
pyttb Documentation

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Tensors (also known as multidimensional arrays or N-way arrays) are used in a variety of applications ranging from chemometrics to network analysis.

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FUNCTIONALITY

pyttb provides the following classes and functions for manipulating dense, sparse, and structured tensors, along with algorithms for computing low-rank tensor models.

- **Tensor Classes**

pyttb supports multiple tensor types, including dense and sparse, as well as specially structured tensors, such as the Krusal format (stored as factor matrices).

- **Algorithms**

CP methods such as alternating least squares, direct optimization, and weighted optimization (for missing data). Also alternative decompositions such as Poisson Tensor Factorization via alternating Poisson regression.

2.1 Python API

2.1.1 Tensor Classes

`pyttb.ktensor`

class `pyttb.ktensor`

Bases: `object`

KTENSOR Class for Kruskal tensors (decomposed).

Contains the following data members:

weights: `numpy.ndarray` vector containing the weights of the rank-1 tensors defined by the outer products of the column vectors of the `factor_matrices`.

factor_matrices: `list` of `numpy.ndarray`. The length of the list is equal to the number of dimensions of the tensor. The shape of the *i*th element of the list is (n_i, r), where n_i is the length dimension *i* and r is the rank of the tensor (as well as the length of the weights vector).

Although the constructor `__init__()` can be used to create an empty `pyttb.ktensor`, there are several class methods that can be used to create an instance of this class:

- `from_data()`
- `from_tensor_type()`
- `from_factor_matrices()`
- `from_function()`
- `from_vector()`

Examples

For all examples listed below, the following module imports are assumed:

```
>>> import pyttb as ttb
>>> import numpy as np
```

classmethod `from_data(weights, *factor_matrices)`

Construct a `pyttb.ktensor` from weights and factor matrices.

The length of the list or the number of arguments specified by `factor_matrices` must equal the length of `weights`. See `pyttb.ktensor` for parameter descriptions.

Parameters

- **weights** (`numpy.ndarray`, required)
- **factor_matrices** (list of `numpy.ndarray` or variable number of `numpy.ndarray`, required)

Returns

`pyttb.ktensor`

Examples

Create a `pyttb.ktensor` from weights and a list of factor matrices:

```
>>> weights = np.array([1., 2.])
>>> fm0 = np.array([[1., 2.], [3., 4.]])
>>> fm1 = np.array([[5., 6.], [7., 8.]])
>>> K = ttb.ktensor.from_data(weights, [fm0, fm1])
>>> print(K)
ktensor of shape 2 x 2
weights=[1. 2.]
factor_matrices[0] =
[[1. 2.]
 [3. 4.]]
factor_matrices[1] =
[[5. 6.]
 [7. 8.]]
```

Create a `pyttb.ktensor` from weights and factor matrices passed as arguments:

```
>>> K = ttb.ktensor.from_data(weights, fm0, fm1)
>>> print(K)
ktensor of shape 2 x 2
weights=[1. 2.]
factor_matrices[0] =
[[1. 2.]
 [3. 4.]]
factor_matrices[1] =
[[5. 6.]
 [7. 8.]]
```

classmethod `from_tensor_type(source) → ktensor`

Construct a `pyttb.ktensor` from another `pyttb.ktensor`. A deep copy of the data from the input `pyttb.ktensor` is used for the new `pyttb.ktensor`.

Parameters

source (`pyttb.ktensor`, required)

Returns

`pyttb.ktensor`

Examples

Create an instance of a *pyttb.ktensor*:

```
>>> fm0 = np.array([[1., 2.], [3., 4.]])
>>> fm1 = np.array([[5., 6.], [7., 8.]])
>>> factor_matrices = [fm0, fm1]
>>> K_source = ttb.ktensor.from_factor_matrices(factor_matrices)
>>> print(K_source)
ktensor of shape 2 x 2
weights=[1. 1.]
factor_matrices[0] =
[[1. 2.]
 [3. 4.]]
factor_matrices[1] =
[[5. 6.]
 [7. 8.]]
```

Create another instance of a *pyttb.ktensor* from the original one above:

```
>>> K = ttb.ktensor.from_tensor_type(K_source)
>>> print(K)
ktensor of shape 2 x 2
weights=[1. 1.]
factor_matrices[0] =
[[1. 2.]
 [3. 4.]]
factor_matrices[1] =
[[5. 6.]
 [7. 8.]]
```

See also *pyttb.ktensor.copy()*

classmethod *from_factor_matrices*(**factor_matrices*)

Construct a *pyttb.ktensor* from factor matrices. The weights of the returned *pyttb.ktensor* will all be equal to 1.

