

# Package ‘rTensor’

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**Type** Package

**Title** Tools for Tensor Analysis and Decomposition

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**Description** A set of tools for creation, manipulation, and modeling of tensors with arbitrary number of modes. A tensor in the context of data analysis is a multidimensional array. rTensor does this by providing a S4 class 'Tensor' that wraps around the base 'array' class. rTensor provides common tensor operations as methods, including matrix unfolding, summing/averaging across modes, calculating the Frobenius norm, and taking the inner product between two tensors. Familiar array operations are overloaded, such as index subsetting via '[' and element-wise operations. rTensor also implements various tensor decomposition, including CP, GLRAM, MPCA, PVD, and Tucker. For tensors with 3 modes, rTensor also implements transpose, t-product, and t-SVD, as defined in Kilmer et al. (2013). Some auxiliary functions include the Khatri-Rao product, Kronecker product, and the Hamadard product for a list of matrices.

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## Description

This package is centered around the [Tensor-class](#), which defines a S4 class for tensors of arbitrary number of modes. A vignette and/or a possible paper will be included in a future release of this package.

## Details

This page will summarize the full functionality of this package. Note that since all the methods associated with S4 class [Tensor-class](#) are documented there, we will not duplicate it here.

The remaining functions can be split into two groups: the first is a set of tensor decompositions, and the second is a set of helper functions that are useful in tensor manipulation.

rTensor implements the following tensor decompositions:

[cp](#) Canonical Polyadic (CP) decomposition

[tucker](#) General Tucker decomposition

[mpca](#) Multilinear Principal Component Analysis; note that for 3-Tensors this is also known as Generalized Low Rank Approximation of Matrices(GLRAM)

[hosvd](#) (Truncated-)Higher-order singular value decomposition

[t\\_svd](#) Tensor singular value decomposition; 3-Tensors only; also note that there is an associated reconstruction function [t\\_svd\\_reconstruct](#)

[pvd](#) Population value decomposition of images; 3-Tensors only

rTensor also provides a set functions for tensors multiplication:

[ttm](#) Tensor times matrix, aka m-mode product

[ttl](#) Tensor times list (of matrices)

[t\\_mult](#) Tensor product based on block circulant unfolding; only implemented for a pair of 3-Tensors

...as well as for matrices:

[hadamard\\_list](#) Computes the Hamadard (element-wise) product of a list of matrices

[kronecker\\_list](#) Computes the Kronecker product of a list of matrices

[khatri\\_rao](#) Computes the Khatri-Rao product of two matrices

[khatri\\_rao\\_list](#) Computes the Khatri-Rao product of a list of matrices

[fold](#) General folding of a matrix into a tensor

[k\\_fold](#) Inverse operation for [k\\_unfold](#)

[unmatvec](#) Inverse operation for [matvec](#)

For more information on any of the functions, please consult the individual man pages.

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as.tensor

*Tensor Conversion*

**Description**

Create a [Tensor-class](#) object from an array, matrix, or vector.

**Usage**

```
as.tensor(x, drop = FALSE)
```

**Arguments**

x	an instance of array, matrix, or vector
drop	whether or not modes of 1 should be dropped

**Value**

a [Tensor-class](#) object

**Examples**

```
#From vector
vec <- runif(100); vecT <- as.tensor(vec); vecT
#From matrix
mat <- matrix(runif(1000),nrow=100,ncol=10)
matT <- as.tensor(mat); matT
#From array
indices <- c(10,20,30,40)
arr <- array(runif(prod(indices)), dim = indices)
arrT <- as.tensor(arr); arrT
```

cp

*Canonical Polyadic Decomposition*

**Description**

Canonical Polyadic (CP) decomposition of a tensor, aka CANDECOMP/PARAFRAC. Approximate a  $K$ -Tensor using a sum of `num_components` rank-1  $K$ -Tensors. A rank-1  $K$ -Tensor can be written as an outer product of  $K$  vectors. There are a total of `num_components * tnsr@num_modes` vectors in the output, stored in `tnsr@num_modes` matrices, each with `num_components` columns. This is an iterative algorithm, with two possible stopping conditions: either relative error in Frobenius norm has gotten below `tol`, or the `max_iter` number of iterations has been reached. For more details on CP decomposition, consult Kolda and Bader (2009).

**Usage**

```
cp(tnsr, num_components = NULL, max_iter = 25, tol = 1e-05)
```

**Arguments**

tnsr	Tensor with K modes
num_components	the number of rank-1 K-Tensors to use in approximation
max_iter	maximum number of iterations if error stays above tol
tol	relative Frobenius norm error tolerance

**Details**

Uses the Alternating Least Squares (ALS) estimation procedure. A progress bar is included to help monitor operations on large tensors.

**Value**

a list containing the following

lambdas a vector of normalizing constants, one for each component

U a list of matrices - one for each mode - each matrix with num\_components columns

conv whether or not  $\text{resid} < \text{tol}$  by the last iteration

norm\_percent the percent of Frobenius norm explained by the approximation

est estimate of tnsr after compression

fnorm\_resid the Frobenius norm of the error  $\text{fnorm}(\text{est} - \text{tnsr})$

all\_resids vector containing the Frobenius norm of error for all the iterations

**References**

T. Kolda, B. Bader, "Tensor decomposition and applications". SIAM Applied Mathematics and Applications 2009.

**See Also**

[tucker](#)

**Examples**

```
subject <- faces_tnsr[, , 14, ]
cpD <- cp(subject, num_components=10)
cpD$conv
cpD$norm_percent
plot(cpD$all_resids)
```

---

cs_fold	<i>Column Space Folding of Matrix</i>
---------	---------------------------------------

---

**Description**

DEPRECATED. Please see [unmatvec](#)

**Usage**

```
cs_fold(mat, m = NULL, modes = NULL)
```

**Arguments**

mat	matrix to be folded
m	the mode corresponding to cs_unfold
modes	the original modes of the tensor

---

cs_unfold-methods	<i>Tensor Column Space Unfolding</i>
-------------------	--------------------------------------

---

**Description**

DEPRECATED. Please see [matvec-methods](#) and [unfold-methods](#).

