T-ALGEBRA: AN IMPLEMENTATION OF MATRIX PARADIGM/PACKAGE OVER A NOVEL SEMISIMPLE COMMUTATIVE ALGEBRA

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https://github.com/liaoliang2020/talgebra

Contents

1. In	ntroduction	9
		3
2. Fr	unctions for T-scalars	S
2.1.	T-SCALAR PRODUCT: tproduct	3
2.2.	FOURIER MATRIX: fourier_matrix	8
2.3.	INVERSE FOURIER MATRIX: ifourier_matrix	S
2.4.	IDENTITY T-SCALAR: E_T	10
2.5.	Zero T-scalar: Z_T	11
2.6.	PSEUDO-INVERSE OF A T-SCLAAR: tpinv_tscalar	12
2.7.	NORM AND ANGLE OF A T-SCALAR: tnorm_angle	13
2.8.	CONJUGATE OF A T-SCALAR: tconj_tscalar	14
2.9.	REAL PART OF A T-SCALAR: treal_part_tscalar	16
2.10.	IMAGINARY PART OF A T-SCALAR: timg_part_tscalar	17
2.11.	Power of A T-scalar: tpower	18
2.12.	IDEMPOTENT T-SCALAR: idempotent	20
2.13.	RANDOM T-SCALAR: randn_tscalar	21
2.14.	GENERALIZED EXPONENTIAL: texp	22
2.15.	NATURAL LOGARITHM OF A T-SCALAR: tlog	24
2.16.	ABSOLUTE VALUE OF A T-SCALAR: tabs_tscalar	25

2.17.	ANGLE OF A T-SCALAR: tangle_tscalar	26		
2.18.	WHETHER INPUT IS A SELF-CONJUGATE: is_self_conjugate	27		
2.19.	WHETHER INPUT IS A NONNEGATIVE: is_nonnegative	29		
2.20.	Infinum of A set of two Self-conjugate T-scalars: tinf	30		
3. Fur	nctions for T-matrices	32		
3.1. N	MULTHPLICATION OF TWO T-MATRICES: tmultiplication	32		
3.2. N	MULTIPLICATION OF THREE T-MATRICES: tmultiplication_arg3	34		
3.3. R	RANK OF A T-MATRIX: trank	35		
3.4. I	DENTITY T-MATRIX: teye	36		
3.5. C	CONJUGATE TRANSPOSE A T-MATRIX: tctranspose	37		
3.6. I	NVERSE OF A T-MATRIX: tinv	38		
3.7. P	SEUDO-INVERSE OF A T-MATRIX: tpinv	39		
3.8. C	COMPACT TENSORIAL SINGULAR VECTOR DECOMPOSITION OF A T-MATRIX: tsvd	43		
3.9. I	DETERMINANT OF A T-MATRIX: tdet	44		
3.10.	DIAGONAL ELEMENTS OF A T-MATRIX: tdiag	46		
3.11.	SQUARE ROOT OF A T-MATRIX: tsqrtm	49		
3.12.	NORM OF A T-MATRIX: tnorm	50		
3.13.	EIGENVALUES AND EIGENVECTORS OF A T-MATRIX: teig	53		
3.14.	TRACE OF A T-MATRIX: ttrace	54		
3.15.	NAIVE TRANSPOSE OF A T-MATRIX: tsimtranspose	55		
3.16.	Generalized Signular Value Thresholding of a t-matrix: tsvt	58		
3.17.	PRODUCT OF THE T-SCALAR ENTRIES OF A T-MATRIX: tprod	59		
4. Fur	nctions for Generalized Tensors	61		
4.1. C	GENERALIZED TENSOR MULTIPLICATION: gtensor_multiplication	61		
5. Demos				
5.1. L	OW-RANK APPROXIMATION OF A HIGHER-ORDER MATRIX	61		
References				

1. Introduction

In the big-data deluge era, the canonical matrix and tensor paradigm over an algebraically closed field plays an essential role in many areas, including but not limited to machine learning, computer vision, pattern analysis, and statistic inference. Under the canonical matrix and tensor paradigm, observed data are given in the form of high-order arrays of canonical scalars (i.e., real or complex numbers). For example, an RGB image is a real number array of order three, two orders for the image's spatial measures, and a third for the image's spectral measures. An RGB image is also said to have three modes or three-way. A color video sequence of images is of order four, with three orders for spatial-spectral measures and the fourth-order chronological tempo.

Therefore, it is a natural question of whether there exists an extension of the field \mathbb{C} over which a generalized matrix and tensor paradigm can be established and backward-compatible to the canonical paradigm over a field. Fortunately, the answer is yes, but one had to sacrifice at least one of the axioms of a field to obtain something extended.

Among these efforts trying to generalizing the field of complex numbers, well-known is Hamilton's \mathbb{H} of quaternions, which, up to isomorphism, is a real division subalgebra of the matrix algebra $M_2(\mathbb{C})$ [Ham48, Roz88, HN11]. However, the multiplication of quaternions is not commutative.

Most hypercomplex number systems, including Hamilton's quaternions, are all subalgebras of Clifford algebra, and the fruits of generating complex numbers to obtain something extended. However, Clifford algebra's hypercomplex number systems are not suitable for general data analytics partially because they are either non-commutative or incompatible with many canonical notions such as euclidean norms. These hypercomplex number systems so far only find narrow niches in geometry and geometry-related branches of physics and computer sciences [Hes03, AS⁺04].

2. Functions for T-scalars

A matrix over C is called a tmatrix. In this article, its canonical counterpart over complex numbers is called a "canonical matrix," or just a "matrix."

The matrix paradigm over t-algebra C generalizes many notions of canonical matrices over complex numbers. In the following, we show how to use these t-matrix generalizations with the their MATLAB implementations.

2.1. T-SCALAR PRODUCT: tproduct. The following function tproduct computes the t-scalar product of two t-scalars tscalar01 and tscalar02.

The multi-way circular convolution can compute the product of two scalars in the spatial domain.

More specifically, let C be the set of all $I_1 \times \cdots \times I_N$ complex arrays, namely $C \equiv \mathbb{C}^{I_1 \times \cdots \times I_N}$. For all $A_T, B_T \in C \equiv \mathbb{C}^{I_1 \times \cdots \times I_N}$, the t-scalar multiplication $C_T \doteq A_T \circ B_T$ is defined by the N-way circular convolution of A_T and B_T as follows.

Definition 2.1 (T-scalar multiplication, i.e., N-way circular convolution). [LM20a] The (i_1, \dots, i_N) -th complex entry $(C_T)_{i_1, \dots, i_N}$ of $C_T \in C \equiv \mathbb{C}^{I_1 \times \dots \times I_N}$ is given by

$$(C_T)_{i_1,\dots,i_N} \doteq \sum_{(m_1,\dots,m_N)\in[I_1]\times\dots\times[I_N]} (A_T)_{m_1,\dots,m_N} \cdot (B_T)_{m'_1,\dots,m'_N} \in \mathbb{C}$$

$$\forall (i_1,\dots,i_N)\in[I_1]\times\dots\times[I_N] . \tag{1}$$

where $m'_n = \text{mod}(i_n - m_n, I - n) + 1$ for all $n \in [N]$.

The t-scalar product (i.e., N-way circular convolution) can be equivalently obtained in the Fourier domain by the Hadamard product. The computation in the Fourier domain is more efficient and is adopted in the following function tproduct.

FUNCTION 1. **tproduct.m**: t-scalar product of two sclalars

```
function result = tproduct(tscalar01, tscalar02)
%This function computes the product of two t-scalars
assert(isequal(size(tscalar01), size(tscalar02)));

%T-scalar product computed in the fourier domain by the Hadamard
%product
result = ifftn(fftn(tscalar01) .* fftn(tscalar02));
end
```

The input arguments tscalar01 and tscalar02 are complex arrays of the same shape. The line confirms their shapes (i.e., "size" by the MATLAB colloquialism) are identical.

In line 4, multi-way Fourier transform fftn and its inverse transform ifftn are invoked. The multi-way Fourier transform and its inverse transform can be equivalently computed by MATLAB's single-way Fourier transform fft and is demonstrated as follows.

FUNCTION 2. Demo: multi-way Fourier transform fftn implemented by single-way Fourier transforms fft

```
function fourier_transformed_tscalar = fftn(tscalar)
assert(isnumeric(tscalar));
number_of_index_axes = ndims(tscalar);
%Compute single-way fourier transform along each index axis.
fourier_transformed_tscalar = tscalar;
for i = 1: number_of_index_axes
fourier_transformed_tscalar = fft(fourier_transformed_tscalar, [], i);
end
end
```

Similarly, the multi-way inverse transform ifftn is computed by single-way inverse transforms ifft along different axes as follows.

FUNCTION 3. Demo: multi-way inverse Fourier transform ifftn implemented by single-way Fourier transforms ifft

```
function tscalar = ifftn(fourier_transformed_tscalar)
assert(isnumeric(fourier_transformed_tscalar));
number_of_index_axes = ndims(fourier_transformed_tscalar);
%Compute single-way inverse fourier transform along each index axis.
tscalar = fourier_transformed_tscalar;
for i = 1: number_of_index_axes
    tscalar = fft(tscalar, [], i);
end
end
```

SYNTAX

```
C = tproduct(A, B)
```

DESCRIPTION

Output C is the multi-way circular convolution of two arrays A and B. Output C and inputs A and B are numeric arrays of the same size. This function generalizes the canonical multiplication of two complex numbers.

EXAMPLES

Compute the multiplication of two t-scalars A and B as follows.

```
\gg A = ones(3)
A =
                     1
      1
             1
      1
             1
                     1
      1
             1
                     1
>> B = 2 * ones(3)
B =
      2
             2
                     2
      2
             2
                     2
      2
             2
                     2
>> tproduct(A, B)
ans =
     18
            18
                    18
     18
            18
                    18
     18
            18
                    18
```

Compute the multiplication of two random t-scalars A and B, both of them in the form of a 3×3 real array.

```
>> A = rand(3)
A =
    0.4984
               0.5853
                         0.2551
    0.9597
                         0.5060
               0.2238
    0.3404
               0.7513
                          0.6991
>> B = rand(3)
B =
    0.8909
               0.1386
                         0.8407
               0.1493
    0.9593
                         0.2543
    0.5472
               0.2575
                          0.8143
>> C = tproduct(A, B)
C =
    2.4311
               2.5358
                          2.7127
                          2.6006
    2.7564
               2.5265
    2.8999
               2.4472
                          2.4721
```

Compute the multiplication of two random t-scalars A and B, both of them in the form of a 2×3 complex array.

```
>> A = randn(2, 3) + i * randn(2, 3)
A =
   0.5377 - 0.4336i - 2.2588 + 3.5784i
                                       0.3188 - 1.3499i
   1.8339 + 0.3426i
                      0.8622 + 2.7694i
                                       -1.3077 + 3.0349i
>> B = randn(2, 3) + i * randn(2, 3)
B =
   0.7254 + 1.4090i
                    0.7147 + 0.6715i
                                       -0.1241 + 0.7172i
  -0.0631 + 1.4172i
                    -0.2050 - 1.2075i
                                        1.4897 + 1.6302i
>> C = tproduct(A, B)
C =
  -0.0499 + 6.6934i - 15.9137 + 1.0122i - 0.5178 + 0.7970i
  -14.5318 + 6.7276i - 7.0977 - 2.0930i 1.2429 + 6.9517i
```

Compute the multiplication of two random t-scalars A and B, both of them in the form of a $2 \times 2 \times 3$ real array.

```
>> A = randn(2, 2, 3)
```

$$A(:,:,1) = \\ -0.0068 & -0.7697 \\ 1.5326 & 0.3714$$

$$A(:,:,2) = \\ -0.2256 -1.0891 \\ 1.1174 0.0326$$

$$A(:,:,3) = 0.5525 1.5442 1.1006 0.0859$$

$$>> B = randn(2, 2, 3)$$

$$B(:,:,1) = \\ -0.1774 & 1.4193 \\ -0.1961 & 0.2916$$

$$B(:,:,2) = 0.1978 -0.8045 \\ 1.5877 0.6966$$

$$B(:,:,3) = 0.8351 \quad 0.2157 \\ -0.2437 \quad -1.1658$$

$$C(:,:,1) = -1.3426 -1.0037$$

 $4.3985 5.0231$

MULTI-WAY FOURIER TRANSFORM AND INVERSE TRANSFORM. Liang Liao and Stephen John Maybank give a rigorous definition of multi-way Fourier transform and inverse transform in [LM20a, LM20b]. For convenience, the definition is organized as follows.

