( p \)-point transforms followed by \( pq - p - q - 1 \) additions. This class of techniques is called Prime Factor Algorithms (PFA) [17, 18].

In Figure 1.b, we show how to compute DCT-II of length \( N(2N + 1) \). This factorization includes \( 2N + 1 \) \( N \)-point DCT-II sub-transforms, and additionally \( N \) \( 2N + 1 \)-point DCT-II transforms, which, in turn include \( N \)-point DST-VII as part of their flowgraph. Hence, a system that implements and uses \( N \)-point DCT-II and DST-VII, can easily compute an \( N(2N + 1) \) transform by reusing them. Same principle more generally applies to computing transforms of lengths \( 2^k N(2N + 1) \).

Embedded factorization structures including DST-VII blocks in flowgraphs for DCT-II can be of interest to hardware implementations, as it offers potential for reducing the area, cost, and power usage of a circuit responsible for computing transforms.

5. DISCUSSION AND CONCLUDING REMARKS

We notice that decomposition (1) looks very similar to the well-known split of even-sized DCT-II (see, e.g. [3]):

\[
C_{2N}^{II} = P_{2N} \begin{pmatrix}
C_N^{II} & C_N^{IV} J_N \\
C_N^{IV} & J_N
\end{pmatrix} \begin{pmatrix}
I_N \\
J_N
\end{pmatrix},
\]

(4)

where \( P_{2N} \) is a certain permutation matrix. This split leads to recursive construction due to reappearance of DCT-II in the upper part of decomposition. In contrast, our decomposition of \( 2N + 1 \)-point DCT-II (1) does not immediately lead to a recursion.

In Figure 3 we offer conceptual illustration of both decompositions (1) and (4). Input data samples are denoted as \( y_N, \ldots, y_1, z_0, x_1, \ldots, x_N \) in a \( 2N + 1 \)-point case (a), and \( y_N, \ldots, y_1, x_1, \ldots, x_N \) in \( 2N \)-point case (b). It is shown that the lower (right) portion of DCT-II transform becomes essentially equivalent to DST-VII (or DCT-IV) transform over residual samples \( y'_i = y_i - x_i \), while the upper (right) portion of DCT-II transform becomes essentially equivalent to DCT-VI (or DCT-II) transform over sums: \( x'_i = x_i + y_i \), \( i = 1, \ldots, N \). In the \( 2N + 1 \)-point case, the upper transform also absorbs the middle sample \( z_0 \).

This illustration may be insightful for understanding meanings of the involved transforms. For instance, in signal processing, it is customary to think of DCT-II as an approximation of KLT for 1-st order Markov source with high correlation coefficient. Decomposition in Figure 3.a shows that DST-VII, as well as DCT-VI (with some permutations and sign changes) can be understood as approximations of KLT over residual or mixed signals with progressively increasing distances between samples. Similarly, decomposition in Figure 3.b shows that in case of a \( 2N \)-sample arrangement, it is DCT-IV and DCT-II that can be understood as approximations of KLT over residual and mixed signals.

The obtained relationship (1) may also be instrumental in showing that DST-VII-based coding of Intra-prediction residual is essentially equivalent to performing L-shaped DCT-II, where one part of L-shape corresponds to boundary pixels, and the other part absorbs pixels predicted based on this boundary. A design of direction-adaptive transforms based on similar idea was proposed in [19].
6. REFERENCES