**Usage**

```
cs_unfold(tnsr, m)

## S4 method for signature 'Tensor'
cs_unfold(tnsr, m = NULL)
```

**Arguments**

tnsr	Tensor instance
m	mode to be unfolded on

**Details**

```
cs_unfold(tnsr, m=NULL)
```

---

`dim-methods`*Mode Getter for Tensor*

---

**Description**

Return the vector of modes from a tensor

**Usage**

```
## S4 method for signature 'Tensor'  
dim(x)
```

**Arguments**

`x` the Tensor instance

**Details**

`dim(x)`

**Value**

an integer vector of the modes associated with `x`

**Examples**

```
tnsr <- rand_tensor()  
dim(tnsr)
```

---

`faces_tnsr`*ORL Database of Faces*

---

**Description**

A dataset containing pictures of 40 individuals under 10 different lightings. Each image has 92 x 112 pixels. Structured as a 4-tensor with modes 92 x 112 x 40 x 10.

**Usage**

```
faces_tnsr
```

**Format**

A Tensor object with modes 92 x 112 x 40 x 10. The first two modes correspond to the image pixels, the third mode corresponds to the individual, and the last mode corresponds to the lighting.

**Source**

<http://www.cl.cam.ac.uk/research/dtg/attarchive/facedatabase.html>

**See Also**

[plot\\_orl](#)

---

fnorm-methods

*Tensor Frobenius Norm*

---

**Description**

Returns the Frobenius norm of the Tensor instance.

**Usage**

```
fnorm(tnsr)

## S4 method for signature 'Tensor'
fnorm(tnsr)
```

**Arguments**

tnsr            the Tensor instance

**Details**

```
fnorm(tnsr)
```

**Value**

numeric Frobenius norm of x

**Examples**

```
tnsr <- rand_tensor()
fnorm(tnsr)
```



---

fold	<i>General Folding of Matrix</i>
------	----------------------------------

---

**Description**

General folding of a matrix into a Tensor. This is designed to be the inverse function to [unfold-methods](#), with the same ordering of the indices. This amounts to following: if we were to unfold a Tensor using a set of `row_idx` and `col_idx`, then we can fold the resulting matrix back into the original Tensor using the same `row_idx` and `col_idx`.

**Usage**

```
fold(mat, row_idx = NULL, col_idx = NULL, modes = NULL)
```

**Arguments**

<code>mat</code>	matrix to be folded into a Tensor
<code>row_idx</code>	the indices of the modes that are mapped onto the row space
<code>col_idx</code>	the indices of the modes that are mapped onto the column space
<code>modes</code>	the modes of the output Tensor

**Details**

This function uses `aperm` as the primary workhorse.

**Value**

Tensor object with modes given by `modes`

**References**

T. Kolda, B. Bader, "Tensor decomposition and applications". SIAM Applied Mathematics and Applications 2009.

**See Also**

[unfold-methods](#), [k\\_fold](#), [unmatvec](#)

**Examples**

```
tnsr <- new("Tensor", 3L, c(3L, 4L, 5L), data = runif(60))
matT3 <- unfold(tnsr, row_idx = 2, col_idx = c(3, 1))
identical(fold(matT3, row_idx = 2, col_idx = c(3, 1), modes = c(3, 4, 5)), tnsr)
```

---

hadamard_list	<i>List hadamard Product</i>
---------------	------------------------------

---

**Description**

Returns the hadamard (element-wise) product from a list of matrices or vectors. Commonly used for n-mode products and various Tensor decompositions.

**Usage**

```
hadamard_list(L)
```

**Arguments**

L                    list of matrices or vectors

**Value**

matrix that is the hadamard product

**Note**

The modes/dimensions of each element in the list must match.

**See Also**

[kronecker\\_list](#), [khatri\\_rao\\_list](#)

**Examples**

```
lizt <- list('mat1' = matrix(runif(40),ncol=4),  
'mat2' = matrix(runif(40),ncol=4),  
'mat3' = matrix(runif(40),ncol=4))  
dim(hadamard_list(lizt))
```

---

head-methods	<i>Head for Tensor</i>
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---

**Description**

Extend head for Tensor

**Usage**

```
## S4 method for signature 'Tensor'  
head(x, ...)
```

**Arguments**

`x` the Tensor instance  
`...` additional parameters to be passed into `head()`

**Details**

`head(x, ...)`

**See Also**

[tail-methods](#)

**Examples**

```
tnsr <- rand_tensor()
head(tnsr)
```

---

hosvd *(Truncated-)Higher-order SVD*

---

**Description**

Higher-order SVD of a K-Tensor. Write the K-Tensor as a (m-mode) product of a core Tensor (possibly smaller modes) and K orthogonal factor matrices. Truncations can be specified via ranks (making them smaller than the original modes of the K-Tensor will result in a truncation). For the mathematical details on HOSVD, consult Lathauwer et. al. (2000).

**Usage**

```
hosvd(tnsr, ranks = NULL)
```

**Arguments**

`tnsr` Tensor with K modes  
`ranks` a vector of desired modes in the output core tensor, default is `tnsr@modes`

**Details**

A progress bar is included to help monitor operations on large tensors.

**Value**

a list containing the following:

`Z` core tensor with modes specified by ranks

`U` a list of orthogonal matrices, one for each mode

`est` estimate of `tnsr` after compression

`fnorm_resid` the Frobenius norm of the error `fnorm(est-tnsr)` - if there was no truncation, then this is on the order of `mach_eps * fnorm`.

**Note**

The length of ranks must match `tnsr@num_modes`.

**References**

L. Lathauwer, B.Moor, J. Vanderwalle "A multilinear singular value decomposition". Journal of Matrix Analysis and Applications 2000.