Parameters

factor_matrices (list of *numpy.ndarray* or variable number of *numpy.ndarray*, required) – The number of columns of each of the factor matrices must be the same.

Returns

pyttb.ktensor

Examples

Create a *pyttb.ktensor* from a list of factor matrices:

```
>>> fm0 = np.array([[1., 2.], [3., 4.]])
>>> fm1 = np.array([[5., 6.], [7., 8.]])
>>> factor_matrices = [fm0, fm1]
>>> K = ttb.ktensor.from_factor_matrices(factor_matrices)
>>> print(K)
ktensor of shape 2 x 2
```

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```
weights=[1. 1.]
factor_matrices[0] =
[[1. 2.]
 [3. 4.]]
factor_matrices[1] =
[[5. 6.]
 [7. 8.]]
```

Create a `pyttb.ktensor` from factor matrices passed as arguments:

```
>>> K = ttb.ktensor.from_factor_matrices(fm0, fm1)
>>> print(K)
ktensor of shape 2 x 2
weights=[1. 1.]
factor_matrices[0] =
[[1. 2.]
 [3. 4.]]
factor_matrices[1] =
[[5. 6.]
 [7. 8.]]
```

classmethod `from_function`(*fun, shape, num_components*)

Construct a `pyttb.ktensor` whose factor matrix entries are set using a function. The weights of the returned `pyttb.ktensor` will all be equal to 1.

Parameters

- **fun** (*function, required*) – A function that can accept a shape (i.e., `tuple` of dimension sizes) and return a `numpy.ndarray` of that shape. Example functions include `numpy.random.random_sample`, `numpy.zeros`, `numpy.ones`.
- **shape** (`tuple`, *required*)
- **num_components** (*int, required*)

Returns

`pyttb.ktensor`

Examples

Create a `pyttb.ktensor` with entries of the factor matrices taken from a uniform random distribution:

```
>>> np.random.seed(1)
>>> K = ttb.ktensor.from_function(np.random.random_sample, (2, 3, 4), 2)
>>> print(K)
ktensor of shape 2 x 3 x 4
weights=[1. 1.]
factor_matrices[0] =
[[4.1702...e-01 7.2032...e-01]
 [1.1437...e-04 3.0233...e-01]]
factor_matrices[1] =
[[0.1467... 0.0923...]
 [0.1862... 0.3455...]
 [0.3967... 0.5388...]]
```

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```
factor_matrices[2] =
[[0.4191... 0.6852...]
 [0.2044... 0.8781...]
 [0.0273... 0.6704...]
 [0.4173... 0.5586...]]
```

Create a `pyttb.ktensor` with entries equal to 1:

```
>>> K = ttb.ktensor.from_function(np.ones, (2, 3, 4), 2)
>>> print(K)
ktensor of shape 2 x 3 x 4
weights=[1. 1.]
factor_matrices[0] =
[[1. 1.]
 [1. 1.]]
factor_matrices[1] =
[[1. 1.]
 [1. 1.]
 [1. 1.]]
factor_matrices[2] =
[[1. 1.]
 [1. 1.]
 [1. 1.]
 [1. 1.]]
```

Create a `pyttb.ktensor` with entries equal to 0:

```
>>> K = ttb.ktensor.from_function(np.zeros, (2, 3, 4), 2)
>>> print(K)
ktensor of shape 2 x 3 x 4
weights=[1. 1.]
factor_matrices[0] =
[[0. 0.]
 [0. 0.]]
factor_matrices[1] =
[[0. 0.]
 [0. 0.]
 [0. 0.]]
factor_matrices[2] =
[[0. 0.]
 [0. 0.]
 [0. 0.]
 [0. 0.]]
```

classmethod `from_vector`(*data*, *shape*, *contains_weights*)

Construct a `pyttb.ktensor` from a vector (given as a `numpy.ndarray`) and shape (given as a `numpy.ndarray`). The rank of the `pyttb.ktensor` is inferred from the shape and length of the vector.