Definition 2.2 (MULTI-WAY FOURIER TRANSFORM). The Fourier transform is a linear isomorphism defined by the N-mode multiplication of tensors, which sends each element $X_T \in C \equiv \mathbb{C}^{I_1 \times \cdots \times I_N}$ to $\tilde{X}_T \in C \equiv \mathbb{C}^{I_1 \times \cdots \times I_N}$ as follows.

$$\tilde{X}_T \doteq F(X_T) \doteq X_T \times_1 W_{mat}^{(I_1)} \cdots \times_k W_{mat}^{(I_n)} \cdots \times_N W_{mat}^{(I_N)} \in C \equiv \mathbb{C}^{I_1 \times \cdots \times I_N}$$
 (2) where $W_{mat}^{(I_n)} \in \mathbb{C}^{I_n \times I_n}$ denotes the $I_n \times I_n$ Fourier matrix for all $n \in [N]$.

The (m_1, m_2) -th complex entry of the matrix $W_{mat}^{(I_n)}$ is given by

$$\left(W_{mat}^{(I_n)}\right)_{m_1,m_2} = e^{2\pi i \cdot (m_1 - 1) \cdot (m_2 - 1) \cdot I_k^{-1}} \in \mathbb{C} , \quad \forall \ (m_1, m_2) \in [I_n] \times [I_n] . \tag{3}$$

The inverse multi-way transform $F^{-1}: \tilde{X}_T \mapsto X_T$ is given by the following N-mode multiplication for tensors as follows.

$$X_T \doteq F^{-1}(\tilde{X}_T) = \tilde{X}_T \times_1 \left(W_{mat}^{(I_1)}\right)^{-1} \cdots \times_n \left(W_{mat}^{(I_n)}\right)^{-1} \cdots \times_N \left(W_{mat}^{(I_N)}\right)^{-1} \in C \equiv \mathbb{C}^{I_1 \times \cdots \times I_N}$$
 (4) where $\left(W_{mat}^{(I_n)}\right)^{-1}$ denotes the inverse of the matrix $W_{mat}^{(I_n)}$ for all $n \in [N]$.

It is possible to employ a different multi-way transform and inverse transform to define a variant t-algebra C. Nevertheless, we only discuss the proposed algebra based on multi-way Fourier transforms or equivalently multi-way circular convolution. Interested readers are referred to [LM20a, LM20b] to connect the multi-way circular convolution and Fourier transform.

2.2. FOURIER MATRIX: fourier_matrix. Function fourier_matrix gives the $n \times n$ Fourier matrix.

FUNCTION 4. fourier_matrix.m: function returning the $n \times n$ Fourier matrix

```
function fourier_matrix_result = fourier_matrix(n)
% this function computes the nxn Fourier matrix
assert(isscalar(n));
assert(n > 0);
fourier_matrix_result = fft(eye(n), [], 1);
end
```

SYNTAX

```
Y = fourier_matrix(n)
```

DESCRIPTION

Input n is a positive integer. Output Y is a $n \times n$ complex matrix (i.e., Fourier matrix), whose (m_1, m_2) -th entry is given by equation (3).

EXAMPLES

The 3×3 fourier matrix is as follows.

The 2×2 fourier matrix is as follows.

Remark 2.3 (Shape of a Fourier matrix). A Fourier matrix must be square.

2.3. INVERSE FOURIER MATRIX: ifourier_matrix. The inverse Fourier transform is given by the following code.

FUNCTION 5. **ifourier_matrix**: function returning the $n \times n$ inverse Fourier matrix

```
function ifourier_matrix_result = ifourier_matrix(n)
% this function computes the n*n inverse Fourier matrix
ifourier_matrix_result = conj(fourier_matrix(n)) / n;
end
```

SYNTAX

```
Y = ifourier_matrix(n)
```

DESCRIPTION

Input n is a positive integer. Output Y is a $n \times n$ complex matrix (i.e., inverse Fourier matrix). The output of ifourier_matrix(n) is the inverse matrix of the output of fourier_matrix(n).

Let the $n \times n$ Fourier matrix be $X_{mat} \doteq W_{mat}^{(I_n)}$ and its inverse matrix be $Y_{mat} \doteq (W_{mat}^{(I_n)})^{-1}$. Besides the equality $X_{mat} \cdot Y_{mat} = Y_{mat} \cdot X_{mat} = I_{mat}$, the following equality

holds

$$(Y_{mat})_{m_1,m_2} = (1/n) \cdot \overline{(X_{mat})_{m_1,m_2}}$$

$$= (1/n) \cdot e^{-2\pi i \cdot (m_1 - 1) \cdot (m_2 - 1) \cdot I_k^{-1}} , \quad \forall (m_1, m_2) \in [I_n] \times [I_n] .$$
(5)

Thus, Function 5 is an efficient way of computing an inverse Fourier matrix.

EXAMPLES

The 3×3 inverse fourier matrix is given as follows.

```
>> ifourier_matrix(3)
ans =
    0.3333 + 0.0000i    0.3333 + 0.0000i    0.3333 + 0.0000i
    0.3333 + 0.0000i    -0.1667 + 0.2887i    -0.1667 - 0.2887i
    0.3333 + 0.0000i    -0.1667 - 0.2887i    -0.1667 + 0.2887i
```

2.4. IDENTITY T-SCALAR: E_T. The function returns the identity t-scalar of a given size.

SYNTAX

```
E = E_T(tsize)
```

DESCRIPTION

Input tsize is a row array of positive integers. Output is an array whose inceptional entry is equal to 1, and the other entries are equal to 0. The identity t-scalar generalizes the canonical identity scalar 1.

EXAMPLES

The 3×3 identity t-scalar is given as follows.

Compute the multi-way Fourer transform of the 3×3 identity t-scalar

The 2×3 identity t-scalar E_T is given as follows.

Compute the multi-way Fourer transform of the 2×3 identity t-scalar.

2.5. Zero T-scalar: Z_T. This function returns the zero t-scalar of a given size.

SYNTAX

$$Y = Z_T(tsize)$$

DESCRIPTION

Input tsize is a row array of positive integers. Output is an array of zeros. When tsize is a single positive integer, output is a square array of zeros.

EXAMPLES

The 3×3 zero t-scalar is given as follows.

When the input argument is a single postive integer n, the output is a $n \times n$ array of zeros.

The 1×3 zero t-scalar is given as follows.

The $2 \times 3 \times 2$ zero t-scalar is given as follows.

$$>> Z_T([2, 3, 2])$$

2.6. PSEUDO-INVERSE OF A T-SCLAAR: tpinv_tscalar. This function returns the pseudo-inverse of a t-scalar.

SYNTAX

```
Y = tpinv_tscalar(tscalar)
```

DESCRIPTION

Input tscalar is a numeric array. Output Y is a numeric array of the same size. This function generalizes the reciprocal of non-zero complex number.

EXAMPLES

Compute the pseudo-inverse of a complex number.

```
>> tpinv_tscalar(5)
ans =
          0.2000
>> tpinv_tscalar(i)
ans =
          0.0000 - 1.0000i
>> tpinv_tscalar(0)
ans =
          0
```

The pseudo-inverse of the 3×3 identity t-scalar E_T is itself.

The pseudo-inverse of the 3×3 zero t-scalar Z_T is still Z_T .

```
>> tpinv_tscalar(Z_T(3))
ans =
```

Compute the pseudo-inverse of a random t-scalar.

Mathematically, the following equality holds

$$E_T^+ \equiv E_T^{-1} \equiv E_T$$

$$Z_T^+ \equiv Z_T . \tag{6}$$

where $(\cdot)^+$ denotes the psuedo-inverse of a t-scalar, and $(\cdot)^{-1}$ denotes the inverse of a t-scalar.

The zero t-scalar Z_T is non-invertible and the identity t-scalar E_T is invertible. Any t-scalar is pseudo-invertible [LM20a, LM20b].

2.7. NORM AND ANGLE OF A T-SCALAR: tnorm_angle. Function tnorm_angle returns the norm (i.e., generalized absolute value, a nonnegative t-scalar) and the generalized angle (a nonnegative t-scalar) of the t-scalar.

SYNTAX

```
[tnorm, tangle] = tnorm_angle(tscalar)
```

DESCRIPTION

Input tscalar and outputs tnorm, tangle are all numeric arrays of the same size.

```
>> [tnorm, tangle] = tnorm_angle(tscalar)
tnorm =
    4.5066   -0.9511   -0.9511
    -0.5076   -0.0028    0.9449
    -0.5076    0.9449   -0.0028

tangle =
    0.0000 + 0.0000i    0.0000 + 0.0686i    0.0000 - 0.0686i
    0.0000 - 0.7277i    0.0000 + 0.4253i    0.0000 - 0.6812i
    0.0000 + 0.7277i    0.0000 + 0.6812i    0.0000 - 0.4253i
```

The multi-way Fourier transform of the norm of a t-scalar is an array of nonnegative real numbers. See the following example.

The multi-way Fourier transform of the abgle of a t-scalar is an array of real numbers. See the following example.

2.8. CONJUGATE OF A T-SCALAR: tconj_tscalar. Function tconj_tscalar returns the conjugate of a given t-scalar.

SYNTAX

```
conj_tscalar = tconj_tscalar(tscalar)
```

DESCRIPTION

Input tscalar and output conj_tscalar are numeric arrays of the same size.

```
>> X = randn(3, 2) + i * randn(3, 2)

X = 0.0774 + 0.3714i -0.0068 - 1.0891i
```

```
-1.2141 - 0.2256i
                     1.5326 + 0.0326i
  -1.1135 + 1.1174i -0.7697 + 0.5525i
>> Y = tconj_tscalar(X)
Y =
  0.0774 - 0.3714i - 0.0068 + 1.0891i
 -1.1135 - 1.1174i -0.7697 - 0.5525i
  -1.2141 + 0.2256i 1.5326 - 0.0326i
>> [fftn(X), fftn(Y)]
ans =
 Columns 1 through 2
  -1.4941 + 0.7592i -3.0064 + 1.7671i
  -0.7605 - 3.3628i 0.9168 + 3.3881i
  2.4662 + 0.4506i 2.3422 - 0.7739i
 Columns 3 through 4
  -1.4941 - 0.7592i -3.0064 - 1.7671i
  -0.7605 + 3.3628i \quad 0.9168 - 3.3881i
   2.4662 - 0.4506i 2.3422 + 0.7739i
```

Remark 2.4. From the above example, it shows fftn(X) is the conjugate array of fftn(Y).

In the following example, A is a random t-scalar, and C is a so-called nonnegative t-scalar.

```
46.7224 - 0.0000i 18.6277 - 0.0000i 6.3444 + 0.0000i 30.8452 - 0.0000i 44.0261 - 0.0000i 9.2967 - 0.0000i
```

Remark 2.5. The above example shows the Fourier transform of a nonnegative t-scalar is an array of nonnegative real numbers.

2.9. REAL PART OF A T-SCALAR: treal_part_tscalar. For all t-scalar $X_T \in C$, the t-scalar in the form of $\frac{X_T + X_T^*}{2}$ is called the real part of the t-scalar X_T .

It is easy to follow that the following equality holds for all $X_T \in C$,

$$\left(\frac{X_T + X_T^*}{2}\right)^* \equiv \frac{X_T + X_T^*}{2} \tag{7}$$

The real part of a t-scalar is self-conjugate. Namely, $\frac{X_T + X_T^*}{2} \in C^{sc}$, $\forall X_T \in C$, where C^{sc} denotes the subalgebra of C [LM20a, LM20b].