**See Also**

[tucker](#)

**Examples**

```
tnsr <- rand_tensor(c(6,7,8))
hosvdD <-hosvd(tnsr)
hosvdD$fnorm_resid
hosvdD2 <-hosvd(tnsr,ranks=c(3,3,4))
hosvdD2$fnorm_resid
```

---

initialize-methods      *Initializes a Tensor instance*

---

**Description**

Not designed to be called by the user. Use as `.tensor` instead.

**Usage**

```
## S4 method for signature 'Tensor'
initialize(.Object, num_modes = NULL, modes = NULL,
  data = NULL)
```

**Arguments**

<code>.Object</code>	the tensor object
<code>num_modes</code>	number of modes of the tensor
<code>modes</code>	modes of the tensor
<code>data</code>	can be vector, matrix, or array

**See Also**

`as.tensor`

---

innerProd-methods	<i>Tensors Inner Product</i>
-------------------	------------------------------

---

**Description**

Returns the inner product between two Tensors

**Usage**

```
innerProd(tnsr1, tnsr2)

## S4 method for signature 'Tensor, Tensor'
innerProd(tnsr1, tnsr2)
```

**Arguments**

tnsr1	first Tensor instance
tnsr2	second Tensor instance

**Details**

```
innerProd(tnsr1, tnsr2)
```

**Value**

inner product between x1 and x2

**Examples**

```
tnsr1 <- rand_tensor()
tnsr2 <- rand_tensor()
innerProd(tnsr1, tnsr2)
```

---

khatri_rao	<i>Khatri-Rao Product</i>
------------	---------------------------

---

**Description**

Returns the Khatri-Rao (column-wise Kronecker) product of two matrices. If the inputs are vectors then this is the same as the Kronecker product.

**Usage**

```
khatri_rao(x, y)
```

**Arguments**

x            first matrix  
y            second matrix

**Value**

matrix that is the Khatri-Rao product

**Note**

The number of columns must match in the two inputs.

**See Also**

[kronecker](#), [khatri\\_rao\\_list](#)

**Examples**

```
dim(khatri_rao(matrix(runif(12),ncol=4),matrix(runif(12),ncol=4)))
```

---

khatri\_rao\_list            *List Khatri-Rao Product*

---

**Description**

Returns the Khatri-Rao product from a list of matrices or vectors. Commonly used for n-mode products and various Tensor decompositions.

**Usage**

```
khatri_rao_list(L, reverse = FALSE)
```

**Arguments**

L            list of matrices or vectors  
reverse      whether or not to reverse the order

**Value**

matrix that is the Khatri-Rao product

**Note**

The number of columns must match in every element of the input list.

**See Also**

[khatri\\_rao](#)

**Examples**

```
smalllizt <- list('mat1' = matrix(runif(12),ncol=4),
  'mat2' = matrix(runif(12),ncol=4),
  'mat3' = matrix(runif(12),ncol=4))
dim(khatri_rao_list(smalllizt))
```

---

kronecker_list	<i>List Kronecker Product</i>
----------------	-------------------------------

---

**Description**

Returns the Kronecker product from a list of matrices or vectors. Commonly used for n-mode products and various Tensor decompositions.

**Usage**

```
kronecker_list(L)
```

**Arguments**

L                    list of matrices or vectors

**Value**

matrix that is the Kronecker product

**See Also**

[hadamard\\_list](#), [khatri\\_rao\\_list](#), [kronecker](#)

**Examples**

```
smalllizt <- list('mat1' = matrix(runif(12),ncol=4),
  'mat2' = matrix(runif(12),ncol=4),
  'mat3' = matrix(runif(12),ncol=4))
dim(kronecker_list(smalllizt))
```

---

`k_fold`*k-mode Folding of Matrix*

---

**Description**

k-mode folding of a matrix into a Tensor. This is the inverse function to `k_unfold` in the `m` mode. In particular, `k_fold(k_unfold(tnsr, m), m, getModes(tnsr))` will result in the original Tensor.

**Usage**

```
k_fold(mat, m = NULL, modes = NULL)
```

**Arguments**

<code>mat</code>	matrix to be folded into a Tensor
<code>m</code>	the index of the mode that is mapped onto the row indices
<code>modes</code>	the modes of the output Tensor

**Details**

This is a wrapper function to [fold](#).

**Value**

Tensor object with modes given by `modes`

**References**

T. Kolda, B. Bader, "Tensor decomposition and applications". SIAM Applied Mathematics and Applications 2009.

**See Also**

[k\\_unfold-methods](#), [fold](#), [unmatvec](#)

**Examples**

```
tnsr <- new("Tensor", 3L, c(3L, 4L, 5L), data=runif(60))
matT2 <- k_unfold(tnsr, m=2)
identical(k_fold(matT2, m=2, modes=c(3, 4, 5)), tnsr)
```



---

k\_unfold-methods      *Tensor k-mode Unfolding*

---

### Description

Unfolding of a tensor by mapping the kth mode (specified through parameter *m*), and all other modes onto the column space. This is the most common type of unfolding operation for Tucker decompositions and its variants. Also known as k-mode matricization.

### Usage

```
k_unfold(tnsr, m)

## S4 method for signature 'Tensor'
k_unfold(tnsr, m = NULL)
```

### Arguments

<code>tnsr</code>	the Tensor instance
<code>m</code>	the index of the mode to unfold on

### Details

```
k_unfold(tnsr, m=NULL)
```

### Value

matrix with `x@modes[m]` rows and `prod(x@modes[-m])` columns

### References

T. Kolda and B. Bader, "Tensor decomposition and applications". SIAM Applied Mathematics and Applications 2009.

### See Also

[matvec-methods](#) and [unfold-methods](#)

### Examples

```
tnsr <- rand_tensor()
matT2<-rs_unfold(tnsr, m=2)
```

---

`matvec-methods`*Tensor Matvec Unfolding*

---

**Description**

For 3-tensors only. Stacks the slices along the third mode. This is the prevalent unfolding for T-SVD and T-MULT based on block circulant matrices.

