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► **To cite this version:**

Ahcen Aliouat, Nasreddine Kouadria, Doru Florin Chiper. x-DTT: A package for calculating real and Integer Discrete Tchebichef Transform kernels based on orthogonal polynomials. *SoftwareX*, 2023, 23, pp.101441. 10.1016/j.softx.2023.101441 . hal-04171822

HAL Id: hal-04171822

<https://hal.science/hal-04171822>

Submitted on 26 Jul 2023

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Original software publication

x-DTT: A package for calculating real and Integer Discrete Tchebichef Transform kernels based on orthogonal polynomials

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ARTICLE INFO

Article history:

Received 1 March 2023

Received in revised form 6 June 2023

Accepted 12 June 2023

Dataset link: <https://github.com/ahcen23/x-DTT/blob/main/data.rar>, <https://github.com/ahcen23/x-DTT/blob/main/data.rar>

Keywords:

Discrete Tchebichef Transform

Integer Discrete Tchebichef Transform

Image compression

Features extraction

ABSTRACT

This paper presents a MATLAB implementation of the exact real Discrete Tchebichef Transform (DTT) and Integer Discrete Tchebichef Transform (IDTT) with their inverse functions. It is a simple and scalable package that is the first of its kind for MATLAB. This software provides an efficient way to advance research related to DTT, IDTT, and their inverses. It is applicable to all domains where Chebyshev moments are employed, including feature extraction, image and video compression, and other related image processing. The package supports matrix generation of any size, from 2×2 to any desired size, which makes it scalable. The presented software is easy to use and generates the exact DTT transformation kernel, integer IDTT kernel, and diagonal matrix. The performance of DTT and IDTT-based JPEG compression compared to DCT-based JPEG compression was evaluated using standard metrics. Overall, this package is a valuable tool for researchers and practitioners working in the field of image and video processing.

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Code metadata

Current code version	v1.0
Permanent link to code/repository used for this code version	https://github.com/ElsevierSoftwareX/SOFTX-D-23-00146
Permanent link to Reproducible Capsule	https://github.com/ahcen23/x-DTT
Legal Code License	The 3-Clause BSD License
Code versioning system used	/Github
Software code languages, tools, and services used	MATLAB
Compilation requirements, operating environments & dependencies	MATLAB 2009a or newer
Link to developer documentation	https://github.com/ahcen23/x-DTT/blob/main/README.md
Support email for questions	ahcen.aliouat@univ-annaba.org

1. Motivation and significance

The Discrete Tchebichef Transform (DTT) is a transformation that has lately been utilized as an alternative to the Discrete Cosine Transform (DCT) [1] for image and video compression [2–6], feature extraction [7–9] and image watermarking [10]. Generating DTT is based on Discrete Chebyshev Polynomials (DCP) which are orthogonal polynomials widely used for data fitting and approximation. The popularity of the DTT is due to its high transform efficiency and energy compaction. Other advantages are the high performance in data compression and relatively low complexity

[11,12]. Consequently, the DTT is a potential competitor to the DCT in low delay video coding standards [13–15].

Various attempts have been made, like the DCT [16,17], to minimize the spatial and temporal complexity of the DTT to enable its implementation in embedded systems [18,19]. One such attempt is the Integer DTT (IDTT), which expresses the DTT transform matrix using only integer coefficients. This approach reduces the floating-point multiplications and division operations required during the transform computation process. However, calculating the integer DTT matrix involves multiple steps, including the calculation of DCP values using a recursive-based approach. This can be a computationally challenging problem for any programming language, especially when considering the scalability of the algorithm.

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We believe that a crucial aspect in advancing research that utilizes DTT and IDTT is the ability to quickly generate scalable and variable-sized kernel with minimal effort. To address this need, we have developed the x-DTT package, which enables the generation of DTT, IDTT, and their inverses using a single MATLAB function for matrices of various sizes.

The paper is structured as follows: The next section provides a brief overview of the real and integer discrete Tchebichef transform, which is necessary to understand the paper and examples presented. Section 3 describes the software, while Section 4 outlines the software functionalities. Implementation examples are presented in Section 1. In Section 4, we compare DTT, IDTT, and DCT to demonstrate the accuracy of xDTT results in image compression applications. Finally, Section 6 provides concluding remarks.

2. Real DTT and IDTT overview

This section offers a concise summary of the mathematical basis for the DTT and IDTT computations. Additionally, we present the depiction of the DTT based on the essential elements of the DCP. The DCP can be expressed as Δ , where Δ is the forward difference operator.

$$t_k(n) = k! \Delta^k \left[\binom{n}{k} \binom{n-N}{k} \right] \text{ for } n = 0, 1 \dots N-1 \quad (1)$$

Some properties of the Chebyshev polynomials are given below.