SYNTAX

```
real_part = treal_part_tscalar(tscalar)
```

DESCRIPTION

Input tscalar and output real_part are both numeric arrays of the same size, where tscalar can be any t-scalar in C and real_part is the so-called real part of tscalar and a self-conjugate t-scalar in C^{sc} .

```
>> A = randn(3)
A =
   1.1006 - 1.4916
                    2.3505
   1.5442 - 0.7423
                      -0.6156
   0.0859 - 1.0616
                    0.7481
>> B = treal_part_tscalar(A)
B =
   1.1006
            0.4294
                       0.4294
            0.0029
   0.8151
                      -0.8386
   0.8151 - 0.8386
                       0.0029
>> C = tconj_tscalar(B)
   1.1006
             0.4294
                      0.4294
   0.8151
            0.0029
                      -0.8386
   0.8151
          -0.8386
                    0.0029
>> fftn(B)
```

Remark 2.6. It follows from the above examples that C is identical to B, namely, B is self-conjugate. The Fourier transform of B, (i.e., a self-conjugate t-scalar) is an array of real numbers.

2.10. IMAGINARY PART OF A T-SCALAR: timg_part_tscalar. For all t-scalar $X_T \in C$, the t-scalar in the form of $\frac{X_T - X_T^*}{2i}$ is called the imaginary part of the t-scalar X_T .

It is easy to follow that the following equality holds for all $X_T \in C$,

$$\left(\frac{X_T - X_T^*}{2i}\right)^* \equiv \frac{X_T - X_T^*}{2i} \tag{8}$$

The imaginary part of a t-scalar is self-conjugate. Namely, $\frac{X_T - X_T^*}{2i} \in C^{sc}$, $\forall X_T \in C$, where C^{sc} denotes the subalgebra of C [LM20a, LM20b].

SYNTAX

imag_part = timg_part_tscalar(tscalar)

DESCRIPTION

Input tscalar and output imag_part are both numeric arrays of the same size, where tscalar can be any t-scalar in C and imag_part is the so-called imaginary part of tscalar and a self-conjugate t-scalar in C^{sc} .

```
0.4136 + 0.6949i  -0.5529 - 0.9312i  -0.8728 + 0.3097i

0.4136 - 0.6949i  -0.8728 - 0.3097i  -0.5529 + 0.9312i

>> fftn(B)

ans =

-0.6694  -1.8103    1.5172

2.4944  -0.7035   -2.8644

2.2401  -3.2276  -7.3080
```

Remark 2.7. Since C is identical to B, B is self-conjugate, and the Fourier transform of B (i.e., a self-conjugate t-scalar) is an array of real numbers.

2.11. POWER OF A T-SCALAR: tpower. Function tpower returns the p power of a t-scalar, $Y_T = (X_T)^p$. When $p = \frac{1}{2}$, the function returns the arithmetic square root of a t-scalar $Y_T = \sqrt{X_T}$.

SYNTAX

```
tscalar_power = tpower(tscalar, p_value)
```

DESCRIPTION

Input tscalar and outputs tscalar_power are numeric arrays of the same size. Input p_value is a canonical scalar, i.e., a complex number. Usually, input p_value is a real number.

```
>> A = randn(3) + i * randn(3)
  1.7119 - 1.9609i -0.8396 + 2.9080i 0.9610 - 1.0582i
 -0.1941 - 0.1977i 1.3546 + 0.8252i 0.1240 - 0.4686i
 -2.1384 - 1.2078i -1.0722 + 1.3790i 1.4367 - 0.2725i
>> B = tpower(A, 2)
B =
  9.3472 + 4.9465i
                    5.5966 + 2.8883i -16.1201 - 9.0661i
  6.2115 +10.0242i 18.5029 - 8.5377i -16.8333 - 0.8297i
                    14.1335 - 6.0189i -8.1149 -10.5458i
-10.9200 + 16.9954i
>> fftn(A)
ans =
                   4.3826 - 2.3565i -7.5884 - 7.6893i
  1.3440 - 0.0535i
  2.3033 - 2.7884i
                     3.8327 - 1.7446i 5.1228 - 4.2926i
  1.8525 + 2.5087i
                     7.0427 + 0.1217i - 2.8853 - 1.3535i
>> [fftn(A) .^2, fftn(B)]
```

```
ans =
  1.0e+02 *
 Columns 1 through 3
  0.0180 - 0.0014i
                   0.1365 - 0.2066i -0.0154 + 1.1670i
 -0.0247 - 0.1285i
                   -0.0286 + 0.0929i
                   0.4959 + 0.0171i
                                   0.0649 + 0.0781i
 Columns 4 through 6
  0.0180 - 0.0014i
                   0.1365 - 0.2066i
                                   -0.0154 + 1.1670i
 -0.0247 - 0.1285i
                   0.1165 - 0.1337i
                                   0.0782 - 0.4398i
                                   0.0649 + 0.0781i
 -0.0286 + 0.0929i
                   0.4959 + 0.0171i
```

Remark 2.8. The above example shows that the array fftn(A).^2 is identical to the array fftn(B).

More algebraically, for all t-scalar $X_T \in C$, let $Y_T \doteq X_T^p$ and $F(X_T) = \tilde{X}_T$ and $F(Y_T) = \tilde{Y}_T$. The following equality holds

$$\left((\tilde{X}_T)_{m_1, m_2} \right)^p = (\tilde{Y}_T)_{m_1, m_2} , \forall m_1, m_2 . \tag{9}$$

When p = 1/2, one has the arithmetical square root of a t-scalar, which helps define the norm of a t-scalar or a t-matrix.

See the following examples.

```
>> C = tpower(A, 0.5)
C =
   1.7072 - 0.7392i - 0.5633 + 0.6266i 0.3416 + 0.1000i
   0.0274 - 0.0826i
                     0.2279 + 0.1667i
                                        0.0453 - 0.0479i
 -0.2045 - 0.3783i - 0.5681 + 0.0536i
                                        0.1459 + 0.2779i
>> [fftn(A) .^0.5, fftn(C)]
ans =
 Columns 1 through 3
   1.1595 - 0.0231i
                                      1.2678 - 3.0325i
                     2.1632 - 0.5447i
                     2.0055 - 0.4350i 2.4296 - 0.8834i
   1.7205 - 0.8104i
                                        0.3884 - 1.7425i
   1.5766 + 0.7956i
                     2.6539 + 0.0229i
 Columns 4 through 6
   1.1595 - 0.0231i
                     2.1632 - 0.5447i
                                      1.2678 - 3.0325i
                                        2.4296 - 0.8834i
   1.7205 - 0.8104i
                     2.0055 - 0.4350i
   1.5766 + 0.7956i 2.6539 + 0.0229i 0.3884 - 1.7425i
```

2.12. IDEMPOTENT T-SCALAR: idempotent. This function returns an idempotent t-scalar.

SYNTAX

idempotent_tscalar = idempotent(k, tsize)

DESCRIPTION

Input tsize is a row array of positive integers, representing $I_1 \times \cdots \times I_N$. Input k in an integer in $\{1, \dots, K\}$ where $K \doteq I_1 \times \cdots \times I_N$.

Output tresult_scalar is a numeric array, i.e., the k-th idempotent t-scalar where $k = 1, \dots, (I_1 \times \dots \times I_N)$.

EXAMPLES

Compute the 9 idempoent t-scalars in $C \equiv \mathbb{C}^{3\times 3}$.

```
>> tsize = [3, 3];
>> idempotent(1, tsize)
ans =
   0.1111
             0.1111
                       0.1111
   0.1111
             0.1111
                       0.1111
   0.1111
             0.1111
                       0.1111
>> idempotent(2, tsize)
ans =
  0.1111 + 0.0000i 0.1111 + 0.0000i 0.1111 + 0.0000i
 -0.0556 + 0.0962i -0.0556 + 0.0962i -0.0556 + 0.0962i
 -0.0556 - 0.0962i -0.0556 - 0.0962i -0.0556 - 0.0962i
>> idempotent(3, tsize)
ans =
  0.1111 + 0.0000i 0.1111 + 0.0000i 0.1111 + 0.0000i
 -0.0556 - 0.0962i -0.0556 - 0.0962i -0.0556 - 0.0962i
 -0.0556 + 0.0962i -0.0556 + 0.0962i -0.0556 + 0.0962i
>> idempotent(4, tsize)
ans =
   0.1111 + 0.0000i -0.0556 + 0.0962i -0.0556 - 0.0962i
   0.1111 + 0.0000i -0.0556 + 0.0962i -0.0556 - 0.0962i
   0.1111 + 0.0000i -0.0556 + 0.0962i -0.0556 - 0.0962i
>> idempotent(5, tsize)
ans =
```

```
0.1111 + 0.0000i -0.0556 + 0.0962i -0.0556 - 0.0962i
 -0.0556 + 0.0962i -0.0556 - 0.0962i 0.1111 + 0.0000i
 >> idempotent(6, tsize)
ans =
  0.1111 + 0.0000i -0.0556 + 0.0962i -0.0556 - 0.0962i
 -0.0556 + 0.0962i -0.0556 - 0.0962i 0.1111 + 0.0000i
>> idempotent(7, tsize)
ans =
  0.1111 + 0.0000i -0.0556 - 0.0962i -0.0556 + 0.0962i
  0.1111 + 0.0000i -0.0556 - 0.0962i -0.0556 + 0.0962i
  0.1111 + 0.0000i -0.0556 - 0.0962i -0.0556 + 0.0962i
>> idempotent(8, tsize)
ans =
  0.1111 + 0.0000i -0.0556 - 0.0962i -0.0556 + 0.0962i
 -0.0556 - 0.0962i -0.0556 + 0.0962i 0.1111 + 0.0000i
>> idempotent(9, tsize)
ans =
  0.1111 + 0.0000i -0.0556 - 0.0962i -0.0556 + 0.0962i
 -0.0556 - 0.0962i -0.0556 + 0.0962i 0.1111 + 0.0000i
 -0.0556 + 0.0962i 0.1111 + 0.0000i -0.0556 - 0.0962i
```

2.13. RANDOM T-SCALAR: randn_tscalar. For the convenience of demonstration, we give function randn_tscalar, which returns a random t-scalar.

SYNTAX

```
random_tscalar = randn_tscalar(tsize, mode)
```

DESCRIPTION

Input tsize is a row numeric array. Input mode is either 'real' or 'complex'. If only one input tsize is given, the second input is set to the default value 'complex'.

```
>> randn_tscalar([2, 3])
```

```
ans =
    1.0984 - 1.5771i    0.7015 + 0.2820i    -0.3538 - 1.3337i
    -0.2779 + 0.5080i    -2.0518 + 0.0335i    -0.8236 + 1.1275i
>> randn_tscalar(3, 'real')
ans =
    -0.7145    -0.5890    -1.1201
    1.3514    -0.2938     2.5260
    -0.2248    -0.8479     1.6555
```

The following examples show that, for all t-scalar $X_T \in C$, the following equality holds [LM20a, LM20b]

$$X_T \circ X_T^* \equiv Re(X_T)^2 + Im(X_T)^2 \in S^{nonneg} \tag{10}$$

The following example shows that, for all $X_T \in C$, the t-scalar $X_T \circ X_T^*$ is nonnegative. In other words, $F(X_T \circ X_T^*)$ returns an array of nonnegative real numbers.

```
>> fftn(tproduct(X, tconj_tscalar(X)))
ans =
   22.6249 + 0.0000i   10.4956 - 0.0000i
   2.3111 + 0.0000i   14.1038 + 0.0000i
```

2.14. Generalized Exponential: texp. Generalized exponential of a t-scalar.

SYNTAX

```
exp_result = texp(tscalar)
```

DESCRIPTION

Input tscalar and output exp_result are numeric arrays of the same size.

EXAMPLES

Calculate the exponential of 1, which is Euler's number, e.

```
>> texp(1)
ans =
2.7183
```

Calculate the exponential of 0, which is equal to 1.

```
>> texp(0) ans = 1
```

Calculate the exponential of i, which is equal to $e^i \equiv \cos 1 + i \sin 1$.

```
>> texp(i)
ans =
0.5403 + 0.8415i
```

Calculate the exponential of E_T of the size 3×3 , i.e. $e^{E_T}=e\cdot E_T$, which generalizes Euler's number e.