**Usage**

```
matvec(tnsr)

## S4 method for signature 'Tensor'
matvec(tnsr)
```

**Arguments**

`tnsr`            the Tensor instance

**Details**

```
matvec(tnsr)
```

**Value**

matrix with  $\prod(x@modes[-m])$  rows and  $x@modes[m]$  columns

**References**

M. Kilmer, K. Braman, N. Hao, and R. Hoover, "Third-order tensors as operators on matrices: a theoretical and computational framework with applications in imaging". *SIAM Journal on Matrix Analysis and Applications* 2013.

**See Also**

[k\\_unfold-methods](#) and [unfold-methods](#)

**Examples**

```
tnsr <- rand_tensor(c(2,3,4))
matT1 <- matvec(tnsr)
```

---

modeMean-methods      *Tensor Mean Across Single Mode*

---

### Description

Given a mode for a K-tensor, this returns the K-1 tensor resulting from taking the mean across that particular mode.

### Usage

```
modeMean(tnsr, m, drop)

## S4 method for signature 'Tensor'
modeMean(tnsr, m = NULL, drop = FALSE)
```

### Arguments

tnsr	the Tensor instance
m	the index of the mode to average across
drop	whether or not mode m should be dropped

### Details

```
modeMean(tnsr, m=NULL, drop=FALSE)
```

### Value

K-1 or K Tensor, where  $K = x@num\_modes$

### See Also

[modeSum](#)

### Examples

```
tnsr <- rand_tensor()
modeMean(tnsr, 1, drop=TRUE)
```

---

modeSum-methods

*Tensor Sum Across Single Mode*

---

### Description

Given a mode for a K-tensor, this returns the K-1 tensor resulting from summing across that particular mode.

### Usage

```
modeSum(tnsr, m, drop)

## S4 method for signature 'Tensor'
modeSum(tnsr, m = NULL, drop = FALSE)
```

### Arguments

tnsr	the Tensor instance
m	the index of the mode to sum across
drop	whether or not mode m should be dropped

### Details

```
modeSum(tnsr, m=NULL, drop=FALSE)
```

### Value

K-1 or K tensor, where  $K = x@num\_modes$

### See Also

[modeMean](#)

### Examples

```
tnsr <- rand_tensor()
modeSum(tnsr, 3, drop=TRUE)
```

**Description**

This is basically the Tucker decomposition of a K-Tensor, [tucker](#), with one of the modes uncompressed. If  $K = 3$ , then this is also known as the Generalized Low Rank Approximation of Matrices (GLRAM). This implementation assumes that the last mode is the measurement mode and hence uncompressed. This is an iterative algorithm, with two possible stopping conditions: either relative error in Frobenius norm has gotten below `tol`, or the `max_iter` number of iterations has been reached. For more details on the MPCA of tensors, consult Lu et al. (2008).

**Usage**

```
mpca(tnsr, ranks = NULL, max_iter = 25, tol = 1e-05)
```

**Arguments**

<code>tnsr</code>	Tensor with K modes
<code>ranks</code>	a vector of the compressed modes of the output core Tensor, this has length K-1
<code>max_iter</code>	maximum number of iterations if error stays above <code>tol</code>
<code>tol</code>	relative Frobenius norm error tolerance

**Details**

Uses the Alternating Least Squares (ALS) estimation procedure. A progress bar is included to help monitor operations on large tensors.

**Value**

a list containing the following:

- `Z_ext` the extended core tensor, with the first K-1 modes given by `ranks`
- `U` a list of K-1 orthogonal factor matrices - one for each compressed mode, with the number of columns of the matrices given by `ranks`
- `conv` whether or not `resid < tol` by the last iteration
- `est` estimate of `tnsr` after compression
- `norm_percent` the percent of Frobenius norm explained by the approximation
- `fnorm_resid` the Frobenius norm of the error `fnorm(est-tnsr)`
- `all_resids` vector containing the Frobenius norm of error for all the iterations

**Note**

The length of `ranks` must match `tnsr@num_modes-1`.

**References**

H. Lu, K. Plataniotis, A. Venetsanopoulos, "Mpca: Multilinear principal component analysis of tensor objects". IEEE Trans. Neural networks, 2008.

**See Also**

[tucker](#), [hosvd](#)

**Examples**

```
subject <- faces_tnsr[, ,21,]
mpcaD <- mpca(subject,ranks=c(10,10))
mpcaD$conv
mpcaD$norm_percent
plot(mpcaD$all_resids)
```

---

Ops-methods

*Conformable elementwise operators for Tensor*

---

**Description**

Overloads elementwise operators for tensors, arrays, and vectors that are conformable (have the same modes).

**Usage**

```
## S4 method for signature 'Tensor, Tensor'
Ops(e1, e2)
```

**Arguments**

e1	left-hand object
e2	right-hand object

**Examples**

```
tnsr <- rand_tensor(c(3,4,5))
tnsr2 <- rand_tensor(c(3,4,5))
tnsrsum <- tnsr + tnsr2
tnsrdiff <- tnsr - tnsr2
tnsrelemprod <- tnsr * tnsr2
tnsrelemquot <- tnsr / tnsr2
for (i in 1:3L){
  for (j in 1:4L){
    for (k in 1:5L){
      stopifnot(tnsrsum@data[i,j,k]==tnsr@data[i,j,k]+tnsr2@data[i,j,k])
      stopifnot(tnsrdiff@data[i,j,k]==(tnsr@data[i,j,k]-tnsr2@data[i,j,k]))
      stopifnot(tnsrelemprod@data[i,j,k]==tnsr@data[i,j,k]*tnsr2@data[i,j,k])
    }
  }
}
```

```
stopifnot(tnsrelemquot@data[i,j,k]==tnsr@data[i,j,k]/tnsr2@data[i,j,k])
}
}
}
```

---

plot\_orl

*Function to plot the ORL Database of Faces*

---

### Description

A wrapper function to image() to allow easy visualization of faces\_tnsr, the ORL Face Dataset.