2.1. Properties of DCP

(1) **Orthogonality:** The DCPs are orthogonal with respect to unit weight

$$\sum_{n=0}^{N-1} t_k(n) t_m(n) = \delta_{km} d_k, \quad (2)$$

where d_k is the squared norm of the DCP defined as

$$d(k, N) = \sum_{n=0}^{N-1} [t_k(n)]^2 = \frac{N(N^2-1)(N^2-2^2) \dots (N^2-k^2)}{2k+1} \quad (3)$$

(2) **Symmetry:** The DCP satisfies the symmetry property

$$t_k(N-1-n) = (-1)^k t_k(n) \quad (4)$$

(3) **Recurrence Relation:** Based on Eq. (1), the DCP satisfies a three-term recurrence relation that can be expressed as follows:

$$t_0(n) = 1, t_1(n) = 2n - N + 1 \quad (5)$$

$$t_k(n) = \left[\frac{2k-1}{k} t_k(1) \right] t_{k-1}(n) - \left[\binom{k-1}{k} (N^2 - (k-1)^2) \right] t_{k-2}(n) \quad (6)$$

2.2. Derivation of real DTT from normalized DCP

DCPs, while orthogonal, can lead to numerical instability and error magnification. To address this issue, we perform normalization using Eq. (2) to obtain $\tau_k(n) = \frac{t_k(n)}{\sqrt{d(n, N)}}$. Here, t_k represents the classical DCP, while the orthonormal DCP is denoted by τ_k . The set $\{\tau_k(n)\}$ is orthonormal, as $\forall n, m : \langle \tau_n, \tau_m \rangle = \delta_{n,m}$, $\{\tau_k(n)\}$. We can obtain the orthonormal DCP iteratively using τ_k in Eq. (7), which can be expressed concisely as:

$$\tau_k(n) = (a_1 n + a_2) \tau_{k-1}(n) + a_3 \tau_{k-2}(n) \quad (7)$$

with

$$\tau_0(n) = \frac{1}{\sqrt{N}} \quad (8)$$

and

$$\tau_1(n) = (2n + 1 - N) \sqrt{\frac{3}{N(N^2-1)}} \quad (9)$$

where

$$a_1 = \frac{2}{k} \sqrt{\frac{4k^2-1}{N^2-k^2}}, \quad a_2 = \frac{1-N}{k} \sqrt{\frac{4k^2-1}{N^2-k^2}} \quad (10)$$

and

$$a_3 = \frac{1-k}{k} \cdot \frac{2k+1}{2k-3} \sqrt{\frac{N^2-(k-1)^2}{N^2-k^2}} \quad (11)$$

2.3. Definition of the DTT

The N -point 1-D DTT, $Y(k)$ of an input sequence $x(n)$, for $n, k = 0, 1 \dots N-1$ and the inverse 1-D DTT can be defined as

$$Y(k) = \sum_{n=0}^{N-1} \tau(k, n) x(n) \text{ and } x(n) = \sum_{k=0}^{N-1} \tau(k, n) Y(k), \quad (12)$$

where the kernel $\tau(k, n)$ represents $\tau_k(n)$, the orthonormal basis of DCP, which is given by (7).

2.4. Formulation of IDTT

The 2-D DTT can be represented in matrix form [20] as shown in

$$Y = \tau X \tau' = (\tau(\tau X))', \quad (13)$$

Here, X is the 2-D input matrix, τ is the Chebyshev basis, and Y is the 2-D matrix of transform coefficients.

As seen in (13), the 2-D DTT of input data X of size $N \times N$, given by Eq. (5), can also be expressed in the form shown above.

$$Y = \tau X \tau', \quad (14)$$

where τ is the ' $N \times N$ ' DTT kernel and can be factorized as shown below to separate the integer and float numbers. Consider a variable

$$q(k) = \frac{p(k)}{\sqrt{d(k, N)}}, \quad (15)$$

where

$$p(k) = \begin{cases} k!k & \text{for } k = 1 \\ k!(k+1) & \text{for } k = 0, 2, 4, 6, \dots, N \\ k!(N-k) & \text{for } k = 3, 5, 7 \dots N-1 \end{cases} \quad (16)$$

and $d(k, N)$ is the squared norm of the DTP given by

$$d(k, N) = \frac{N(N^2-1)(N^2-2^2) \dots (N^2-k^2)}{2k+1} \quad (17)$$

Let \mathbf{Q} be a diagonal matrix with $q(k)$ as the diagonal element of the k^{th} row. The transformation matrix τ can be expressed as

$$\tau = \mathbf{Q} \hat{\tau} \\ \hat{\tau} = \mathbf{Q}^{-1} \tau \quad (18)$$

It is worth noting that the matrix \mathbf{Q} can be shifted in various scenarios during the process without introducing any computational overhead. The linear representation of the DTP polynomials of the order 16×16 and 32×32 are shown in Figs. 1 and 2.