Calculate the exponential of Z_T of the size 3×3 , i.e. $e^{Z_T} = E_T$, which generalizes the canonical scalar 1.

Calculate the exponential of a random t-scalar of the size 2×3 .

```
>> X = randn_tscalar([2, 3])
X =
     0.3335 + 0.8620i     0.4517 + 0.4550i     0.1837 - 0.3349i
     0.3914 - 1.3617i     -0.1303 - 0.8487i     -0.4762 + 0.5528i
>> texp(X)
ans =
     0.2069 + 0.4511i     -0.6209 - 0.7939i     -0.1891 + 0.4643i
```

Remark 2.9 (Expotential of a t-scalar). For all $X_T \in C$, the expotential of X_T is a t-scalar given by the following series

$$e^{X_T} \doteq \sum_{k=1}^{+\infty} \frac{X_T^{(k-1)}}{k-1} = E_T + X_T + \frac{X_T^2}{2} + \frac{X_T^3}{6} + \frac{X_T^4}{24} + \dots \in C$$
 (11)

For all $X_T \in C$, let $F(X_T) = \tilde{X}_T$ and $F(e^{X_T}) = \tilde{Y}_T$. By the semisimplicity of C introduced in [LM20a], the following equality holds in the following domain

$$e^{\alpha} = \beta \in \mathbb{C} \tag{12}$$

where $\alpha \doteq (\tilde{X}_T)_{m_1,m_2} \in \mathbb{C}$ and $\beta \doteq (\tilde{Y}_T)_{m_1,m_2}$ for all m_1, m_2 .

2.15. NATURAL LOGARITHM OF A T-SCALAR: tlog. Function tlog generalizes the natural logarithm of a canonical scalar and is the inverse function of function texp.

Function tlog is the MATLAB implementation of the map $\ln : X_T \mapsto \ln X_T \in C$.

The following equalities hold

$$\ln X_T = \sum_{k=1}^{+\infty} \frac{2}{2k-1} \left(\frac{X_T - E_T}{X_T + E_T} \right)^{2k-1}, \ \forall X_T \in S^{nonneg}$$

$$\ln X_T = \sum_{k=1}^{+\infty} (-1)^{k+1} \frac{(X_T - E_T)^k}{k}, \ \forall X_T \text{ satisfying } |X_T - E_T| \le E_T.$$
(13)

whehre $|\cdot| \in S^{nonneg}$ denotes the norm of a t-scalar.

SYNTAX

log_result = tlog(tscalar)

DESCRIPTION

Input tscalar and output log_result are also numeric arrays of the same size.

EXAMPLES

Calculate the natural logarithm of 1, which is equal to 0.

Calculate the natural logarithm of e, which is equal to 1.

Compute the natural logarithm of the 3×3 identity t-scalar E_T , which is the 3×3 zero t-scalar Z_T .

Function tlog is the inverse function of texp.

```
>> X = randn_tscalar(3)
X =
   0.5377 + 2.7694i
                     0.8622 + 0.7254i
                                      -0.4336 - 0.2050i
   1.8339 - 1.3499i
                     0.3188 - 0.0631i
                                        0.3426 - 0.1241i
 -2.2588 + 3.0349i -1.3077 + 0.7147i
                                      3.5784 + 1.4897i
>> Y = tloq(texp(X))
   0.5377 - 0.0231i 0.8622 + 0.0273i
                                       -0.4336 - 0.9031i
 -0.5845 - 2.0480i
                    1.5280 - 0.7612i
                                       -0.2620 + 0.2249i
   0.1596 + 2.3368i - 0.7031 + 1.0638i
                                       2.3692 + 0.7916i
```

2.16. ABSOLUTE VALUE OF A T-SCALAR: tabs_tscalar. Function tabs_tscalar returns the abosolute value of a t-scalar and is an efficient implementation of function tnorm_angle retuning only the first output.

SYNTAX

```
Y = tabs_tscalar(X)
```

DESCRIPTION

Function tabs_tscalar returns the norm (i.e., generalized absolute value, a non-negative t-scalar) of a t-scalar. It generalizes the absolute value of a complex number.

EXAMPLES

Compute the absolute value of 3 + 4i, which is equal to 5.

Compute the absolute value of 1+i, which is equal to $\sqrt{2}$.

```
>> tabs_tscalar(1 + i)
ans =
```

1.4142

Compute the absolute value of $3 \cdot E_T + 4i \cdot E_T$, which is equal to $5 \cdot E_T$.

Compute the absolute value of $(1+i) \cdot E_T$, which is equal to $\sqrt{2} \cdot E_T$.

Compute the absolute value of a random t-scalar, which is a nonnegative t-scalar.

```
>> X = randn tscalar(3)
X =
   1.4090 - 0.3034i - 1.2075 + 0.8884i 0.4889 - 0.8095i
   1.4172 + 0.2939i
                     0.7172 - 1.1471i
                                         1.0347 - 2.9443i
   0.6715 - 0.7873i 1.6302 - 1.0689i 0.7269 + 1.4384i
>> Y = tabs_tscalar(X)
Y =
   4.4986 + 0.0000i
                      0.3981 + 0.7577i
                                       0.3981 - 0.7577i
 -0.1722 + 0.3195i
                     0.2519 + 0.3154i
                                         1.3705 + 0.3892i
 -0.1722 - 0.3195i 1.3705 - 0.3892i
                                         0.2519 - 0.3154i
>> fftn(Y)
ans =
   8.1951
              3.3182
                        0.9491
   5.6186
              8.0814
                        1.9729
   2.0707
              4.8391
                        5.4425
```

2.17. ANGLE OF A T-SCALAR: tangle_tscalar. Function tabs_tscalar returns the abosolute value of a t-scalar and is an efficient implementation of function tnorm_angle retuning only the second output.

SYNTAX

```
tangle = tangle_tscalar(tscalar)
```

DESCRIPTION

Input tscalar and outputs tangleare also numeric arrays of the same size. Function tangle_tscalar returns a nonnegative t-scalar which generalizes the phrase angle of a complex number.

EXAMPLES

Compute the phrase angle of complex number 1+i, which is equal to $\frac{\pi}{4}$.

```
>> tangle_tscalar(1 + i)
ans =
     0.7854
```

The phrase angle of the t-scalar $(1+i) \cdot E_T$ is equal to $\frac{\pi}{4} \cdot E_T$.

Compate the phrase angle of a random t-scalar in C, which is a self-conjuate in C^{cs} .

```
>> X = randn tscalar(3)
X =
  0.3252 - 0.0301i - 1.7115 + 1.0933i 0.3192 + 0.0774i
 -0.7549 - 0.1649i -0.1022 + 1.1093i 0.3129 - 1.2141i
  1.3703 + 0.6277i -0.2414 - 0.8637i -0.8649 - 1.1135i
>> Y = tangle_tscalar(X)
Y =
  0.2518 + 0.0000i 0.0937 + 0.2899i 0.0937 - 0.2899i
 -0.9435 + 0.4960i -0.2052 - 0.1029i -0.4711 - 0.6768i
 -0.9435 - 0.4960i -0.4711 + 0.6768i -0.2052 + 0.1029i
>> fftn(Y)
ans =
            0.4435
                    -2.5487
  -2.8003
            1.9041
   1.5677
                    2.6915
   2.5503
            -0.3668
                      -1.1751
```

2.18. WHETHER INPUT IS A SELF-CONJUGATE: is_self_conjugate. This function determines whether input is a self-conjugate t-scalar.

SYNTAX

```
bool_result = is_self_conjugate(tscalar)
```

DESCRIPTION

Input tscalar is a numeric array and output bool_result is boolean, and true if tscalar is self-conjugate, otherwsie, false.

```
The number 1 is self-conjugate.
```

```
>> is_self_conjugate(1)
ans =
  logical
   1
  Any real number is self-conjugate.
>> is_self_conjugate(0.5)
ans =
  logical
   1
>> is_self_conjugate(pi)
ans =
  logical
   1
 Any element in \mathbb{C} \setminus \mathbb{R} is not self-conjugate
>> is_self_conjugate(i)
ans =
  logical
   0
>> is_self_conjugate(1 + i)
ans =
  logical
   0
  A random t-scalar is not self-conjugate.
>> is_self_conjugate(randn_tscalar(3))
ans =
  logical
   0
  The real part of a random t-scalar is self-conjugate.
>> is_self_conjugate(treal_part_tscalar(randn_tscalar(3)) )
ans =
  logical
```

1

The imaginary part of a random t-scalar is self-conjugate.

```
>> is_self_conjugate(timg_part_tscalar(randn_tscalar(3)) )
ans =
  logical
  1
```

The absolute value of a random t-scalar is self-conjugate.

```
>> is_self_conjugate(tabs_tscalar(randn_tscalar(3)) )
ans =
  logical
  1
```

2.19. WHETHER INPUT IS A NONNEGATIVE: is_nonnegative. This function determines whether input is a nonnegative t-scalar.

SYNTAX

```
bool_result = is_nonnegative(tscalar)
```

DESCRIPTION

The input tscalar is a numeric array and the output bool_result is boolean, and true if tscalar is a nonnegative t-scalar, otherwsie, false.

```
>> is_nonnegative(1)
ans =
  logical
   1
>> is_nonnegative(0)
ans =
  logical
   1
>> is_nonnegative(-1)
ans =
  logical
   0
>> is_nonnegative(E_T([2, 3, 4]))
ans =
  logical
   1
```

```
>> is_nonnegative(Z_T([3, 5, 7, 9]))
ans =
  logical
   1
 The t-scalar X_T = -1 \cdot E_T is not nonnegative, but self-conjugate
>> is_nonnegative(-1 * E_T([5, 11, 23]))
ans =
  logical
   0
>> is_self_conjugate(-1 * E_T([5, 11, 23]))
ans =
  logical
   1
 The absolute value of any t-scalar is nonnegative.
>> is_nonnegative(tabs_tscalar(randn_tscalar([3, 4, 5, 9]))
ans =
  logical
   1
```

2.20. Infinum of A set of two Self-conjugate T-scalars: tinf. This function returns the infinum of a set of two self-conjugate t-scalars.

SYNTAX

```
infimum_tscalar = tinf(tscalar1, tscalar2)
```

DESCRIPTION

Inputs tscalar1 and tscalar2 and output infimum_tscalar are numeric arrays of the same size. Function tinf generalizes $\min(\alpha, \beta)$ of two real numbers α and β .

```
The infimum of \{-1,3\} is -1.

>> tinf(-1, 3)

ans =

-1

The infimum of \{Z_T, E_T\} is Z_T.

>> tinf(Z_T(3), E_T(3))

ans =
```

```
0 0 0
0 0 0
0 0 0
```

Compute the infinum of a set of two random self-conjugate t-scalars.

```
>> X = treal_part_tscalar(randn_tscalar(3))
X =
  -0.3523 + 0.0000i -0.2677 - 0.5828i -0.2677 + 0.5828i
  1.4030 - 0.9531i 0.6374 + 0.8028i -0.4042 + 0.4081i
  1.4030 + 0.9531i -0.4042 - 0.4081i 0.6374 - 0.8028i
>> Y = treal_part_tscalar(randn_tscalar(3))
Y =
 -1.5246 + 0.0000i -0.4078 + 0.8015i -0.4078 - 0.8015i
  0.4044 + 0.7344i 0.9958 - 0.0609i 0.1535 + 0.1239i
  0.4044 - 0.7344i 0.1535 - 0.1239i 0.9958 + 0.0609i
>> tinf(X, Y)
ans =
 -2.4893 + 0.0000i 0.0746 + 0.1859i 0.0746 - 0.1859i
  0.8868 - 0.1011i 1.2877 + 0.6646i -0.6208 + 0.2339i
  0.8868 + 0.1011i - 0.6208 - 0.2339i 1.2877 - 0.6646i
>> fftn(X)
ans =
            2.1624 2.8137
   2.3847
  -2.0774
            -6.9840
                      -1.1567
  -2.9705
            1.5397 1.1174
>> fftn(Y)
ans =
   0.7672 \quad -0.3891 \quad -2.5254
  -2.5127
            0.5556 - 0.0140
            0.6478
  -5.2749
                    -4.9756
>> fftn(tinf(X, Y))
ans =
   0.7672 \quad -0.3891 \quad -2.5254
  -2.5127 -6.9840 -1.1567
  -5.2749 0.6478 -4.9756
```

3. Functions for T-matrices

3.1. MULTIIPLICATION OF TWO T-MATRICES: tmultiplication. This function generalizes the canonical multiplication of two complex matrices.