### Usage

```
plot_orl(subject = 1, condition = 1)
```

### Arguments

subject	which subject to plot (1-40)
condition	which lighting condition (1-10)

### References

AT&T Laboratories Cambridge. <http://www.cl.cam.ac.uk/research/dtg/attarchive/facedatabase.html>

F. Samaria, A. Harter, "Parameterisation of a Stochastic Model for Human Face Identification".  
IEEE Workshop on Applications of Computer Vision 1994.

### See Also

[faces\\_tnsr](#)

### Examples

```
plot_orl(subject=5,condition=4)
plot_orl(subject=2,condition=7)
```

---

print-methods      *Print for Tensor*

---

### Description

Extend print for Tensor

### Usage

```
## S4 method for signature 'Tensor'
print(x, ...)
```

### Arguments

x                    the Tensor instance  
 ...                  additional parameters to be passed into print()

### Details

```
print(x, ...)
```

### See Also

[show](#)

### Examples

```
tnsr <- rand_tensor()
print(tnsr)
```

---

pvd                    *Population Value Decomposition*

---

### Description

The default Population Value Decomposition (PVD) of a series of 2D images. Constructs population-level matrices P, V, and D to account for variances within as well as across the images. Structurally similar to Tucker ([tucker](#)) and GLRAM ([mpca](#)), but retains crucial differences. Requires  $2 \cdot n_3 + 2$  parameters to specified the final ranks of P, V, and D, where  $n_3$  is the third mode (how many images are in the set). Consult Crainiceanu et al. (2013) for the construction and rationale behind the PVD model.

### Usage

```
pvd(tnsr, uranks = NULL, wranks = NULL, a = NULL, b = NULL)
```



**Arguments**

tnsr	3-Tensor with the third mode being the measurement mode
uranks	ranks of the U matrices
wranks	ranks of the W matrices
a	rank of $P = U \times \times t(U)$
b	rank of $D = W \times \times t(W)$

**Details**

The PVD is not an iterative method, but instead relies on  $n_3 + 2$  separate PCA decompositions. The third mode is for how many images are in the set.

**Value**

a list containing the following:

P population-level matrix  $P = U \times \times t(U)$ , where U is constructed by stacking the truncated left eigenvectors of slicewise PCA along the third mode

V a list of image-level core matrices

D population-level matrix  $D = W \times \times t(W)$ , where W is constructed by stacking the truncated right eigenvectors of slicewise PCA along the third mode

est estimate of tnsr after compression

norm\_percent the percent of Frobenius norm explained by the approximation

fnorm\_resid the Frobenius norm of the error  $fnorm(est - tnsr)$

**References**

C. Crainiceanu, B. Caffo, S. Luo, V. Zippunikov, N. Punjabi, "Population value decomposition: a framework for the analysis of image populations". Journal of the American Statistical Association, 2013.

**Examples**

```
subject <- faces_tnsr[, , 8, ]
pvdD <- pvd(subject, uranks=rep(46, 10), wranks=rep(56, 10), a=46, b=56)
```

---

rand_tensor	<i>Tensor with Random Entries</i>
-------------	-----------------------------------

---

**Description**

Generate a Tensor with specified modes with iid normal(0,1) entries.

**Usage**

```
rand_tensor(modes = c(3, 4, 5), drop = FALSE)
```

**Arguments**

modes	the modes of the output Tensor
drop	whether or not modes equal to 1 should be dropped

**Value**

a Tensor object with modes given by modes

**Note**

Default rand\_tensor() generates a 3-Tensor with modes c(3, 4, 5).

**Examples**

```
rand_tensor()  
rand_tensor(c(4,4,4))  
rand_tensor(c(10,2,1), TRUE)
```

---

rs_fold	<i>Row Space Folding of Matrix</i>
---------	------------------------------------

---

**Description**

DEPRECATED. Please see [k\\_fold](#).

**Usage**

```
rs_fold(mat, m = NULL, modes = NULL)
```

**Arguments**

mat	matrix to be folded
m	the mode corresponding to rs_unfold
modes	the original modes of the tensor

---

rs\_unfold-methods      *Tensor Row Space Unfolding*

---

**Description**

DEPRECATED. Please see [k\\_unfold-methods](#) and [unfold-methods](#).

**Usage**

```
rs_unfold(tnsr, m)

## S4 method for signature 'Tensor'
rs_unfold(tnsr, m = NULL)
```

**Arguments**

tnsr	Tensor instance
m	mode to be unfolded on

**Details**

```
rs_unfold(tnsr,m=NULL)
```

---

show-methods      *Show for Tensor*

---

**Description**

Extend show for Tensor

**Usage**

```
## S4 method for signature 'Tensor'
show(object)
```

**Arguments**

object	the Tensor instance
--------	---------------------

**Details**

```
show(object)
```

**See Also**

[print](#)

**Examples**

```
tnsr <- rand_tensor()
tnsr
```

---

t-methods

*Tensor Transpose*

---

**Description**

Implements the tensor transpose based on block circulant matrices (Kilmer et al. 2013) for 3-tensors.

**Usage**

```
## S4 method for signature 'Tensor'
t(x)
```

**Arguments**

x                    a 3-tensor

**Details**

t(x)

**Value**

tensor transpose of x

**References**

M. Kilmer, K. Braman, N. Hao, and R. Hoover, "Third-order tensors as operators on matrices: a theoretical and computational framework with applications in imaging". *SIAM Journal on Matrix Analysis and Applications* 2013.