Parameters

- **data** (`numpy.ndarray`, required) – Vector containing either elements of the factor matrices (when ``contains_weights`==False`) or elements of the weights and factor matrices (when ``contains_weights`==True`). When both the elements of the weights and the factor_matrices are present, the weights come first and the columns of the factor matrices

come next.

- **shape** (`numpy.ndarray`, required) – Vector containing the shape of the tensor (i.e., lengths of the dimensions).
- **contains_weights** (*bool, required*) – Flag to specify whether or not *data* contains weights. If False, all weights are set to 1.

Returns

`pyttb.ktensor`

Examples

Create a `pyttb.ktensor` from a vector containing only elements of the factor matrices:

```
>>> rank = 2
>>> shape = np.array([2, 3, 4])
>>> data = np.arange(1, rank*sum(shape)+1).astype(float)
>>> K = ttb.ktensor.from_vector(data[:,], shape, False)
>>> print(K)
ktensor of shape 2 x 3 x 4
weights=[1. 1.]
factor_matrices[0] =
[[1. 3.]
 [2. 4.]]
factor_matrices[1] =
[[ 5.  8.]
 [ 6.  9.]
 [ 7. 10.]]
factor_matrices[2] =
[[11. 15.]
 [12. 16.]
 [13. 17.]
 [14. 18.]]
```

Create a `pyttb.ktensor` from a vector containing elements of both the weights and the factor matrices:

```
>>> weights = 2 * np.ones(rank).astype(float)
>>> weights_and_data = np.concatenate((weights, data), axis=0)
>>> K = ttb.ktensor.from_vector(weights_and_data[:,], shape, True)
>>> print(K)
ktensor of shape 2 x 3 x 4
weights=[2. 2.]
factor_matrices[0] =
[[1. 3.]
 [2. 4.]]
factor_matrices[1] =
[[ 5.  8.]
 [ 6.  9.]
 [ 7. 10.]]
factor_matrices[2] =
[[11. 15.]
 [12. 16.]
 [13. 17.]
 [14. 18.]]
```

arrange(*weight_factor=None, permutation=None*)

Arrange the rank-1 components of a `pyttb.ktensor` in place. If *permutation* is passed, the columns of *self.factor_matrices* are arranged using the provided permutation, so you must make a copy before calling this method if you want to store the original `pyttb.ktensor`. If *weight_factor* is passed, then the values in *self.weights* are absorbed into *self.factor_matrices[weight_factor]*. If no parameters are passed, then the columns of *self.factor_matrices* are normalized and then permuted such that the resulting *self.weights* are sorted by magnitude, greatest to least. Passing both parameters leads to an error.

Parameters

- **weight_factor** (*int, optional*) – The index of the factor matrix that the weights will be absorbed into.
- **permutation** (*tuple, list, or numpy.ndarray, optional*) – The new order of the components of the `pyttb.ktensor` into which to permute. The permutation must be of length equal to the number of components of the `pyttb.ktensor`, *self.ncomponents* and must be a permutation of $[0, \dots, \text{self.ncomponents}-1]$.

Examples

Create the initial `pyttb.ktensor`:

```
>>> weights = np.array([1., 2.])
>>> fm0 = np.array([[1., 2.], [3., 4.]])
>>> fm1 = np.array([[5., 6.], [7., 8.]])
>>> K = ttb.ktensor.from_data(weights, [fm0, fm1])
>>> print(K)
ktensor of shape 2 x 2
weights=[1. 2.]
factor_matrices[0] =
[[1. 2.]
 [3. 4.]]
factor_matrices[1] =
[[5. 6.]
 [7. 8.]]
```

Arrange the columns of the factor matrices using a permutation:

```
>>> p = [1,0]
>>> K.arrange(permutation=p)
>>> print(K)
ktensor of shape 2 x 2
weights=[2. 1.]
factor_matrices[0] =
[[2. 1.]
 [4. 3.]]
factor_matrices[1] =
[[6. 5.]
 [8. 7.]]
```

Normalize and permute columns such that *weights* are sorted in decreasing order:

```
>>> K.arrange()
>>> print(K)
```

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```

ktensor of shape 2 x 2
weights=[89.4427... 27.2029...]
factor_matrices[0] =
[[0.4472... 0.3162...]
 [0.8944... 0.9486...]]
factor_matrices[1] =
[[0.6... 0.5812...]
 [0.8... 0.8137...]]

```

Absorb the weights into the second factor:

```

>>> K.arrange(weight_factor=1)
>>> print(K)
ktensor of shape 2 x 2
weights=[1. 1.]
factor_matrices[0] =
[[0.4472... 0.3162...]
 [0.8944... 0.9486...]]
factor_matrices[1] =
[[53.6656... 15.8113...]
 [71.5541... 22.1359...]]

```

copy()

Make a deep copy of a *pyttb.ktensor*.

Returns

pyttb.ktensor

Examples

Create a random *pyttb.ktensor* with weights of 1:

```

>>> np.random.seed(1)
>>> K = ttb.ktensor.from_function(np.random.random_sample, (2, 3, 4), 2)
>>> print(K)
ktensor of shape 2 x 3 x 4
weights=[1. 1.]
factor_matrices[0] =
[[4.1702...e-01 7.2032...e-01]
 [1.1437...e-04 3.0233...e-01]]
factor_matrices[1] =
[[0.1467... 0.0923...]
 [0.1862... 0.3455...]
 [0.3967... 0.5388...]]
factor_matrices[2] =
[[0.4191... 0.6852...]
 [0.2044... 0.8781...]
 [0.0273... 0.6704...]
 [0.4173... 0.5586...]]

```

Create a copy of the *pyttb.ktensor* and change the weights:


```

>>> K2 = K.copy()
>>> K2.weights = np.array([2., 3.])
>>> print(K2)
ktensor of shape 2 x 3 x 4
weights=[2. 3.]
factor_matrices[0] =
[[4.1702...e-01 7.2032...e-01]
 [1.1437...e-04 3.023...e-01]]
factor_matrices[1] =
[[0.1467... 0.0923...]
 [0.1862... 0.3455...]
 [0.3967... 0.5388...]]
factor_matrices[2] =
[[0.4191... 0.6852...]
 [0.2044... 0.8781...]
 [0.0273... 0.6704...]
 [0.4173... 0.5586...]]

```

Show that the original `pyttb.ktensor` is unchanged:

```

>>> print(K)
ktensor of shape 2 x 3 x 4
weights=[1. 1.]
factor_matrices[0] =
[[4.1702...e-01 7.2032...e-01]
 [1.1437...e-04 3.0233...e-01]]
factor_matrices[1] =
[[0.1467... 0.0923...]
 [0.1862... 0.3455...]
 [0.3967... 0.5388...]]
factor_matrices[2] =
[[0.4191... 0.6852...]
 [0.2044... 0.8781...]
 [0.0273... 0.6704...]
 [0.4173... 0.5586...]]

```

`double()`

Convert `pyttb.ktensor` to `numpy.ndarray`.

Returns

`numpy.ndarray`

Examples

```

>>> weights = np.array([1., 2.])
>>> fm0 = np.array([[1., 2.], [3., 4.]])
>>> fm1 = np.array([[5., 6.], [7., 8.]])
>>> factor_matrices = [fm0, fm1]
>>> K = ttb.ktensor.from_data(weights, factor_matrices)
>>> K.double()
array([[29., 39.],
       [63., 85.]])

```

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```
>>> type(K.double())
<class 'numpy.ndarray'>
```

end(*k=None*)

Last index of indexing expression for *pyttb.ktensor*.

Parameters

k (*int, optional*) – dimension for subscripted indexing

Returns

int (*index*)

Examples

```
>>> K = ttb.ktensor.from_function(np.ones, (2, 3, 4), 2)
>>> print(K.end(2))
3
```

extract(*idx=None*)

Creates a new *pyttb.ktensor* with only the specified components.

Parameters

idx (*int, tuple, list, numpy.ndarray, optional*) – Index set of components to extract. It should be the case that *idx* is a subset of $[0, \dots, \text{self.ncomponents}]$. If this parameter is *None* or is empty, a copy of the *pyttb.ktensor* is returned.