Let C be the set of all $I_1 \times \cdots \times I_N$ complex arrays, namely $C \equiv \mathbb{C}^{I_1 \times \cdots \times I_N}$. For all t-matrices $A_{TM} \in C^{M_1 \times M} \equiv \mathbb{C}^{I_1 \times \cdots \times I_N \times M_1 \times M}$, and $B_{TM} \in C^{D \times D_2} \equiv \mathbb{C}^{I_1 \times \cdots \times I_N \times M \times M_2}$, their t-matrix multiplication $C_{TM} \doteq A_{TM} \circ B_{TM}$ is an element in $C^{M_1 \times M_2} \equiv \mathbb{C}^{I_1 \times \cdots \times I_N \times M_1 \times M_2}$ defined as follows.

Definition 3.1 (T-matrix multiplication). The (m_1, m_2) -th t-scalar entry $(C_{TM})_{m_1, m_2} \in C \equiv \mathbb{C}^{I_1 \times \cdots \times I_N}$ of C_{TM} is given by

$$(C_{TM})_{m_1,m_2} \doteq \sum_{m=1}^{M} (A_{TM})_{m_1,m} \circ (B_{TM})_{m,m_2} , \ \forall (m_1,m_2) \in [M_1] \times [M_2] .$$
 (14) where $(A_{TM})_{m_1,m} \circ (A_{TM})_{m,m_2}$ denotes the t-scalar product (i.e., N-way circular convo-

SYNTAX

tmatrix_result = tmultiplication(tmatrix1, tmatrix2, tsize)

DESCRIPTION

Inputs tmatrix1 and tmatrix2 are numeric arrays, and input tsize is a row array of positive integers, representing the size $I_1 \times \cdots \times I_N$. Output tmatrix_result is a numeric array.

This function generalizes the multiplication of two complex matrices.

lution) of two $I_1 \times \cdots \times I_N$ complex arrays (see Defintion 2.1).

EXAMPLES

Compute the multiplication of two (canonical) complex matrices A and B, where A is a 5×3 complex matrix and B is a 3×2 complex matrix.

```
>> A = randn(5, 3) + i * randn(5, 3)

A =

-0.8095 + 1.0933i    1.3703 - 1.1135i    0.3129 - 0.2256i
-2.9443 + 1.1093i    -1.7115 - 0.0068i    -0.8649 + 1.1174i
1.4384 - 0.8637i    -0.1022 + 1.5326i    -0.0301 - 1.0891i
0.3252 + 0.0774i    -0.2414 - 0.7697i    -0.1649 + 0.0326i
-0.7549 - 1.2141i    0.3192 + 0.3714i    0.6277 + 0.5525i

>> B = randn(3, 2) + i * randn(3, 2)

B =

1.1006 + 2.3505i    -1.4916 - 0.1924i
1.5442 - 0.6156i    -0.7423 + 0.8886i
0.0859 + 0.7481i    -1.0616 - 0.7648i

>> A * B
```

```
ans =
  -1.8344 - 3.0478i
                     0.8854 + 0.5695i
  -9.4052 - 5.2075i
                      7.6544 - 3.1285i
   5.2108 + 4.7439i - 4.3987 + 0.9620i
  -0.7091 - 0.3109i \quad 0.5929 + 0.2704i
   >> squeeze(tmultiplication(reshape(A, [1, 5, 3]), ...
reshape(B, [1, 3, 2]), 1))
ans =
  -1.8344 - 3.0478i
                      0.8854 + 0.5695i
  -9.4052 - 5.2075i
                      7.6544 - 3.1285i
   5.2108 + 4.7439i - 4.3987 + 0.9620i
  -0.7091 - 0.3109i
                     0.5929 + 0.2704i
   Compute t-matrix multiplication when A is a 3\times3\times5\times7 real array, B is a 3\times3\times7\times11
real array, and tsize is equal to [3, 3]. The result t-matrix C is a 3 \times 3 \times 5 \times 11
real array.
>> A = randn([3, 3, 5, 7]); B = randn([3, 3, 7, 11]); ...
C = tmultiplication(A, B, [3, 3]); whos;
  Name
            Size
                                Bytes Class
                                                  Attributes
  Α
            3x3x5x7
                                  2520 double
            3x3x7x11
                                  5544 double
  В
  С
            3x3x5x11
                                  3960 double
>> tsize = [5, 7]; ...
row_num1 = 11; col_num1 = 13; row_num2 = 13; col_num2 = 17; ...
A = randn([tsize, row_num1, col_num1]) + ...
   i * randn([tsize, row_num1, col_num1]); ....
B = randn([tsize, row_num2, col_num2]) + ...
   i * randn([tsize, row_num2, col_num2]); ...
C = tmultiplication(A, B, tsize); ...
whos A; whos B; whos C;
  Name
            Size
                                  Bytes
                                         Class
                                                   Attributes
            5x7x11x13
  Α
                                  80080
                                         double
                                                   complex
  Name
            Size
                                  Bytes Class
                                                   Attributes
```

123760 double complex

5x7x13x17

Name	Size	Bytes	Class	Attributes
С	5x7x11x17	104720	double	complex

3.2. MULTIPLICATION OF THREE T-MATRICES: tmultiplication_arg3. This function computes the multiplication of three t-matrices.

SYNTAX

tmatrix_result = tmultiplication_arg3(tmatrix1, tmatrix2, tmatrix3,
tsize)

DESCRIPTION

Inputs tmatrix1, tmatrix2 and tmatrix3 are numeric arrays, i.e., three multiplicable t-matrices. Input tsize is a row array of positive integers, representing $I_1 \times \cdots \times I_N$.

Output tresult_scalar is a numeric array, i.e., a t-matrix which is the product of three input t-matrices.

EXAMPLES

Compute the multiplication of three t-matrices.

```
>> A = randn(3, 2); ...
B = randn(2, 2); \dots
C = randn(2, 5); \dots
tsize = [1, 1];
D = tmultiplication_arg3(reshape(A, [tsize, size(A)]), ...
reshape(B, [tsize, size(B)]), ...
reshape(C, [tsize, size(C)]), tsize );
>>
>> squeeze(D)
ans =
   1.2400
              0.8862
                      -1.5429
                                  0.8178
                                           1.8499
                       1.9570
  -0.0630
              1.6955
                                  0.2248
                                           -0.1399
    0.8769
              0.2671
                      -1.4791
                                 0.5233
                                           1.3176
>> A * B * C
ans =
   1.2400
              0.8862
                       -1.5429
                                  0.8178
                                            1.8499
              1.6955
                        1.9570
                                  0.2248
                                           -0.1399
  -0.0630
    0.8769
              0.2671
                       -1.4791
                                  0.5233
                                           1.3176
```

Compute the multiplication of three random t-matrices.

```
>> tsize = [3, 3];
>> A = randn([tsize, 3, 5]);
```

```
>> B = randn([tsize, 5, 7]);
>> C = randn([tsize, 7, 11]);
>> D = tmultiplication arg3(A, B, C, tsize);
>> whos A; whos B; whos C; whos D;
  Name
            Size
                         Bytes
                                 Class
                                           Attributes
            3x3x3x5
                          1080
                                 double
  Α
  Name
            Size
                         Bytes
                                 Class
                                           Attributes
            3x3x5x7
                          2520
                                 double
  Name
            Size
                         Bytes
                                 Class
                                           Attributes
            3x3x7x11
                          5544
  C
                                 double
  Name
            Size
                         Bytes
                                 Class
                                           Attributes
            3x3x3x11
                          2376
  D
                                 double
```

3.3. RANK OF A T-MATRIX: trank. This function computes the rank, a nonegative t-scalar, of a t-matrix and generalizes the rank of a complex matrix.

SYNTAX

```
generalized_rank = trank(tmatrix, tsize)
```

DESCRIPTION

Input tmatrix is a numeric array. input tsize is a row array of positive integers, representing the size $I_1 \times \cdots \times I_N$.

Output generalized_rank is a $I_1 \times \cdots \times I_N$ numerical array, namely, a nonnegative t-scalar.

EXAMPLES

Compute the rank of a random 3×3 t-scalar.

Compute the rank of a random rank-deficient t-scalar.

```
>> B = randn(3) + i * randn(3); B(3) = 0; B = ifftn(B)
B =
 -0.1815 - 0.1793i 0.1117 - 0.5259i
                                      -0.0164 + 0.1007i
  0.1309 - 0.8840i 0.2987 - 0.1716i
                                      0.0873 + 0.1710i
  0.1853 + 0.5892i -0.1343 - 0.2487i 0.4068 - 0.5628i
>> trank(B, [3, 3])
ans =
  0.8889 + 0.0000i -0.1111 + 0.0000i -0.1111 + 0.0000i
  0.0556 + 0.0962i 0.0556 + 0.0962i
                                      0.0556 + 0.0962i
  0.0556 - 0.0962i 0.0556 - 0.0962i 0.0556 - 0.0962i
>> fftn(trank(B, [3, 3]))
ans =
   1.0000
             1.0000
                       1.0000
   1.0000
             1.0000
                       1.0000
             1.0000
                       1.0000
```

Compute the rank a random 2×5 t-matrix with 3×3 t-scalar entries. The maximum rank a 2×5 t-matrix is $2 \cdot E_T$.

Compute the rank a random 2×5 t-matrix with 3×3 t-scalar entries and the first row t-vectors are set to zero.

```
>> B = randn(3, 3, 2, 5) + i * randn(3, 3, 2, 5); ...
B(:, :, 1, :) = 0;
>> trank(B, [3, 3])
ans =
    1     0     0
    0     0     0
    0     0     0
```

3.4. IDENTITY T-MATRIX: teye. This function returns the identity t-matrix of a given size.

SYNTAX

```
t_identity_matrix = teye(row_col_num, tsize)
```

DESCRIPTION

Input row_col_num is a nonnegative t-scalar. Input tsize is a row array of positive integers, representing $I_1 \times \cdots \times I_N$. Output t_identity_matrix is a numeric array, a diagonal t-matrix. This function generalizes the identity matrix.

EXAMPLES

Compute the 2×2 identity t-matrix, whose t-scalar entries are 3×3 numerical arrays.

$$A(:,:,2,2) = 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ \end{bmatrix}$$

3.5. CONJUGATE TRANSPOSE A T-MATRIX: tctranspose. This function computes the conjugate transpose of a t-matrix and generalizes the conjugate transpose of a complex matrix.

SYNTAX

tmatrix_result = tctranspose(tmatrix, tsize)

DESCRIPTION

Input tmatrix is a $I_1 \times \cdots \times I_N \times M_1 \times M_2$ numeric array, and input tsize is a row array of positive integers, representing $I_1 \times \cdots \times I_N$.

Output tmatrix_result is a $I_1 \times \cdots \times I_N \times M_2 \times M_1$ numeric array.

Given a $M_1 \times M_2$ t-matrix $A_{TM} \in C^{M_1 \times M_2} \equiv \mathbb{C}^{I_1 \times \cdots \times I_N \times M_1 \times M_2}$, function tetranspose returns a $M_2 \times M_1$ t-matrix $A_{TM}^* \in C^{M_1 \times M_2} \equiv \mathbb{C}^{I_1 \times \cdots \times M_2 \times M_1}$ such that the following condition holds

$$(A_{TM})_{m_1,m_2}^* = (A_{TM}^*)_{m_2,m_1} \in \mathbb{C}^{I_1 \times \dots \times I_N}, \ \forall (m_1, m_2) \in [M_1] \times [M_2].$$
 (15)

EXAMPLES

Compute the conjugate transpose of a 5×7 t-matrix, whose t-sclalar entries are 3×3 arrays.

```
>> A = randn(3, 3, 5, 7) + i * randn(3, 3, 5, 7); ...
B = tctranspose(A, [3, 3]);
>> whos
 Name
            Size
                               Bytes Class
                                                 Attributes
  Α
            3x3x5x7
                                 5040
                                       double
                                                 complex
            3x3x7x5
  В
                                 5040
                                       double
                                                 complex
>> norm(A(:, :, 1, 3) - tconj_tscalar(B(:, :, 3, 1)), 'F')
ans =
   1.1233e-15
```

3.6. Inverse of a T-Matrix: tinv. This function computes the inverse of a t-matrix and generalizes the inverse of a complex matrix.