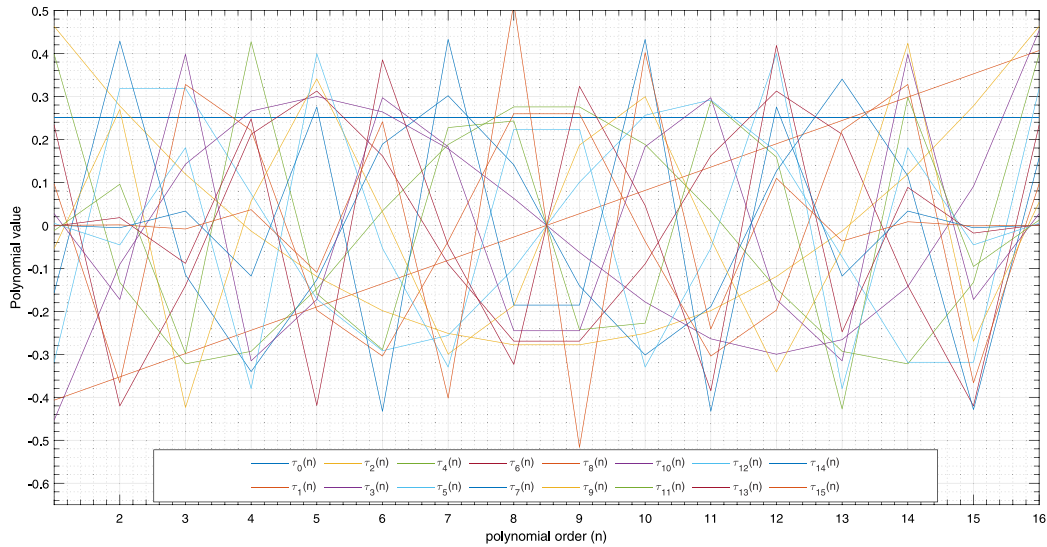


Fig. 1. An illustration of the first 16 orthonormal DCPs ($\tau_0(n)$ to $\tau_{15}(n)$).

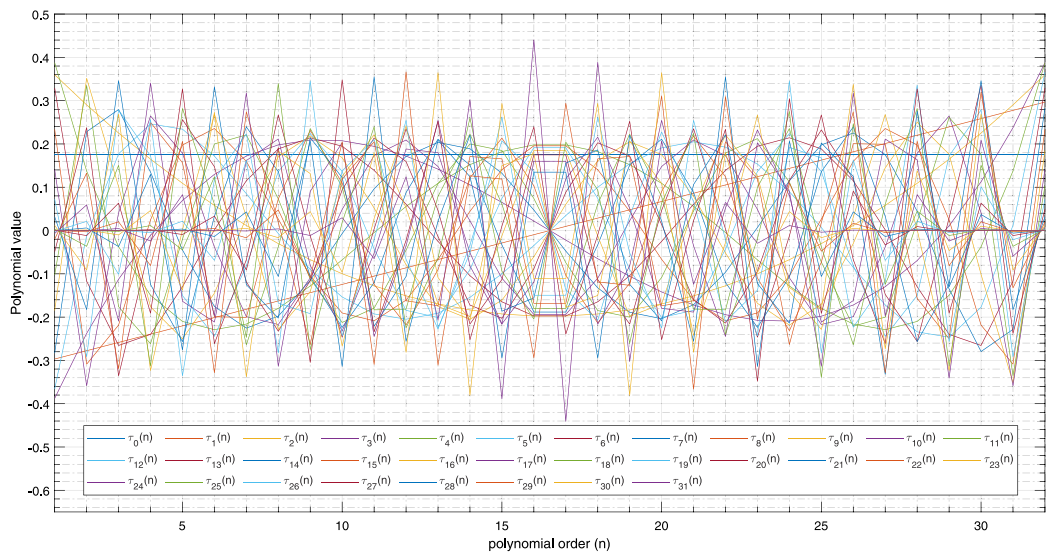


Fig. 2. An illustration of the first 32 orthonormal DCPs ($\tau_0(n)$ to $\tau_{31}(n)$).

3. Software description

The proposed software package includes a function named DTT, which takes an input N , an integer value representing the size of the generated matrix. The DTT function calculates the recursive polynomials based on the equations described in Section 2 and outputs three matrices:

- **Matrix τ :** This output is a real matrix that represents the exact DTT transformation kernel of size $N \times N$.
- **Matrix $\hat{\tau}$:** This output represents an integer matrix that is the integer kernel of the IDTT.
- **Matrix $diag$:** This output represents the diagonal matrix as shown previously in Eq. (18).