Returns

pyttb.ktensor

Examples

Create a *pyttb.ktensor*:

```
>>> weights = np.array([1., 2.])
>>> fm0 = np.array([[1., 2.], [3., 4.]])
>>> fm1 = np.array([[5., 6.], [7., 8.]])
>>> K = ttb.ktensor.from_data(weights, [fm0, fm1])
>>> print(K)
ktensor of shape 2 x 2
weights=[1. 2.]
factor_matrices[0] =
[[1. 2.]
 [3. 4.]]
factor_matrices[1] =
[[5. 6.]
 [7. 8.]]
```

Create a new *pyttb.ktensor*, extracting only the second component from each factor of the original *pyttb.ktensor*:

```
>>> K.extract([1])
ktensor of shape 2 x 2
```

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```

weights=[2.]
factor_matrices[0] =
[[2.]
 [4.]]
factor_matrices[1] =
[[6.]
 [8.]]

```

fixsigns(*other=None*)

Change the elements of a *pyttb.ktensor* in place so that the largest magnitude entries for each column vector in each factor matrix are positive, provided that the sign on pairs of vectors in a rank-1 component can be flipped.

Parameters

other (*pyttb.ktensor*, optional) – If not None, returns a version of the *pyttb.ktensor* where some of the signs of the columns of the factor matrices have been flipped to better align with *other*. In not None, both *pyttb.ktensor* objects are first normalized (using *normalize()*).

Returns

pyttb.ktensor – The changes are made in place and a reference to the updated tensor is returned

Examples

Create a *pyttb.ktensor* with negative large magnitude entries:

```

>>> weights = np.array([1., 2.])
>>> fm0 = np.array([[1., 2.], [3., 4.]])
>>> fm1 = np.array([[5., 6.], [7., 8.]])
>>> K = ttb.ktensor.from_data(weights, [fm0, fm1])
>>> K.factor_matrices[0][1, 1] = -K.factor_matrices[0][1, 1]
>>> K.factor_matrices[1][1, 1] = -K.factor_matrices[1][1, 1]
>>> print(K)
ktensor of shape 2 x 2
weights=[1. 2.]
factor_matrices[0] =
[[ 1.  2.]
 [ 3. -4.]]
factor_matrices[1] =
[[ 5.  6.]
 [ 7. -8.]]

```

Fix the signs of the largest magnitude entries:

```

>>> print(K.fixsigns())
ktensor of shape 2 x 2
weights=[1. 2.]
factor_matrices[0] =
[[ 1. -2.]
 [ 3.  4.]]
factor_matrices[1] =

```

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```
[[ 5. -6.]
 [ 7.  8.]]
```

Fix the signs using another *pyttb.ktensor*:

```
>>> K = ttb.ktensor.from_data(weights, [fm0, fm1])
>>> K2 = K.copy()
>>> K2.factor_matrices[0][1, 1] = -K2.factor_matrices[0][1, 1]
>>> K2.factor_matrices[1][1, 1] = -K2.factor_matrices[1][1, 1]
>>> K = K.fixsigns(K2)
>>> print(K)
ktensor of shape 2 x 2
weights=[27.2029... 89.4427...]
factor_matrices[0] =
[[ 0.3162... -0.4472...]
 [ 0.9486... -0.8944...]]
factor_matrices[1] =
[[ 0.5812... -0.6...]
 [ 0.8137... -0.8...]]
```

full()

Convert a *pyttb.ktensor* to a *pyttb.tensor*.

Returns

pyttb.tensor

Examples

```
>>> weights = np.array([1., 2.])
>>> fm0 = np.array([[1., 2.], [3., 4.]])
>>> fm1 = np.array([[5., 6.], [7., 8.]])
>>> K = ttb.ktensor.from_data(weights, [fm0, fm1])
>>> print(K)
ktensor of shape 2 x 2
weights=[1. 2.]
factor_matrices[0] =
[[1. 2.]
 [3. 4.]]
factor_matrices[1] =
[[5. 6.]
 [7. 8.]]
>>> print(K.full())
tensor of shape 2 x 2
data[:, :] =
[[29. 39.]
 [63. 85.]]
```

innerprod(*other*)

Efficient inner product with a *pyttb.ktensor*.

Efficiently computes the inner product between two tensors, *self*

and *other*. If *other* is a *pyttb.ktensor*, the inner product is computed using inner products of the

factor matrices. Otherwise, the inner product is computed using the *ttv* (tensor times vector) of *other* with all of the columns of *self.factor_matrices*.

Parameters

other (*pyttb.ktensor*, *pyttb.sptensor*, *pyttb.tensor*, or *pyttb.ttensor*, required)
– Tensor with which to compute the inner product.