**Examples**

```
tnsr <- rand_tensor()
identical(t(tnsr@data[, , 1]), t(tnsr@data[, , 1]))
identical(t(tnsr@data[, , 2]), t(tnsr@data[, , 5]))
identical(t(t(tnsr)), tnsr)
```

---

tail-methods

*Tail for Tensor*

---

### Description

Extend tail for Tensor

### Usage

```
## S4 method for signature 'Tensor'  
tail(x, ...)
```

### Arguments

x                    the Tensor instance  
...                   additional parameters to be passed into tail()

### Details

```
tail(x, ...)
```

### See Also

[head-methods](#)

### Examples

```
tnsr <- rand_tensor()  
tail(tnsr)
```

---

Tensor-class

*S4 Class for a Tensor*

---

### Description

An S4 class for a tensor with arbitrary number of modes. The Tensor class extends the base 'array' class to include additional tensor manipulation (folding, unfolding, reshaping, subsetting) as well as a formal class definition that enables more explicit tensor algebra.

## Details

This can be seen as a wrapper class to the base array class. While it is possible to create an instance using `new`, it is also possible to do so by passing the data into `as.tensor`.

Each slot of a Tensor instance can be obtained using `@`.

The following methods are overloaded for the Tensor class: `dim-methods`, `head-methods`, `tail-methods`, `print-methods`, `show-methods`, element-wise array operations, array subsetting (extract via `'[']`), array subset replacing (replace via `'[<-']`), and `tperm-methods`, which is a wrapper around the base `aperm` method.

To sum across any one mode of a tensor, use the function `modeSum-methods`. To compute the mean across any one mode, use `modeMean-methods`.

You can always unfold any Tensor into a matrix, and the `unfold-methods`, `k_unfold-methods`, and `matvec-methods` methods are for that purpose. The output can be kept as a Tensor with 2 modes or a matrix object. The vectorization function is also provided as `vec`. See the attached vignette for a visualization of the different unfoldings.

Conversion from array/matrix to Tensor is facilitated via `as.tensor`. To convert from a Tensor instance, simply invoke `@data`.

The Frobenius norm of the Tensor is given by `fnorm-methods`, while the inner product between two Tensors (of equal modes) is given by `innerProd-methods`. You can also sum through any one mode to obtain the K-1 Tensor sum using `modeSum-methods`. `modeMean-methods` provides similar functionality to obtain the K-1 Tensor mean. These are primarily meant to be used internally but may be useful in doing statistics with Tensors.

For Tensors with 3 modes, we also overloaded `t` (transpose) defined by Kilmer et.al (2013). See `t-methods`.

To create a Tensor with i.i.d. random normal(0, 1) entries, see `rand_tensor`.

## Slots

**num\_modes** number of modes (integer)

**modes** vector of modes (integer), aka sizes/extents/dimensions

**data** actual data of the tensor, which can be 'array' or 'vector'

## Methods

`[ signature(tnsr = "Tensor"): ...`

`[<- signature(tnsr = "Tensor"): ...`

**matvec** signature(tnsr = "Tensor"): ...

**dim** signature(tnsr = "Tensor"): ...

**fnorm** signature(tnsr = "Tensor"): ...

**head** signature(tnsr = "Tensor"): ...

**initialize** signature(.Object = "Tensor"): ...

**innerProd** signature(tnsr1 = "Tensor", tnsr2 = "Tensor"): ...

**modeMean** signature(tnsr = "Tensor"): ...

```
modeSum signature(tnsr = "Tensor"): ...  
Ops signature(e1 = "array", e2 = "Tensor"): ...  
Ops signature(e1 = "numeric", e2 = "Tensor"): ...  
Ops signature(e1 = "Tensor", e2 = "array"): ...  
Ops signature(e1 = "Tensor", e2 = "numeric"): ...  
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tperm signature(tnsr = "Tensor"): ...  
image signature(tnsr = "Tensor"): ...
```

### Note

All of the decompositions and regression models in this package require a Tensor input.

### Author(s)

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### References

James Li, Jacob Bien, Martin T. Wells (2018). rTensor: An R Package for Multidimensional Array (Tensor) Unfolding, Multiplication, and Decomposition. *Journal of Statistical Software*, 87(10), 1-31. URL <http://www.jstatsoft.org/v087/i10/>.

### See Also

[as.tensor](#)

### Examples

```
tnsr <- rand_tensor()  
class(tnsr)  
tnsr  
print(tnsr)  
dim(tnsr)  
tnsr@num_modes  
tnsr@data
```

---

tperm-methods	<i>Mode Permutation for Tensor</i>
---------------	------------------------------------

---

**Description**

Overloads aperm for Tensor class for convenience.

**Usage**

```
tperm(tnsr, perm, ...)
```

```
## S4 method for signature 'Tensor'
```

```
tperm(tnsr, perm, ...)
```

**Arguments**

tnsr	the Tensor instance
perm	the new permutation of the current modes
...	additional parameters to be passed into aperm

**Details**

```
tperm(tnsr, perm=NULL, ...)
```

**Examples**

```
tnsr <- rand_tensor(c(3,4,5))
```

```
dim(tperm(tnsr, perm=c(2,1,3)))
```

```
dim(tperm(tnsr, perm=c(1,3,2)))
```

---

ttl	<i>Tensor Times List</i>
-----	--------------------------

---

**Description**

Contracted (m-Mode) product between a Tensor of arbitrary number of modes and a list of matrices. The result is folded back into Tensor.

**Usage**

```
ttl(tnsr, list_mat, ms = NULL)
```

**Arguments**

tnsr	Tensor object with K modes
list_mat	a list of matrices
ms	a vector of modes to contract on (order should match the order of list_mat)



**Details**

Performs `ttm` repeated for a single Tensor and a list of matrices on multiple modes. For instance, suppose we want to do multiply a Tensor object `tnsr` with three matrices `mat1`, `mat2`, `mat3` on modes 1, 2, and 3. We could do `ttm(ttm(ttm(tnsr,mat1,1),mat2,2),3)`, or we could do `ttl(tnsr,list(mat1,mat2,mat3),c(1,2,3))`. The order of the matrices in the list should obviously match the order of the modes. This is a common operation for various Tensor decompositions such as CP and Tucker. For the math on the m-Mode Product, see Kolda and Bader (2009).