If we let $diag$ represent the Q matrix, the function can be called using the following line of MATLAB code: $[\tau, \hat{\tau}, diag] = DTT(N)$.

Executing this call generates the three matrices based on the theory shown in previous sections. The DTT function is easy to use, just like the integrated DCT function in MATLAB

The function can generate kernels of any size, ranging from 2×2 to a maximum tested size of 256×256 , which is considerably high. However, it should be noted that generating a large kernel may take a long time due to the exponential complexity of the DCP generation process.

4. Comparison with DCT in a JPEG image compression

To confirm the correctness of the generated kernel matrices, we compared the image compression performance of a standard JPEG technique using DCT, quantization, zigzag, and Huffman encoding with a modified JPEG chain, where DCT transformation is replaced by DTT matrices generated using the proposed software.

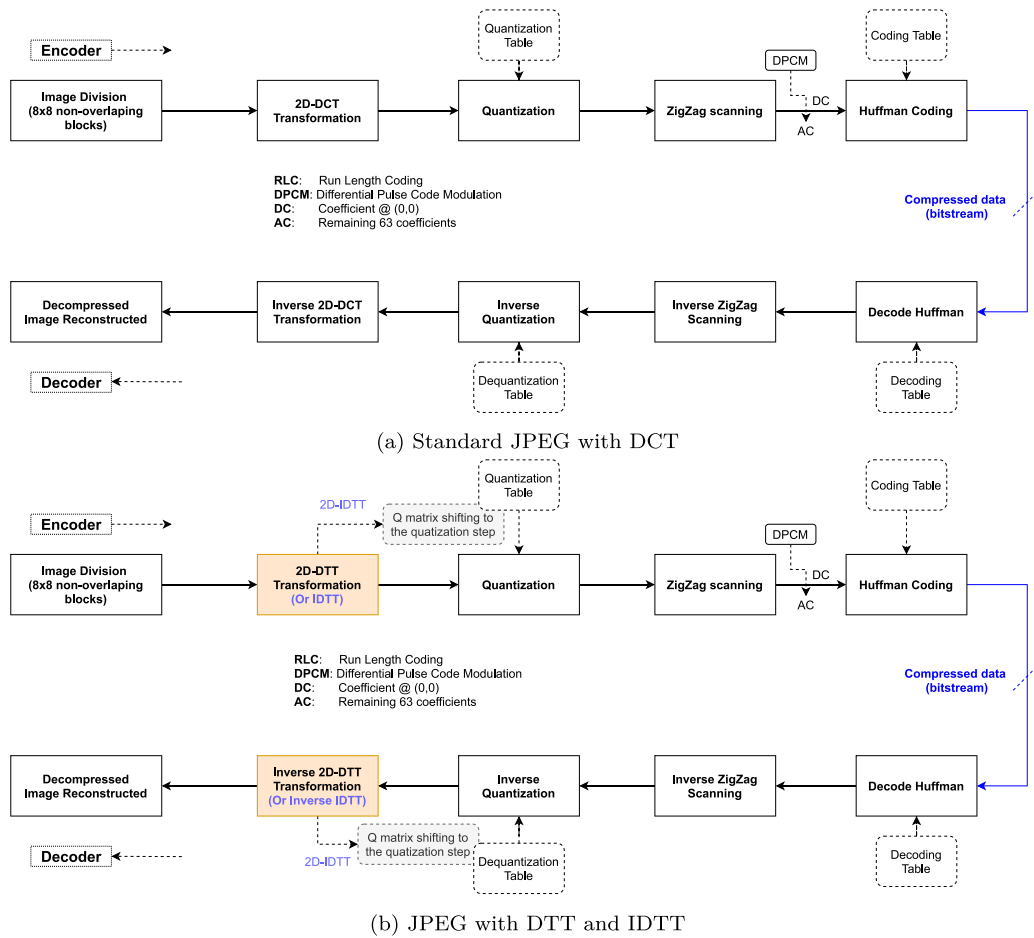


Fig. 3. JPEG chain using DCT and DTT.



Fig. 4. Used images dataset (part from USC-SIPI Dataset [21]) for experimentation: *lena* 512×512 , *cameraman* 256×256 and *boat* 512×512 .

Fig. 3(a) shows the scheme and steps of the JPEG transform and the replacement of DCT with DTT in Fig. 3(b) (see Fig. 3).