Returns

float

Examples

```
>>> K = ttb.ktensor.from_function(np.ones, (2,3,4), 2)
>>> print(K.innerprod(K))
96.0
```

isequal(*other*)

Equal comparator for *pyttb.ktensor* objects.

Parameters

other (*pyttb.ktensor*, required) – *pyttb.ktensor* with which to compare.

Returns

bool

Examples

```
>>> K1 = ttb.ktensor.from_function(np.ones, (2,3,4), 2)
>>> weights = np.ones((2, 1))
>>> factor_matrices = [np.ones((2, 2)), np.ones((3, 2)), np.ones((4, 2))]
>>> K2 = ttb.ktensor.from_data(weights, factor_matrices)
>>> print(K1.isequal(K2))
True
```

issymmetric(*return_diffs=False*)

Returns True if the *pyttb.ktensor* is exactly symmetric for every permutation.

Parameters

return_diffs (*bool*, *optional*) – If True, returns the matrix of the norm of the differences between the factor matrices.

Returns

- *bool*
- *numpy.ndarray*, *optional* – Matrix of the norm of the differences between the factor matrices

Examples

Create a *pyttb.ktensor* that is symmetric and test if it is symmetric:

```
>>> K = ttb.ktensor.from_function(np.ones, (3, 3, 3), 2)
>>> print(K.issymmetric())
True
```

Create a *pyttb.ktensor* that is not symmetric and return the differences:

```
>>> weights = np.array([1., 2.])
>>> fm0 = np.array([[1., 2.], [3., 4.]])
>>> fm1 = np.array([[5., 6.], [7., 8.]])
>>> K2 = ttb.ktensor.from_data(weights, [fm0, fm1])
>>> issym, diffs = K2.issymmetric(return_diffs=True)
>>> print(diffs)
[[0. 8.]
 [0. 0.]]
```

mask(W)

Extract *pyttb.ktensor* values as specified by *W*, a *pyttb.tensor* or *pyttb.sptensor* containing only values of zeros (0) and ones (1). The values in the *pyttb.ktensor* corresponding to the indices for the ones (1) in *W* will be returned as a column vector.

Parameters

W (*pyttb.tensor* or *pyttb.sptensor*, required)

Returns

numpy.ndarray

Examples

Create a *pyttb.ktensor*:

```
>>> weights = np.array([1., 2.])
>>> fm0 = np.array([[1., 2.], [3., 4.]])
>>> fm1 = np.array([[5., 6.], [7., 8.]])
>>> K = ttb.ktensor.from_data(weights, [fm0, fm1])
```

Create a mask *pyttb.tensor* and extract the elements of the *pyttb.ktensor* using the mask:

```
>>> W = ttb.tensor.from_data(np.array([[0, 1], [1, 0]]))
>>> print(K.mask(W))
[[63.]
 [39.]]
```

mttkrp(U, n)

Matricized tensor times Khatri-Rao product for *pyttb.ktensor*.

Efficiently calculates the matrix product of the *n*-mode matricization of the *ktensor* with the Khatri-Rao product of all entries in *U*, a *list* of factor matrices, except the *nth*.

Parameters

- *U* (*list* of factor matrices, required)
- *n* (*int*, required) – Multiply by all modes except *n*.

Returns*numpy.ndarray***Examples**

```

>>> K = ttb.ktensor.from_function(np.ones, (2, 3, 4), 2)
>>> U = [np.ones((2, 2)), np.ones((3, 2)), np.ones(((4, 2)))]
>>> print(K.mttkrp(U, 0))
[[24. 24.]
 [24. 24.]]

```

property ncomponents

Number of components in the *pyttb.ktensor* (i.e., number of columns in each factor matrix) of the *pyttb.ktensor*.

Returns*int***Examples**

```

>>> K = ttb.ktensor.from_function(np.ones, (2, 3, 4), 2)
>>> print(K.ncomponents)
2

```

property ndims

Number of dimensions (i.e., number of factor matrices) of the *pyttb.ktensor*.

Returns*int***Examples**

```

>>> K = ttb.ktensor.from_function(np.ones, (2, 3, 4), 2)
>>> print(K.ndims)
3

```

norm()

Compute the norm (i.e., square root of the sum of squares of entries) of a *pyttb.ktensor*.