SYNTAX

```
tresult = tinv(tmatrix, tsize)
```

DESCRIPTION

Input tmatrix is numeric arrays, and input tsize is a row array of positive integers, representing $I_1 \times \cdots \times I_N$. Output tresult is a numeric array, i.e., a nonnegative t-scalar.

If the t-matrix tmatrix is non-invertible, output tresult is a NaN.

EXAMPLES

Compute a zero t-matrix.

```
>> tsize = [3, 3]; row_num = 5; col_num = 5;
>> A = zeros([tsize, row_num, col_num]);
>> tinv(A, tsize)
ans =
   NaN
```

Given a full-rank square t-matrix $A_T \in C^{M \times M} \equiv \mathbb{C}^{I_1 \times \cdots \times I_N \times M \times M}$, let the inverse t-matrix of A_T be $B_T \doteq A_T^{-1} \in C^{M \times M} \equiv \mathbb{C}^{I_1 \times \cdots \times I_N \times M \times M}$. Then,

$$A_T \circ B_T = B_T \circ A_T = \operatorname{diag}(\underbrace{E_T, \cdots, E_T}_{M \text{ copies}})$$
 (16)

Compute the inverse of a random t-matrix.

```
>> tsize = [3, 3]; row_num = 5; col_num = 5;
>> A = randn([tsize, row_num, col_num]);
>> B = tinv(A, tsize);
>> norm(reshape(tmultiplication(A, B, tsize) ...
- teye(row_num, tsize), 1, []) )
ans =
    9.5911e-15
```

Given a t-matrix, if its row number and column number is equal to 1, function tinv computes the rank a t-scalar.

```
>> tsize = [3, 3];
>> A = randn(tsize);
>> B = tinv(A, tsize);
>> tproduct(A, B)
ans =
    1.0000     0.0000     0.0000
    0.0000     -0.0000     0.0000
    0.0000     0.0000     -0.0000
```

The above example can be rewritten as follows.

3.7. PSEUDO-INVERSE OF A T-MATRIX: tpinv. This function computes the pseudo-inverse of a t-matrix and generalizes the function tinv.

SYNTAX

```
tresult = tpinv(tmatrix, tsize)
```

DESCRIPTION

Input tmatrix is a numeric array, and input tsize is a row array of positive integers, representing $I_1 \times \cdots \times I_N$. Output tmatrix_result is a numeric array.

This function generalizes the inverse of a complex matrix.

EXAMPLES

Compute the pseudo-inverse of a canonical complex matrix.

```
>> tsize = 1; row_num = 2; col_num = 3; ...
A = randn(row_num, col_num) + i * randn(row_num, col_num)
A =
  0.8998 - 0.2130i
                   1.0294 - 1.0431i
                                      1.0128 - 0.4381i
 -0.3001 - 0.8657i - 0.3451 - 0.2701i 0.6293 - 0.4087i
>> B = tpinv(reshape(A, [tsize, row_num, col_num]), tsize);
>> B = reshape(B, col_num, row_num)
B =
  0.0627 - 0.1945i - 0.4116 + 0.5594i
  0.3892 + 0.2522i - 0.1208 - 0.3298i
  >> A * B
ans =
  1.0000 + 0.0000i 0.0000 - 0.0000i
 -0.0000 - 0.0000i
                    1.0000 + 0.0000i
 Compute the pseudo-inverse of a random t-matrix.
>> tsize = [3, 3]; row_num = 2; col_num = 3;
>> A = randn([tsize, row_num, col_num]) ...
+ i * randn([tsize, row_num, col_num])
A(:,:,1,1) =
  0.9835 - 0.0029i - 0.5316 + 1.4049i 0.1766 - 0.7777i
 -0.2977 + 0.9199i \quad 0.9726 + 1.0341i \quad 0.9707 + 0.5667i
  1.1437 + 0.1498i -0.5223 + 0.2916i -0.4140 - 1.3826i
A(:,:,2,1) =
 -0.4383 + 0.2445i -0.4320 + 0.8797i 0.7059 + 0.2669i
  0.9510 + 0.2130i -0.3601 + 0.9239i -1.6045 + 0.4255i
A(:,:,1,2) =
  1.0289 - 1.3147i 1.7463 - 0.0436i -2.1935 + 0.0645i
  1.4580 - 0.4164i 0.1554 + 0.5824i -0.3334 + 0.6003i
```

```
0.0475 + 1.2247i - 1.2371 - 1.0065i 0.7135 - 1.3615i
A(:,:,2,2) =
  0.4136 - 0.1818i -1.6387 - 1.8963i 0.5197 - 0.9905i
 -0.5771 - 0.9395i -0.7601 - 2.1280i -0.0142 - 1.1730i
A(:,:,1,3) =
 -1.1555 - 1.7254i -0.6667 + 0.1102i 0.3984 + 0.0931i
 -0.0095 + 0.2882i 0.8641 + 0.7871i 0.8840 - 0.3782i
 A(:,:,2,3) =
  0.5509 - 0.0438i 0.4759 - 0.4302i -0.0479 + 0.3763i
  0.6830 + 0.9608i 1.4122 - 1.6273i 1.7013 - 0.2270i
  1.1706 + 1.7382i 0.0226 + 0.1663i -0.5097 - 1.1489i
>> B = tpinv(A, tsize)
B(:,:,1,1) =
  0.0033 - 0.0302i - 0.0271 + 0.0519i - 0.0066 - 0.0199i
 -0.0413 + 0.0218i 0.0510 + 0.0235i 0.0298 - 0.0601i
B(:,:,2,1) =
  0.0781 + 0.0278i -0.0235 - 0.0051i 0.0389 + 0.0553i
  0.0017 - 0.0328i - 0.0149 + 0.0183i - 0.0075 - 0.0017i
  0.0149 + 0.0142i 0.0198 - 0.0264i 0.0134 - 0.0411i
B(:,:,3,1) =
 -0.0462 + 0.0672i -0.0009 - 0.0330i -0.0244 + 0.0291i
 -0.0357 + 0.0438i \quad 0.0513 + 0.0686i \quad -0.0227 - 0.0023i
  0.0402 - 0.0297i - 0.0572 + 0.0064i 0.0421 - 0.0040i
B(:,:,1,2) =
 -0.0644 + 0.0163i 0.0391 + 0.0651i -0.0222 - 0.0799i
  0.0526 + 0.0593i - 0.0431 - 0.0449i - 0.0190 + 0.0167i
  B(:,:,2,2) =
 -0.0152 - 0.0432i 0.0067 + 0.0262i 0.0173 - 0.0415i
  0.0167 + 0.0230i -0.0255 - 0.0035i -0.0283 + 0.0600i
```

```
0.0130 - 0.0222i - 0.0130 + 0.0038i - 0.0140 + 0.0617i
B(:,:,3,2) =
  0.0165 - 0.0425i 0.0354 + 0.0081i -0.0227 + 0.0563i
  0.0639 - 0.0559i - 0.0584 + 0.0370i 0.0061 - 0.0496i
 -0.0555 + 0.0312i \quad 0.0217 - 0.0412i \quad 0.0253 + 0.0218i
>> C = tmultiplication(A, B, tsize)
C(:,:,1,1) =
  1.0000 - 0.0000i 0.0000 + 0.0000i 0.0000 - 0.0000i
 -0.0000 + 0.0000i -0.0000 - 0.0000i 0.0000 + 0.0000i
 -0.0000 - 0.0000i -0.0000 - 0.0000i 0.0000 + 0.0000i
C(:,:,2,1) =
  1.0e-15 *
  0.0540 + 0.0669i - 0.0426 + 0.0245i - 0.1317 - 0.0636i
  0.0417 - 0.0060i -0.0233 + 0.1073i 0.0477 - 0.0826i
 -0.0865 + 0.0287i -0.0554 - 0.1601i -0.1926 - 0.0262i
C(:,:,1,2) =
  1.0e-15 *
 -0.0050 + 0.0482i -0.0031 - 0.0589i 0.1365 + 0.1842i
 -0.1083 - 0.0726i -0.0789 - 0.0489i 0.1417 - 0.0024i
 -0.1076 - 0.0126i 0.0449 - 0.0411i -0.0028 + 0.0459i
C(:,:,2,2) =
  1.0000 - 0.0000i 0.0000 - 0.0000i 0.0000 - 0.0000i
 -0.0000 - 0.0000i -0.0000 - 0.0000i -0.0000 + 0.0000i
 -0.0000 + 0.0000i 0.0000 - 0.0000i 0.0000 + 0.0000i
>> whos A; whos B; whos C;
 Name
           Size
                              Bytes Class
                                              Attributes
 Α
           3x3x2x3
                                864 double
                                               complex
 Name
           Size
                              Bytes Class
                                              Attributes
           3x3x3x2
                                864 double
                                               complex
 Name
           Size
                              Bytes Class
                                              Attributes
```

C 3x3x2x2

576 double complex

3.8. COMPACT TENSORIAL SINGULAR VECTOR DECOMPOSITION OF A T-MATRIX: tsvd. This function computes the compact SVD (Tensorial Singular Vector Decomposition) of a t-matrix. This function generalizes the SVD of a complex matrix.

SYNTAX

DESCRIPTION

Input tmatrix is a numeric array, and input tsize is a row array of positive integers, representing $I_1 \times \cdots \times I_N$. Outputs TU, TS, TV are all numeric arrays, i.e., three t-matrices.

Given a t-matrix $X_{TM} \in C^{M_1 \times M_2} \equiv \mathbb{C}^{I_1 \times \cdots \times I_N \times M_1 \times M_2}$, let $D \doteq \min(M_1, M_2)$. Then, the TSVD decompose the t-matrix X_{TM} to three t-matrices

$$U_{TM} \in C^{M_1 \times D} \equiv \mathbb{C}^{I_1 \times \dots \times I_N \times M_1 \times D}$$

$$V_{TM} \in C^{M_2 \times D} \equiv \mathbb{C}^{I_1 \times \dots \times I_N \times M_2 \times D}$$

$$S_{TM} \doteq \operatorname{diag}(\lambda_{T, 1}, \dots, \lambda_{T, D}) \in C^{D \times D} \equiv \mathbb{C}^{I_1 \times \dots \times I_N \times D \times D}$$

$$(17)$$

where $\lambda_{T,d} \in S^{nonneg}$ for all $d \in [D]$, and $\lambda_{T,d} \geq \lambda_{T,(d+1)}$ for all $d \in [D-1]$.

The binary relation \geq is the partial order generalizing the usual total order \geqslant of nonnegative real numbers.

Given two nonnegative $A_T, B_T \in S^{nonneg}, A_T \geq B_T$ if and only if $(A_T - B_T) \in S^{nonneg}$. Otherwise, A_T and B_T are incomparable.

The following t-matrix multiplication hold

$$X_{TM} = U_{TM} \circ S_{TM} \circ V_{TM}^* \tag{18}$$

where V_{TM}^* denotes the conjugate transpose (hemitian transpose) of the t-matrix V_{TM} [LM20a, LM20b].

EXAMPLES

FUNCTION 6. Example: TSVD of a random t-matrix

```
function tsvd_demo
2  % The following code demonstrate the TSVD of a random t-matrix
3  clear; close all; clc;
4
5  % The shape of t-scalars is set to $3*3*3$
6  tsize = [3, 3];
7  M1 = 5;
8  M2 = 3;
9
```

```
10 t_matrix = randn([tsize, M1, M2]);
  [U, S, V] = tsvd(t_matrix, tsize);
12
   whos t_matrix;
   whos U;
    whos S;
    whos V;
17
    tmatrix_product_result = tmultiplication(U, S, tsize);
    tmatrix_product_result = tmultiplication(tmatrix_product_result, ...
18
                             tctranspose(V, tsize), tsize);
19
    whos tmatrix_product_result;
20
21
    residual = norm(tmatrix_product_result(:) - t_matrix(:));
22
    fprintf('residual = %f \n', residual);
23
24 end
```

The result of the above script is as follows.