The used standard images are shown in Fig. 4. We use peak signal-to-noise ratio (PSNR) and Structural similarity (SSIM) with multiple bit-per-pixel (bpp) values for quality evaluation. Where bpp represented the ratio of the compression size to the number of original bits formulated as $bpp =$

S_{comp}/N_{pixels} , where S_{comp} is the size of the compressed data and N_{pixels} is the number of pixels to compress. The formula of PSNR has been presented in [22], while the SSIM is described in [23]. The experiments are performed using MATLAB 2021a software running on a quad-core i7 2.5 GHz laptop.

Figs. 5, 6, and 7 show the compression results for the dataset used in the experiment when using DTT and IDTT-based JPEG

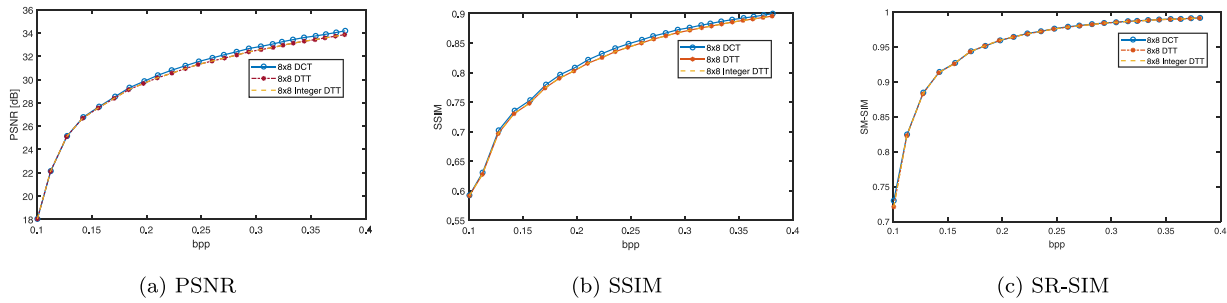


Fig. 5. Compression results comparison for **Lena** image.

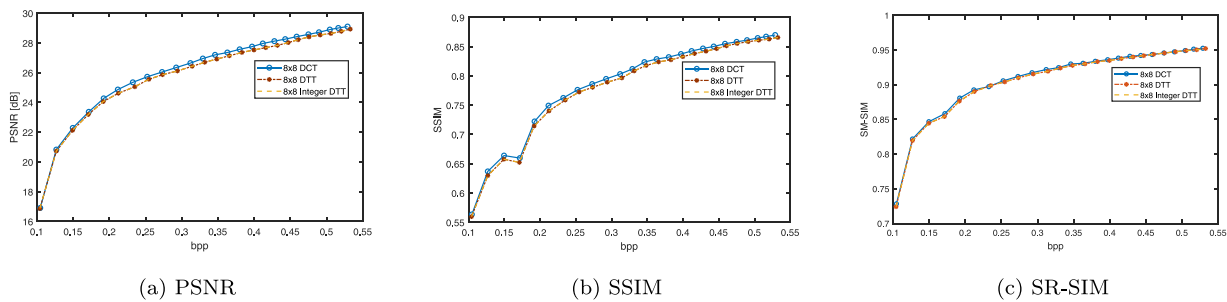


Fig. 6. Compression results comparison for **Cameraman** image.

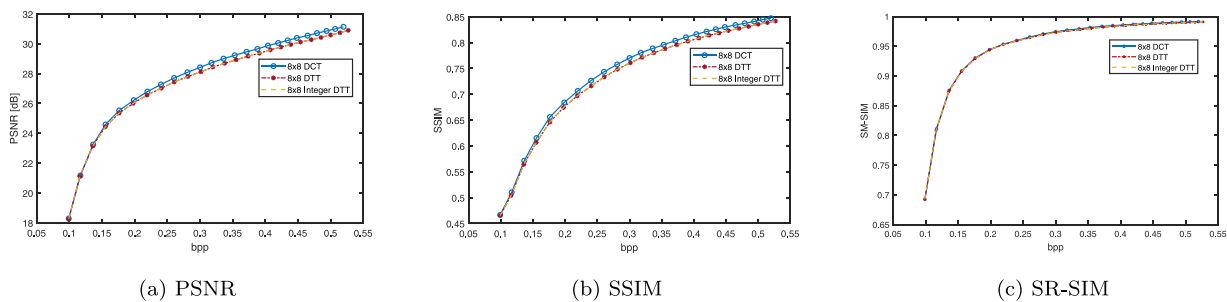


Fig. 7. Compression results comparison for **Boat** image.

compression. It is observed that the qualitative results for DTT and IDTT-based compression are almost similar to those of DCT-based compression.

However, for relatively high bpp values, DCT slightly outperforms DTT and IDTT for all the used sequences. These results are consistent with the theoretical findings reported in the literature regarding the difference between DTT and DCT for natural images [11].