Returns*int*

Examples

```
>>> K = ttb.ktensor.from_function(np.ones, (2, 3, 4), 2)
>>> K.norm()
9.797958971132712
```

normalize(*weight_factor=None, sort=False, normtype=2, mode=None*)

Normalize the columns of the factor matrices of a [pyttb.ktensor](#) in place.

Parameters

- **weight_factor** (*{“all”, int}, optional*) – Absorb the weights into one or more factors. If “all”, absorb weight equally across all factors. If *int*, absorb weight into a single dimension (value must be in `range(self.ndims)`).
- **sort** (*bool, optional*) – Sort the columns in descending order of the weights.
- **normtype** (*{non-negative int, -1, -2, np.inf, -np.inf}, optional*) – Order of the norm (see [numpy.linalg.norm\(\)](#) for possible values).
- **mode** (*int, optional*) – Index of factor matrix to normalize. A value of *None* means normalize all factor matrices.

Returns

[pyttb.ktensor](#)

Examples

```
>>> K = ttb.ktensor.from_function(np.ones, (2, 3, 4), 2)
>>> print(K.normalize())
ktensor of shape 2 x 3 x 4
weights=[4.898... 4.898...]
factor_matrices[0] =
[[0.7071... 0.7071...]
 [0.7071... 0.7071...]]
factor_matrices[1] =
[[0.5773... 0.5773...]
 [0.5773... 0.5773...]
 [0.5773... 0.5773...]]
factor_matrices[2] =
[[0.5 0.5]
 [0.5 0.5]
 [0.5 0.5]
 [0.5 0.5]]
```

nvecs(*n, r, flipsign=True*)

Compute the leading mode-*n* vectors for a [pyttb.ktensor](#).

Computes the *r* leading eigenvectors of $X_n * X_n.T$ (where X_n is the mode-*n* matricization/unfolding of self), which provides information about the mode-*N* fibers. In two-dimensions, the *r* leading mode-1 vectors are the same as the *r* left singular vectors and the *r* leading mode-2 vectors are the same as the *r* right singular vectors. By default, this method computes the top *r* eigenvectors of $X_n * X_n.T$.

Parameters

- **n** (*int, required*) – Mode for tensor matricization.

- **r** (*int, required*) – Number of eigenvectors to compute and use.
- **flipsign** (*bool, optional*) – If True, make each column's largest element positive.

Returns`numpy.ndarray`**Examples**

Compute single eigenvector for dimension 0:

```
>>> K = ttb.ktensor.from_function(np.ones, (2, 3, 4), 2)
>>> nvecs1 = K.nvecs(0, 1)
>>> print(nvecs1)
[[0.70710678...]
 [0.70710678...]]
```

Compute first 2 leading eigenvectors for dimension 0:

```
>>> nvecs2 = K.nvecs(0, 2)
>>> print(nvecs2)
[[ 0.70710678...  0.70710678...]
 [ 0.70710678... -0.70710678...]]
```

permute(*order*)

Permute *pyttb.ktensor* dimensions.

Rearranges the dimensions of a *pyttb.ktensor* so that they are in the order specified by *order*. The corresponding tensor has the same components as *self* but the order of the subscripts needed to access any particular element is rearranged as specified by *order*.

Parameters

order (`numpy.ndarray`) – Permutation of [0,...,self.ndimensions].

Returns*pyttb.ktensor***Examples**

```
>>> weights = np.array([1., 2.])
>>> fm0 = np.array([[1., 2.], [3., 4.]])
>>> fm1 = np.array([[5., 6.], [7., 8.]])
>>> factor_matrices = [fm0, fm1]
>>> K = ttb.ktensor.from_data(weights, factor_matrices)
>>> print(K)
ktensor of shape 2 x 2
weights=[1. 2.]
factor_matrices[0] =
[[1. 2.]
 [3. 4.]]
factor_matrices[1] =
[[5. 6.]
 [7. 8.]]
```

Permute the order of the dimension so they are in reverse order:

```
>>> K1 = K.permute(np.array([1, 0]))
>>> print(K1)
ktensor of shape 2 x 2
weights=[1. 2.]
factor_matrices[0] =
[[5. 6.]
 [7. 8.]]
factor_matrices[1] =
[[1. 2.]
 [3. 4.]]
```

redistribute(*mode*)

Distribute weights of a *pyttb.ktensor* to the specified mode. The redistribution is performed in place.