**Value**

Tensor object with K modes

**Note**

The returned Tensor does not drop any modes equal to 1.

**References**

T. Kolda, B. Bader, "Tensor decomposition and applications". SIAM Applied Mathematics and Applications 2009.

**See Also**

[ttm](#)

**Examples**

```
tnsr <- new("Tensor",3L,c(3L,4L,5L),data=runif(60))
litz <- list('mat1' = matrix(runif(30),ncol=3),
'mat2' = matrix(runif(40),ncol=4),
'mat3' = matrix(runif(50),ncol=5))
ttl(tnsr,litz,ms=c(1,2,3))
```

---

ttm

*Tensor Times Matrix (m-Mode Product)*

---

**Description**

Contracted (m-Mode) product between a Tensor of arbitrary number of modes and a matrix. The result is folded back into Tensor.

**Usage**

```
ttm(tnsr, mat, m = NULL)
```

**Arguments**

<code>tnsr</code>	Tensor object with K modes
<code>mat</code>	input matrix with same number columns as the <code>m</code> th mode of <code>tnsr</code>
<code>m</code>	the mode to contract on

**Details**

By definition, `rs_unfold(ttm(tnsr,mat),m) = mat%*%rs_unfold(tnsr,m)`, so the number of columns in `mat` must match the `m`th mode of `tnsr`. For the math on the `m`-Mode Product, see Kolda and Bader (2009).

**Value**

a Tensor object with K modes

**Note**

The `m`th mode of `tnsr` must match the number of columns in `mat`. By default, the returned Tensor does not drop any modes equal to 1.

**References**

T. Kolda, B. Bader, "Tensor decomposition and applications". SIAM Applied Mathematics and Applications 2009.

**See Also**

[ttl](#), [rs\\_unfold-methods](#)

**Examples**

```
tnsr <- new("Tensor", 3L, c(3L, 4L, 5L), data=runif(60))
mat <- matrix(runif(50), ncol=5)
ttm(tnsr, mat, m=3)
```

---

tucker

*Tucker Decomposition*

---

**Description**

The Tucker decomposition of a tensor. Approximates a K-Tensor using a `n`-mode product of a core tensor (with modes specified by `ranks`) with orthogonal factor matrices. If there is no truncation in one of the modes, then this is the same as the MPCA, [mpca](#). If there is no truncation in all the modes (i.e. `ranks = tnsr@modes`), then this is the same as the HOSVD, [hosvd](#). This is an iterative algorithm, with two possible stopping conditions: either relative error in Frobenius norm has gotten below `tol`, or the `max_iter` number of iterations has been reached. For more details on the Tucker decomposition, consult Kolda and Bader (2009).

**Usage**

```
tucker(tnsr, ranks = NULL, max_iter = 25, tol = 1e-05)
```

**Arguments**

tnsr	Tensor with K modes
ranks	a vector of the modes of the output core Tensor
max_iter	maximum number of iterations if error stays above tol
tol	relative Frobenius norm error tolerance

**Details**

Uses the Alternating Least Squares (ALS) estimation procedure also known as Higher-Order Orthogonal Iteration (HOOI). Initialized using a (Truncated-)HOSVD. A progress bar is included to help monitor operations on large tensors.

**Value**

a list containing the following:

- Z the core tensor, with modes specified by ranks
- U a list of orthogonal factor matrices - one for each mode, with the number of columns of the matrices given by ranks
- conv whether or not  $\text{resid} < \text{tol}$  by the last iteration
- est estimate of tnsr after compression
- norm\_percent the percent of Frobenius norm explained by the approximation
- fnorm\_resid the Frobenius norm of the error  $\text{fnorm}(\text{est} - \text{tnsr})$
- all\_resids vector containing the Frobenius norm of error for all the iterations

**Note**

The length of ranks must match `tnsr@num_modes`.

**References**

T. Kolda, B. Bader, "Tensor decomposition and applications". SIAM Applied Mathematics and Applications 2009.

**See Also**

[hosvd](#), [mpca](#)

**Examples**

```
tnsr <- rand_tensor(c(4,4,4,4))
tuckerD <- tucker(tnsr,ranks=c(2,2,2,2))
tuckerD$conv
tuckerD$norm_percent
plot(tuckerD$all_resids)
```

---

t_mult	<i>Tensor Multiplication (T-MULT)</i>
--------	---------------------------------------

---

**Description**

Implements T-MULT based on block circulant matrices (Kilmer et al. 2013) for 3-tensors.

**Usage**

```
t_mult(x, y)
```

**Arguments**

x	a 3-tensor
y	another 3-tensor

**Details**

Uses the Fast Fourier Transform (FFT) speed up suggested by Kilmer et al. 2013 instead of explicitly constructing the block circulant matrix. For the mathematical details of T-MULT, see Kilmer et al. (2013).

**Value**

tensor product between x and y

**Note**

This only works (so far) between 3-Tensors.

**References**

M. Kilmer, K. Braman, N. Hao, and R. Hoover, "Third-order tensors as operators on matrices: a theoretical and computational framework with applications in imaging". SIAM Journal on Matrix Analysis and Applications 2013.

**Examples**

```
tnsr <- new("Tensor", 3L, c(3L, 4L, 5L), data=runif(60))
tnsr2 <- new("Tensor", 3L, c(4L, 3L, 5L), data=runif(60))
t_mult(tnsr, tnsr2)
```

---

`t_svd`*Tensor Singular Value Decomposition*

---

**Description**

TSVD for a 3-Tensor. Constructs 3-Tensors  $U, S, V$  such that  $\text{tnsr} = \text{t\_mult}(\text{t\_mult}(U, S), \text{t}(V))$ .  $U$  and  $V$  are orthogonal 3-Tensors with orthogonality defined in Kilmer et al. (2013), and  $S$  is a 3-Tensor consists of facewise diagonal matrices. For more details on the TSVD, consult Kilmer et al. (2013).