Fig. 8 shows the percentage of the error between the original image and the compression image using different transformation kernels and for different compression rates.

5. Impact

The x-DTT package offers a user-friendly solution for generating DCPs utilized in DTT, IDTT, and their corresponding inverses. This enables researchers to effortlessly engage in and advance the fields of image compression [2], Feature Extraction [8,9], and

fast transform algorithms [3]. The versatility of x-DTT lies in its capability to generate orthonormal Chebyshev polynomials of various sizes. The impact of this remarkable feature lies in its ability to simplify the development of theories behind the algorithms based on DCPs. Notably, the application of DTT and IDTT as compression techniques continues to be a prominent and rapidly evolving research area in the literature. It has been identified as a promising candidate for replacing DCT-based compression techniques. We firmly believe that the x-DTT package has the potential to significantly contribute to the advancement of this field.

6. Conclusion

The x-DTT software presented in this paper is a user-friendly and powerful package for generating the kernels of exact and integer DTT transformations in MATLAB. This package enables researchers to easily develop, utilize, and study the features and

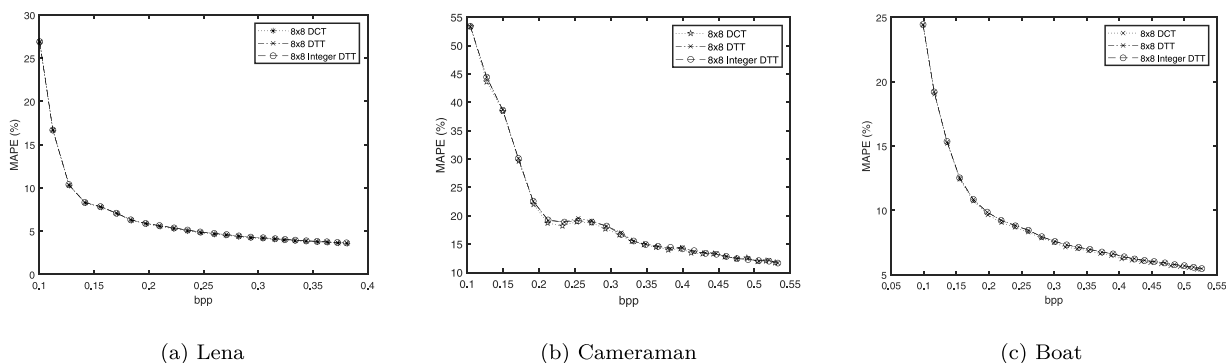


Fig. 8. Average percentage error for different compression rate for each transform.

```

MATLAB Command Window
27 décembre 2022
Page 1
03:57:20

>> [a,b,c]=DTT(2)

a =

    0.7071    0.7071
   -0.7071    0.7071

b =

     1     1
    -1     1

c =

    1.4142     0
         0    1.4142

>>
    
```

Fig. A.9. 2 × 2 DTT results.

concepts related to the DTT transform. The strength of the proposed package lies in its simplicity and scalability for almost all transform sizes. In future work, we plan to add other algorithms for fast computing of the DTT in the same package.

Declaration of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

Data availability

The raw data required to reproduce the above findings are available to download from <https://github.com/ahcen23/x-DTT/blob/main/data.rar>. The processed data required to reproduce the

above findings are also available to download from <https://github.com/ahcen23/x-DTT/blob/main/data.rar>.

Appendix A. Examples: output matrices of the DTT function

In this section, we demonstrate the generation of kernels of different sizes using our function. In the examples below, matrix *a* represents τ , matrix *b* represents $\hat{\tau}$, and matrix *c* represents the diagonal matrix *diag*.

Figs. A.9–A.11 show examples for generating 2 × 2, 4 × 4 and 8 × 8 DTT kernels. Figs. A.12–A.14 show the exact DTT matrix τ , the IDTT $\hat{\tau}$, and the diagonal matrix *diag* for 16 × 16 size.