Name	Size	Bytes	Class	Attributes
t_matrix	3x3x5x3	1080	double	
Name	Size	Bytes	Class	Attributes
U	3x3x5x3	1080	double	
Name	Size	Bytes	Class	Attributes
S	3x3x3x3	648	double	
Name	Size	Bytes	Class	Attributes
V	3x3x3x3	648	double	
Name tmatrix_p	roduct_result	_	tes Class 80 double	Attributes

residual = 0.000000

3.9. Determinant of a t-matrix. This function generalizes the determinant of a complex matrix.

SYNTAX

```
tresult_scalar = tdet(tmatrix, tsize)
```

DESCRIPTION

Input tmatrix is a numeric array, and input tsize is a row array of positive integers, representing $I_1 \times \cdots \times I_N$. Output tresult_scalar is a numeric array, i.e., a t-scalar.

Input tmatrix must be a square t-matrix. This function generalizes the determinant of a complex matrix.

EXAMPLES

Compute the determinant of a canonical matrix.

```
>> row_col_num = 3; tsize = [1, 1];
>> A = randn(row_col_num)
A =
   -0.3034
              0.8884
                        -0.8095
    0.2939
             -1.1471
                       -2.9443
  -0.7873
             -1.0689
                        1.4384
>> B = tdet(reshape(A, [tsize, row_col_num, row_col_num]), tsize);
>> whos B
 Name
            Size
                             Bytes
                                    Class
                                               Attributes
                                    double
 В
            1x1
>> B
B =
    4.1247
>> C = det(A)
C =
    4.1247
 Compute the determinant of a random t-matrix.
>> row_col_num = 2; tsize = [3, 3];
>> A = randn([tsize, row_col_num, row_col_num])
```

```
A(:,:,1,1) =
  -1.3617
             -0.3349
                        -1.1176
                         1.2607
    0.4550
              0.5528
   -0.8487
              1.0391
                         0.6601
A(:,:,2,1) =
   -0.0679
             -0.3031
                         0.8261
   -0.1952
              0.0230
                         1.5270
   -0.2176
              0.0513
                         0.4669
```

3.10. DIAGONAL ELEMENTS OF A T-MATRIX: tdiag. This function gets diagonal elements of t-matrix.

SYNTAX

```
tresult = tdiag(tmatrix, tsize)
```

DESCRIPTION

Input tmatrix is a numeric array (i.e., a t-matrix), and input tsize is a row array of positive integers, representing $I_1 \times \cdots \times I_N$. Output tresult is a numeric array (i.e., a column t-vector).

EXAMPLES

Get the diagnal t-scalar entries of a random t-matrix in $C^{3\times 2}$.

```
>> row_num = 3; col_num = 2; tsize = [3, 3];
>> A = randn([tsize, row_num, col_num])
A(:,:,1,1) =
  -0.9047
             1.4790
                       0.3086
  -0.4677
            -0.8608
                      -0.2339
  -0.1249
            0.7847
                      -1.0570
A(:,:,2,1) =
             0.1922
  -0.2841
                      0.3362
   -0.0867 -0.8223
                    -0.9047
```

-1.4694	-0.0942	-0.2883			
-1.8359	2.4245 0.9594 -0.3158	-1.0360			
0.7873	- 0.3199 -0.5583 -0.3114	-1.0257			
-1.6989	-0.1178 0.6992 0.2696	-1.4831			
0.1097	-0.2900 1.2616 0.4754	0.1269			
-0.4677	g(A, tsize) 1.4790 -0.8608 0.7847	-0.2339			
-1.6989	-0.1178 0.6992 0.2696	-1.4831			
>> whos A Name A	Size 3x3x3x2		-	Class double	Attributes
>> whos B Name B	Size 3x3x2	В	_	Class double	Attributes

Get the diagnal t-scalar entries of the identity t-matrix in $C^{3\times 3}$.

>> tsize = [2, 2]; row_col_num = 3;

>> A = teye(row_col_num, tsize)

A(:,:,1,1) =

1 0

0 0

A(:,:,2,1) =

0 0

0 0

A(:,:,3,1) =

0

0 0

A(:,:,1,2) =

0 0

0 0

A(:,:,2,2) =

1 0

0 0

A(:,:,3,2) =

0

0 0

A(:,:,1,3) =

0 0

0 0

A(:,:,2,3) =

0 0

0 (

A(:,:,3,3) =

1 (

0 0

>> B = tdiag(A, tsize)

B(:,:,1) =

1 0

0 0

Get the diagnal elements of a canonical random matrix.

```
>> tsize = [1, 1]; row_num = 3; col_num = 5; ...
A = randn(row_nu, col_num)
A =
    1.8045
             -0.2603
                        -2.1860
                                   0.4018
                                              0.8123
   -0.7231
              0.6001
                        -1.3270
                                   1.4702
                                              0.5455
    0.5265
              0.5939
                        -1.4410
                                  -0.3268
                                             -1.0516
>> B = tdiag(reshape(A, [tsize, row_num, col_num]), tsize);
>> whos B
  Name
            Size
                              Bytes
                                      Class
                                                Attributes
            1x1x3
                                 24
                                      double
  В
>> reshape(B, 3, 1)
ans =
    1.8045
    0.6001
   -1.4410
```

3.11. Square Root of A T-matrix: tsqrtm. This function computes the square root of a t-matrix.

Given a t-matrix A_{TM} , this function yields a t-matrix B_{TM} such that the following equality holds

$$A_{TM} = B_{TM} \circ B_{TM} . (19)$$

SYNTAX

tmatrix_result = tsqrtm(tmatrix, tsize)

DESCRIPTION

Input tmatrix is a numeric array, i.e., a t-matrix. Input tsize is a row array of positive integers, representing $I_1 \times \cdots \times I_N$. Output tmatrix_result is a numeric array, which is the square root of the input t-matrix.

Function tsqrtm generates the MATLAB function sqrtm.

EXAMPLES

Compute the square root of an order-six t-matrix. The input t-matrix is a compound image version of the image "cameraman.tif". ¹

```
>> A = load('compound_img_layer002'); ...
A = struct2cell(A); A = A{1};
A = double(A);
assert(isnumeric(A));
tsize = [3, 3, 3, 3]; \dots
row_num = 256; col_num = 256;
assert(isequal(size(A), [tsize, row_num, col_num])); ...
B = tsqrtm(A, tsize); ...
residual = tmultiplication(B, B, tsize) - A;
>> norm(residual(:))
ans =
   2.3579e-08
 Compute the square root of the real matrix by the image "cameraman.tif".
>> A = imread('cameraman.tif'); ...
A = double(A); \dots
tsize = 1;
B = tsqrtm(reshape(A, [tsize, size(A)]), tsize); ...
B = squeeze(B); \dots
residual = A - B * B;
norm(residual(:))
ans =
   3.9978e-10
```

3.12. NORM OF A T-MATRIX: tnorm. This function computes the norm of a t-matrix. This function generalizes the norm of a complex matrix.

SYNTAX

```
generalized_norm = tnorm(tmatrix, tsize, norm_type, p_value)
```

DESCRIPTION

Input tmatrix is a numeric array, i.e., a t-matrix. Input tsize is a row array of positive integers, representing $I_1 \times \cdots \times I_N$.

¹ The compound image "compound_img_layer002" is located at "code_repository/demo/compound_img".

Input norm_type is one of the following 'fro' (generalized Frobenius norm), 'p' (generalized p norm), 'max' (generalized max norm), 'ell_one' (generalized ℓ_1 norm), 'spectral' (generalized spectral norm), 'nuclear' (generalized nuclear norm).

Output generalized_norm is a nonnegative t-scalar.

If function thorm is given two inputs tmatrix and tsize, input norm_type is set to the default value 'fro'.

Input p is a nonnegative real number, only valid when input norm_type is 'p'.

EXAMPLES

Compute norms of a real matrix recast by the "cameraman,tif" image.

```
>> A = imread('cameraman.tif');
A = double(A);
tsize = [1, 1];
tnorm(reshape(A, [tsize, size(A)]), tsize)
ans =
   3.4329e+04
>> norm(A(:))
ans =
   3.4329e+04
>> tnorm(reshape(A, [tsize, size(A)]), tsize, 'max')
ans =
     7780728
>> tnorm(reshape(A, [tsize, size(A)]), tsize, 'spectral')
ans =
   3.2036e+04
>> tnorm(reshape(A, [tsize, size(A)]), tsize, 'nuclear')
ans =
   1.1159e+05
>> tnorm(reshape(A, [tsize, size(A)]), tsize, 'p', 0.5)
ans =
   4.5173e+11
 Compute the norms of an order-four t-matrix.
>> A = load('compound img layer001');
```

```
A = struct2cell(A);
A = A\{1\};
tsize = [3, 3]; row num = 256; col num = 256;
assert(isnumeric(A));
assert(isequal(size(A), [tsize, row_num, col_num]));
>>
>> tnorm(A, tsize, 'fro')
ans =
  1.0e+04 *
   4.6639
            3.3189
                      3.3189
                    3.0177
    3.5154
            3.0165
   3.5154
            3.0177 3.0165
>> tnorm(A, tsize)
ans =
  1.0e+04 *
   4.6639 3.3189 3.3189
    3.5154
            3.0165
                      3.0177
   3.5154 3.0177 3.0165
>> tnorm(A, tsize, 'max')
ans =
  1.0e + 06 *
   9.2494 7.6526 7.6526
             7.2950
                      7.2950
    7.7857
   7.7857 7.2950
                      7.2950
>> tnorm(A, tsize, 'spectral')
ans =
  1.0e+04 *
   3.70263.21143.21143.26122.98142.9953
   3.2612
            2.9953
                      2.9814
>> tnorm(A, tsize, 'nuclear')
ans =
  1.0e+05 *
```

```
0.7733
    1.8944
                         0.7733
    0.9075
               0.5990
                         0.5986
    0.9075
               0.5986
                         0.5990
>> tnorm(A, tsize, 'p', 0.3)
ans =
   1.0e+18 *
    1.2435
               1.1184
                         1.1184
    1.1153
               1.0903
                         1.0902
    1.1153
               1.0902
                         1.0903
```

3.13. EIGENVALUES AND EIGENVECTORS OF A T-MATRIX: teig. This function computes the eigenvalues and eigenvectors of a t-matrix. This function generalizes the eigenvalues and eigenvectors of a complex matrix.

SYNTAX

```
[TV, TD] = teig(tmatrix, tsize)
```

DESCRIPTION

Input tmatrix is a numeric array, a t-matrix Input tsize is a row array of positive integers, representing $I_1 \times \cdots \times I_N$.

Outputs TV TD are numeric arrays, i.e., two t-matrices, whose multiplucation is equal to tmatrix. The t-matrix TD is a diagonal t-matrix.

Given t-matrix X_{TM} , the t-matrix V_{TM} contains column eigenvectors of the t-matrix X_{TM} and the diagonal t-matrix contains the eigenvalues of the t-matrix X_{TM} . Namely,

$$X_{TM} \circ V_{TM} = V_{TM} \circ D_{TM} . \tag{20}$$

EXAMPLES

Compute the eigenvectors and eigvalues of an order-six matrix.²

```
>> A = load('compound_img_layer002'); ...
A = struct2cell(A); A = A{1}; ...
tsize = [3, 3, 3, 3]; row_num = 256; col_num = 256; ...
assert(isequal(size(A), [tsize, row_num, col_num]));
>> [TV, TD] = teig(A, tsize);
>> residual = tmultiplication(A, TV, tsize) - ...
```

 $^{^2}$ The mat data file "compound_img_layer002" is located in "code_repository/demo/compound_img".

```
tmultiplication(TV, TD, tsize); ...
>> norm(residual(:))
ans =
   2.4285e-09
```

3.14. TRACE OF A T-MATRIX: ttrace. This function computes the sum of the diagnal t-scalar entries of a t-matrix. This function generalizes the trace of a complex matrix.