**Usage**

```
t_svd(tnsr)
```

**Arguments**

`tnsr`                    3-Tensor to decompose via TSVD

**Value**

a list containing the following:

$U$  the left orthogonal 3-Tensor

$V$  the right orthogonal 3-Tensor

$S$  the middle 3-Tensor consisting of face-wise diagonal matrices

**Note**

Computation involves complex values, but if the inputs are real, then the outputs are also real. Some loss of precision occurs in the truncation of the imaginary components during the FFT and inverse FFT.

**References**

M. Kilmer, K. Braman, N. Hao, and R. Hoover, "Third-order tensors as operators on matrices: a theoretical and computational framework with applications in imaging". *SIAM Journal on Matrix Analysis and Applications* 2013.

**See Also**

[t\\_mult](#), [t\\_svd\\_reconstruct](#)

**Examples**

```
tnsr <- rand_tensor()
tsvd <- t_svd(tnsr)
```

---

t_svd_reconstruct	<i>Reconstruct Tensor From TSVD</i>
-------------------	-------------------------------------

---

**Description**

Reconstruct the original 3-Tensor after it has been decomposed into U, S, V via [t\\_svd](#).

**Usage**

```
t_svd_reconstruct(L)
```

**Arguments**

L                    list that is an output from [t\\_svd](#)

**Value**

a 3-Tensor

**See Also**

[t\\_svd](#)

**Examples**

```
tnsr <- rand_tensor(c(10,10,10))
tsvdD <- t_svd(tnsr)
1 - fnorm(t_svd_reconstruct(tsvdD)-tnsr)/fnorm(tnsr)
```

---

unfold-methods	<i>Tensor Unfolding</i>
----------------	-------------------------

---

**Description**

Unfolds the tensor into a matrix, with the modes in *rs* onto the rows and modes in *cs* onto the columns. Note that *c(rs,cs)* must have the same elements (order doesn't matter) as *x@modes*. Within the rows and columns, the order of the unfolding is determined by the order of the modes. This convention is consistent with Kolda and Bader (2009).

**Usage**

```
unfold(tnsr, row_idx, col_idx)

## S4 method for signature 'Tensor'
unfold(tnsr, row_idx = NULL, col_idx = NULL)
```

**Arguments**

tnsr            the Tensor instance  
row\_idx        the indices of the modes to map onto the row space  
col\_idx        the indices of the modes to map onto the column space

**Details**

For Row Space Unfolding or m-mode Unfolding, see [rs\\_unfold-methods](#). For Column Space Unfolding or matvec, see [cs\\_unfold-methods](#).

[vec-methods](#) returns the vectorization of the tensor.

```
unfold(tnsr, row_idx=NULL, col_idx=NULL)
```

**Value**

matrix with `prod(row_idx)` rows and `prod(col_idx)` columns

**References**

T. Kolda, B. Bader, "Tensor decomposition and applications". SIAM Applied Mathematics and Applications 2009.

**See Also**

[k\\_unfold-methods](#) and [matvec-methods](#)

**Examples**

```
tnsr <- rand_tensor()
matT3 <- unfold(tnsr, row_idx=2, col_idx=c(3,1))
```

---

unmatvec

*Unmatvec Folding of Matrix*


---

**Description**

The inverse operation to [matvec-methods](#), turning a matrix into a Tensor. For a full account of matrix folding/unfolding operations, consult Kolda and Bader (2009).

**Usage**

```
unmatvec(mat, modes = NULL)
```

**Arguments**

mat            matrix to be folded into a Tensor  
modes         the modes of the output Tensor

**Value**

Tensor object with modes given by modes

**References**

T. Kolda, B. Bader, "Tensor decomposition and applications". SIAM Applied Mathematics and Applications 2009.

**See Also**

[matvec-methods](#), [fold](#), [k\\_fold](#)

**Examples**

```
tnsr <- new("Tensor", 3L, c(3L, 4L, 5L), data=runif(60))
matT1 <- matvec(tnsr)
identical(unmatvec(matT1, modes=c(3, 4, 5)), tnsr)
```

---

vec-methods

*Tensor Vec*

---

**Description**

Turns the tensor into a single vector, following the convention that earlier indices vary slower than later indices.

**Usage**

```
vec(tnsr)

## S4 method for signature 'Tensor'
vec(tnsr)
```

**Arguments**

tnsr            the Tensor instance

**Details**

```
vec(tnsr)
```

**Value**

vector with length `prod(x@modes)`

**References**

T. Kolda, B. Bader, "Tensor decomposition and applications". SIAM Applied Mathematics and Applications 2009.



**Examples**

```
tnsr <- rand_tensor(c(4,5,6,7))
vec(tnsr)
```

[-methods

*Extract or Replace Subtensors***Description**

Extends '[' and '[<-' from the base array class for the Tensor class. Works exactly as it would for the base 'array' class.

**Usage**

```
## S4 method for signature 'Tensor'
x[i, j, ..., drop = TRUE]

## S4 replacement method for signature 'Tensor'
x[i, j, ...] <- value
```

**Arguments**

x	Tensor to be subset
i, j, ...	indices that specify the extents of the sub-tensor
drop	whether or not to reduce the number of modes to exclude those that have '1' as the mode
value	either vector, matrix, or array that will replace the subtensor

**Details**

```
x[i,j,...,drop=TRUE]
```

**Value**

an object of class Tensor

**Examples**

```
tnsr <- rand_tensor()
tnsr[1,2,3]
tnsr[3,1,]
tnsr[, ,5]
tnsr[, ,5,drop=FALSE]

tnsr[1,2,3] <- 3; tnsr[1,2,3]
tnsr[3,1,] <- rep(0,5); tnsr[3,1,]
tnsr[,2,] <- matrix(0,nrow=3,ncol=5); tnsr[,2,]
```

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