Appendix B. Supplementary data

Supplementary material related to this article can be found online at <https://doi.org/10.1016/j.softx.2023.101441>.

```

MATLAB Command Window Page 1

>> [a,b,c]=DTT(4)

a =

    0.5000    0.5000    0.5000    0.5000
   -0.6708   -0.2236    0.2236    0.6708
    0.5000   -0.5000   -0.5000    0.5000
   -0.2236    0.6708   -0.6708    0.2236

b =

     1     1     1     1
    -3    -1     1     3
     1    -1    -1     1
    -1     3    -3     1

c =

    2.0000         0         0         0
         0    4.4721         0         0
         0         0    2.0000         0
         0         0         0    4.4721

>>
    
```

Fig. A.10. 4 × 4 DTT results.

```

MATLAB Command Window Page 1

>> [a,b,c]=DTT(8)

a =

    0.3536    0.3536    0.3536    0.3536    0.3536    0.3536    0.3536    0.3536
   -0.5401   -0.3858   -0.2315   -0.0772    0.0772    0.2315    0.3858    0.5401
    0.5401    0.0772   -0.2315   -0.3858   -0.3858   -0.2315    0.0772    0.5401
   -0.4308    0.3077    0.4308    0.1846   -0.1846   -0.4308   -0.3077    0.4308
    0.2820   -0.5238   -0.1209    0.3626    0.3626   -0.1209   -0.5238    0.2820
   -0.1498    0.4922   -0.3638   -0.3210    0.3210    0.3638   -0.4922    0.1498
    0.0615   -0.3077    0.5539   -0.3077   -0.3077    0.5539   -0.3077    0.0615
   -0.0171    0.1195   -0.3585    0.5974   -0.5974    0.3585   -0.1195    0.0171

b =

     1     1     1     1     1     1     1     1
    -7    -5    -3    -1     1     3     5     7
     7     1    -3    -5    -5    -3     1     7
    -7     5     7     3    -3    -7    -5     7
     7   -13    -3     9     9    -3   -13     7
    -7    23   -17   -15    15    17   -23     7
     1    -5     9    -5    -5     9    -5     1
    -1     7   -21    35   -35    21    -7     1

c =

    2.8284         0         0         0         0         0         0         0
         0    12.9615         0         0         0         0         0         0
         0         0    12.9615         0         0         0         0         0
         0         0         0    16.2481         0         0         0         0
         0         0         0         0    24.8193         0         0         0
         0         0         0         0         0    46.7333         0         0
         0         0         0         0         0         0    16.2481         0
         0         0         0         0         0         0         0    58.5833

>>
    
```

Fig. A.11. 8 × 8 DTT results.


```

MATLAB Command Window

>> [a,~,~]=DTT(16)
a =
0.2500 0.2500 0.2500 0.2500 0.2500 0.2500 0.2500 0.2500 0.2500 0.2500 0.2500 0.2500 0.2500 0.2500 0.2500 0.2500
-0.4067 -0.3525 -0.2983 -0.2440 -0.1898 -0.1356 -0.0813 -0.0271 0.0271 0.0813 0.1356 0.1898 0.2440 0.2983 0.3525 0.4067
0.4631 0.2779 0.1191 -0.0132 -0.1191 -0.1985 -0.2514 -0.2779 -0.2779 -0.2514 -0.1985 -0.1191 -0.0132 0.1191 0.2779 0.4631
-0.4532 -0.0906 0.1424 0.2660 0.2998 0.2640 0.1783 0.0628 -0.0628 -0.1783 -0.2640 -0.2998 -0.2660 -0.1424 0.0906 0.4532
0.3981 -0.1327 -0.3223 -0.2931 -0.1473 0.0335 0.1881 0.2756 0.2756 0.1881 0.0335 -0.1473 -0.2931 -0.3223 -0.1327 0.3981
-0.3185 0.3185 0.3185 0.0735 -0.1715 -0.2918 -0.2562 -0.1002 0.1002 0.2562 0.2918 0.1715 -0.0735 -0.3185 -0.3185 0.3185
0.2335 -0.4202 -0.1401 0.2119 0.3125 0.1616 -0.0898 -0.2694 -0.2694 -0.0898 0.1616 0.3125 0.2119 -0.1401 -0.4202 0.2335
-0.1569 0.4288 -0.1150 -0.3403 -0.1263 0.1890 0.3017 0.1408 -0.1408 -0.3017 -0.1890 0.1263 0.3403 0.1150 -0.4288 0.1569
0.0964 -0.3664 0.3278 0.2210 -0.1973 -0.3041 -0.0371 0.2596 0.2596 -0.0371 -0.3041 -0.1973 0.2210 0.3278 -0.3664 0.0964
-0.0539 0.2697 -0.4238 0.0563 0.3408 0.0314 -0.2993 -0.1867 0.1867 0.2993 -0.0314 -0.3408 -0.0563 0.4238 -0.2697 0.0539
0.0272 -0.1725 0.3982 -0.3152 -0.1725 0.2970 0.1829 -0.2452 -0.2452 0.1829 0.2970 -0.1725 -0.3152 0.3982 -0.1725 0.0272
-0.0123 0.0957 -0.2976 0.4273 -0.1588 -0.2892 0.2275 0.2429 -0.2429 -0.2275 0.2892 0.1588 -0.4273 0.2976 -0.0957 0.0123
0.0048 -0.0454 0.1808 -0.3800 0.3993 -0.0532 -0.3297 0.2233 0.2233 -0.3297 -0.0532 0.3993 -0.3800 0.1808 -0.0454 0.0048
-0.0016 0.0180 -0.0884 0.2479 -0.4187 0.3851 -0.0462 -0.3235 0.3235 0.0462 -0.3851 0.4187 -0.2479 0.0884 -0.0180 0.0016
0.0004 -0.0056 0.0333 -0.1180 0.2754 -0.4328 0.4328 -0.1855 -0.1855 0.4328 -0.4328 0.2754 -0.1180 0.0333 -0.0056 0.0004
-0.0001 0.0012 -0.0084 0.0365 -0.1096 0.2411 -0.4019 0.5167 -0.5167 0.4019 -0.2411 0.1096 -0.0365 0.0084 -0.0012 0.0001
>>
    
```