SYNTAX

```
tresult = ttrace(tmatrix, tsize)
```

DESCRIPTION

Input tmatrix is a numeric array, i.e., a square t-matrix. Input tsize is a row array of positive integers, representing the shape $I_1 \times \cdots \times I_N$ of t-scalars. Input tmatrix must be a square t-matrix. Output tresult_scalar is a $I_1 \times \cdots \times I_N$ numeric array, i.e., a t-scalar.

This function generalizes the trace of a complex matrix.

EXAMPLES

>> A = randn(3)

Compute the trace of a square matrix.

Compute the trace of a square t-matrix.

```
>> tsize = [3, 3];
>> A = randn([tsize, 2, 2])
A(:,:,1,1) =
  -0.8637
            -1.1135
                      -0.7697
   0.0774
            -0.0068
                       0.3714
  -1.2141
            1.5326
                      -0.2256
A(:,:,2,1) =
             0.5525
    1.1174
                     0.0859
```

```
-1.0891
              1.1006
                        -1.4916
    0.0326
              1.5442
                        -0.7423
A(:,:,1,2) =
   -1.0616
              0.7481
                        -0.7648
    2.3505
             -0.1924
                        -1.4023
              0.8886
                        -1.4224
   -0.6156
A(:,:,2,2) =
    0.4882
              1.4193
                         1.5877
   -0.1774
              0.2916
                        -0.8045
   -0.1961
              0.1978
                         0.6966
>> ttrace(A, tsize)
ans =
   -0.3755
              0.3058
                         0.8180
   -0.1000
              0.2847
                        -0.4331
              1.7304
   -1.4102
                         0.4710
>> A(:,:,1,1) + A(:,:,2,2)
ans =
   -0.3755
                         0.8180
              0.3058
   -0.1000
              0.2847
                        -0.4331
   -1.4102
              1.7304
                         0.4710
```

3.15. NAIVE TRANSPOSE OF A T-MATRIX: tsimtranspose. This function computes the permute the rows and columns of a t-matrix. This function generalizes the permuation of a canonical matrix.

SYNTAX

```
tmatrix_result = tsimtranspose(tmatrix, tsize)
```

DESCRIPTION

Input tmatrix is a numeric array, i.e., a t-matrix. Input tsize is a row array of positive integers, representing the shape $I_1 \times \cdots \times I_N$ of t-scalars.

Output tmatrix_result is a numeric array, i.e., a t-matrix which permutes the rows and columns of the t-matrix tmatrix.

EXAMPLES

Permute the rows and columns of a canonical matrix.

>> A = randn(3, 5)
$$A =$$

```
0.8351
           -1.1658
                      0.7223
                               0.1873
                                       -0.4390
  -0.2437 \quad -1.1480
                      2.5855 -0.0825 -1.7947
   0.2157
            0.1049 \quad -0.6669 \quad -1.9330
                                         0.8404
>> tsize = [1, 1];
>> B = tsimtranspose(reshape(A, [tsize, size(A)]), tsize);
>> squeeze(B)
ans =
   0.8351 - 0.2437 0.2157
  -1.1658 -1.1480
                      0.1049
   0.7223
            2.5855 -0.6669
   0.1873 -0.0825
                    -1.9330
  -0.4390 \quad -1.7947
                      0.8404
```

Permute the rows and columns of a t-matrix in $C^{2\times 3} \equiv \mathbb{C}^{3\times 3\times 2\times 3}$.

```
>> tsize = [3, 3]; row_num = 2; col_num = 3;
>> A = randn([tsize, row_num, col_num])
A(:,:,1,1) =
  -0.8880
            0.3035
                      0.7394
   0.1001 - 0.6003
                       1.7119
  -0.5445
             0.4900 - 0.1941
A(:,:,2,1) =
  -2.1384 \quad -1.0722 \quad 1.4367
   -0.8396
            0.9610 -1.9609
    1.3546
            0.1240 - 0.1977
A(:,:,1,2) =
  -1.2078
            1.3790 -0.2725
    2.9080 -1.0582
                      1.0984
    0.8252 - 0.4686 - 0.2779
A(:,:,2,2) =
   0.7015 - 0.8236
                      0.2820
  -2.0518 \quad -1.5771
                       0.0335
  -0.3538
             0.5080
                      -1.3337
A(:,:,1,3) =
    1.1275
            0.0229 - 0.2857
   0.3502 \quad -0.2620 \quad -0.8314
  -0.2991 \quad -1.7502 \quad -0.9792
```

-0.5336	0.9642 0.5201 -0.0200 ntranspose(A	-0.7982 1.0187		
0.1001	0.3035 -0.6003 0.4900	1.7119		
2.9080	1.3790 -1.0582 -0.4686	1.0984		
0.3502	= 0.0229 -0.2620 -1.7502	-0.8314		
-0.8396	-1.0722	-1.9609		
-2.0518	-0.8236 -1.5771 0.5080	0.0335		
B(:,:,3,2) -1.1564 -0.5336 -2.0026	0.9642 0.5201	-0.0348 -0.7982 1.0187		
<pre>>> whos A; Name A >> whos B;</pre>	Size 3x3x2x3	Bytes 432	Class double	Attributes
Name B	Size 3x3x3x2	Bytes 432	Class double	Attributes

3.16. Generalized Signular Value Thresholding of a t-matrix: tsvt. This function computes the general singular value thresholding of a t-matrix. This function generalizes the SVT (Singular Value Threasholding) of a canonical matrix.

SYNTAX

```
approximation_tmatrix = tsvt(tmatrix, tau, tsize)
```

DESCRIPTION

Input tmatrix is a numeric array, i.e., a t-matrix. Input tsize is a row array of positive integers, representing $I_1 \times \cdots \times I_N$.

Input tau is a nonnegative t-scalar or nonnegative real numbers. When tau is a nonnegative real number τ , it is equivalent that tau is set to the t-scalar $\tau \cdot E_T$.

Output tresult_scalar is a numeric array, i.e., a low-rank approximation of tmatrix.

EXAMPLES

FUNCTION 7. Example: TSVT of the generalized "cameraman" image (i.e., a $3 \times 3 \times 256 \times 256$ compound image)

```
1 function tsvt_simple_demo
    clear; close all; clc;
3
    tsize = [3, 3];
4
    row_num = 256;
5
6
    col num = 256;
    A = load('compound_img_layer001');
    A = struct2cell(A);
    A = A\{1\};
    A = double(A);
11
    assert(isequal(size(A), [tsize, row_num, col_num]));
12
13
14
15
    tau = 2000 * E_T(tsize);
    approximation_A = tsvt(A, tau, tsize);
16
17
    PSNR_OF_Approximation = PSNR(A, approximation_A);
18
19
    fprintf('PSNR_OF_Approximation = %f \n', PSNR_OF_Approximation);
20
21
    A_slice = squeeze(A(1, 1, :, :));
    approximation_A = squeeze(approximation_A(1, 1, :, :));
22
    imshow([A_slice, approximation_A], []);
23
24 end
```

The output of the above script is as follows.

 $PSNR_OF_Approximation = 24.774847$



3.17. PRODUCT OF THE T-SCALAR ENTRIES OF A T-MATRIX: tprod. This function computes the circular convolution product of the t-scalar entries of a t-matrix.

SYNTAX

```
tresult = tprod(tmatrix, tsize)
```

DESCRIPTION

Input tmatrix is a numeric array, a t-matrix. I nput tsize is a row array of positive integers, representing $I_1 \times \cdots \times I_N$. Output tresult_scalar is a numeric array, i.e., the product of the t-scalar entries of the t-matrix tmatrix.

EXAMPLES

Compute the product of the scalar entries of a canonical matrix.

```
-0.0892 \quad -0.6994 \quad 1.2261
```

Compute the product of the t-scalar entries of a matrix.

```
>> row_num = 2; col_num = 3;
tsize = [3, 3];
A = randn([tsize, row_num, col_num])
A(:,:,1,1) =
   -0.8095
              0.3252
                       -1.7115
   -2.9443 \quad -0.7549 \quad -0.1022
    1.4384
              1.3703
                       -0.2414
A(:,:,2,1) =
    0.3192
             -0.0301
                        1.0933
    0.3129 - 0.1649
                       1.1093
   -0.8649
             0.6277
                     -0.8637
A(:,:,1,2) =
    0.0774 - 0.0068
                       0.3714
   -1.2141
             1.5326
                       -0.2256
   -1.1135
             -0.7697
                        1.1174
A(:,:,2,2) =
   -1.0891
                     -1.4916
              1.1006
    0.0326
              1.5442
                       -0.7423
    0.5525
              0.0859
                       -1.0616
A(:,:,1,3) =
    2.3505
             -0.1924
                      -1.4023
             0.8886 -1.4224
   -0.6156
    0.7481
            -0.7648
                       0.4882
A(:,:,2,3) =
   -0.1774
              0.2916
                       -0.8045
   -0.1961
              0.1978
                       0.6966
    1.4193
              1.5877
                        0.8351
>> B = tprod(A, tsize)
B(:,:,1,1) =
    5.2927
            -2.9262
                        2.0104
   -3.9515 \quad -2.0013
                       -4.8892
```

$$-0.4323 \qquad 0.9548 \qquad 0.6643$$

$$B(:,:,1,2) = \\ 0.2780 \qquad -2.0533 \qquad 0.4430 \\ -0.3332 \qquad -4.5325 \qquad 5.4209 \\ 2.1468 \qquad -4.1028 \qquad 2.9800$$

$$B(:,:,1,3) = \\ -3.6439 \qquad 1.5816 \qquad -2.5461 \\ -0.6945 \qquad 0.8428 \qquad 2.9841 \\ 2.0323 \qquad 0.9619 \qquad -1.2185$$

4. Functions for Generalized Tensors

4.1. GENERALIZED TENSOR MULTIPLICATION: gtensor_multiplication. This function computes a mode k generalized tensor multiplication with a t-matrix.

SYNTAX

gtensor_product = gtensor_multiplication(tmatrix, gtensor, mod_k, tsize)

DESCRIPTION

Input tmatrix is a numeric array, a t-matrix in $C^{M\times I_k}\equiv\mathbb{C}^{I_1\times\cdots\times I_N\times M\times M_k}$. Input gtensor is a numeric array, i.e., a generalized tensor in $C^{M_1\times\cdots M_k\cdots\times M_K}\equiv C^{I_1\times\cdots\times I_N\times M_1\times\cdots M_k\cdots\times M_K}$. Input mod_k is a positive integer not larger than $M_1\cdots M_N$.

Input tsize is row array of positive integers, representing $I_1 \times \cdots \times I_N$.

EXAMPLES

To be updated.

5. Demos

5.1. Low-rank Approximation Of A Higher-order Matrix. With the t-matrix paradigm, many applications can be straightforwardly generalized. To this end, we discuss a high-order generalization of the Eckart-Young-Mirsky theorem, named after the authors of the theorem [EY36, Mir60].

Low-rank approximation plays an important role in modeling many applications in machine learning and data analytics. The problem is to find a low-rank optimal approximation to a given matrix. Specifically, given a matrix $X_{mat} \in \mathbb{C}^{M_1 \times M_2}$, one seeks an approximation matrix \hat{X}_{mat} to X_{TM} , satisfying

$$||X_{mat} - \hat{X}_{mat}||_F = \min_{\operatorname{rank}(Y_{mat}) \leq r} ||X_{mat} - Y_{mat}||_F$$
subject to $\operatorname{rank}(\hat{X}_{mat}) \leq r \leq \operatorname{rank}(X_{mat})$. (21)

With the canonical paradigm where a nonnegative integer defines rank, a solution to a higher-order generalization of equation (21) is NP-hard [HL13, VNVM14, BDHR17], and as a consequence, "naive approach to this problem is doomed to failure". [DSL08].

With the notion of rank defined on t-algebra rather than the field of complex numbers, the higher-order generalization of the Eckart-Young-Mirsky theorem is straightforward.

This straightforward higher-order generalization is "a cut of the Gordian knot" [Gor]. Interested readers are referred to Section 5.3 [LM20a] for more details.

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