Fig. A.12. The “a” matrix results for 16 × 16 DTT.

```

MATLAB Command Window

>> [~,~,~]=DTT(16),num2str(b,'%10d')
ans =
16x154 char array
' 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1'
'-15 -13 -11 -9 -7 -5 -3 -1 1 3 5 7 9 11 13 15'
' 35 21 9 -1 -9 -15 -19 -21 -21 -19 -15 -9 -1 9 21 35'
'-35 -7 11 21 23 20 14 5 -5 -14 -20 -23 -21 -11 7 35'
'-273 -91 -221 -201 -101 23 129 189 189 129 23 -101 -201 -221 -91 273'
'-273 273 273 63 -147 -250 -86 220 -86 220 250 147 -63 -273 -273 273'
' 715 -1287 -429 649 957 495 -275 -825 -825 -275 495 957 649 -429 -1287 715'
'-715 1954 -524 -1551 -576 862 1375 642 -642 -1375 -862 576 1551 524 -1954 715'
' 715 -2717 2431 1639 -1463 -2255 -275 1925 1925 -275 -2255 -1463 1639 2431 -2717 715'
'-715 3575 -5618 746 4518 416 -3968 -2475 2475 3968 -416 -4518 -746 5618 -3575 715'
' 273 -1729 3991 -3159 -1729 2977 1833 -2457 -2457 1833 2977 -1729 -3159 3991 -1729 273'
'-273 2129 -6622 9508 -3533 -6435 5062 5405 -5405 -5062 6435 3533 -9508 6622 -2129 273'
' 35 -329 1309 -2751 2891 -385 -2387 1617 1617 -2387 -385 2891 -2751 1309 -329 35'
'-35 390 -1916 5369 -9070 8342 -1001 -7007 7007 1001 -8342 9070 -5369 1916 -390 35'
' 1 -13 77 -273 637 -1001 1001 -429 -429 1001 -1001 637 -273 77 -13 1'
'-1 15 -105 455 -1365 3003 -5005 6435 -6435 5005 -3003 1365 -455 105 -15 1'
>>
    
```

Fig. A.13. The “b” matrix results for 16 × 16 DTT.

```

MATLAB Command Window

>> [~,~,~]=DTT(16),num2str(c,'%10d')
ans =
16x181 char array
'4 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0'
'03.687818e+01 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0'
'0 07.557777e+01 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0'
'0 0 07.722096e+01 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0'
'0 0 0 06.857755e+02 0 0 0 0 0 0 0 0 0 0 0 0 0 0'
'0 0 0 0 08.570776e+02 0 0 0 0 0 0 0 0 0 0 0 0 0'
'0 0 0 0 0 03.062666e+03 0 0 0 0 0 0 0 0 0 0 0 0'
'0 0 0 0 0 0 04.557937e+03 0 0 0 0 0 0 0 0 0 0 0'
'0 0 0 0 0 0 0 07.415665e+03 0 0 0 0 0 0 0 0 0 0'
'0 0 0 0 0 0 0 0 01.325619e+04 0 0 0 0 0 0 0 0'
'0 0 0 0 0 0 0 0 0 01.002198e+04 0 0 0 0 0 0 0'
'0 0 0 0 0 0 0 0 0 0 02.225339e+04 0 0 0 0 0'
'0 0 0 0 0 0 0 0 0 0 0 07.240098e+03 0 0 0 0'
'0 0 0 0 0 0 0 0 0 0 0 0 02.166062e+04 0 0'
'0 0 0 0 0 0 0 0 0 0 0 0 0 02.312765e+03 0'
'0 0 0 0 0 0 0 0 0 0 0 0 0 0 01.245462e+04'
>>
    
```

Fig. A.14. The “c” matrix results for 16 × 16 DTT.

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