Parameters

mode (*int*) – Must be value in $[0, \dots, \text{self.ndims}]$.

Example

Create a *pyttb.ktensor*:

```
>>> weights = np.array([1., 2.])
>>> fm0 = np.array([[1., 2.], [3., 4.]])
>>> fm1 = np.array([[5., 6.], [7., 8.]])
>>> factor_matrices = [fm0, fm1]
>>> K = ttb.ktensor.from_data(weights, factor_matrices)
>>> print(K)
ktensor of shape 2 x 2
weights=[1. 2.]
factor_matrices[0] =
[[1. 2.]
 [3. 4.]]
factor_matrices[1] =
[[5. 6.]
 [7. 8.]]
```

Distribute weights of that *pyttb.ktensor* to mode 0:

```
>>> K.redistribute(0)
>>> print(K)
ktensor of shape 2 x 2
weights=[1. 1.]
factor_matrices[0] =
[[1. 4.]
 [3. 8.]]
factor_matrices[1] =
[[5. 6.]
 [7. 8.]]
```

property shape

Shape of a *pyttb.ktensor*.

Returns the lengths of all dimensions of the *pyttb.ktensor*.

Returns`tuple`**score**(*other*, *weight_penalty*=True, *threshold*=0.99, *greedy*=True)

Checks if two `pyttb.ktensor` instances with the same shapes but potentially different number of components match except for permutation.

Matching is defined as follows. If *self* and *other* are single- component `pyttb.ktensor` instances that have been normalized so that their weights are *self.weights* and *other.weights*, and their factor matrices are single column vectors containing $[a_1, a_2, \dots, a_n]$ and $[b_1, b_2, \dots, b_n]$, respectively, then the score is defined as

$$\text{score} = \text{penalty} * (a_1.T * b_1) * (a_2.T * b_2) * \dots * (a_n.T * b_n),$$

where the penalty is defined by the weights such that

$$\text{penalty} = 1 - \text{abs}(\text{self.weights} - \text{other.weights}) / \max(\text{self.weights}, \text{other.weights}).$$

The score of multi-component `pyttb.ktensor` instances is a normalized sum of the scores across the best permutation of the components of *self*. *self* can have more components than *other*; any extra components are ignored in terms of the matching score.

Parameters

- **other** (`pyttb.ktensor`, required) – `pyttb.ktensor` with which to match.
- **weight_penalty** (*bool*, optional) – Flag indicating whether or not to consider the weights in the calculations.
- **threshold** (*float*, optional) – Threshold specified in the formula above for determining a match.
- **greedy** (*bool*, optional) – Flag indicating whether or not to consider all possible matchings (exponentially expensive) or just do a greedy matching.

Returns

- *int* – Score (between 0 and 1).
- `pyttb.ktensor` – Copy of *self*, which has been normalized and permuted to best match *other*.
- *bool* – Flag indicating a match according to a user-specified threshold.
- `numpy.ndarray` – Permutation (i.e. array of indices of the modes of *self*) of the components of *self* that was used to best match *other*.

Examples

Create two `pyttb.ktensor` instances and compute the score between them:

```
>>> K = ttb.ktensor.from_data(np.array([2., 1., 3.]), np.ones((3,3)), np.
↳ ones((4,3)), np.ones((5,3)))
>>> K2 = ttb.ktensor.from_data(np.array([2., 4.]), np.ones((3,2)), np.ones((4,
↳ 2)), np.ones((5,2)))
>>> score, Kperm, flag, perm = K.score(K2)
>>> print(score)
0.875
>>> print(perm)
[0 2 1]
```

Compute score without using weights:

```
>>> score, Kperm, flag, perm = K.score(K2, weight_penalty=False)
>>> print(score)
1.0
>>> print(perm)
[0 1 2]
```

symmetrize()

Symmetrize a *pyttb.ktensor* in all modes.

Symmetrize a *pyttb.ktensor* with respect to all modes so that the resulting *pyttb.ktensor* is symmetric with respect to any permutation of indices.

Returns

pyttb.ktensor

Examples

Create a *pyttb.ktensor*:

```
>>> weights = np.array([1., 2.])
>>> fm0 = np.array([[1., 2.], [3., 4.]])
>>> fm1 = np.array([[5., 6.], [7., 8.]])
>>> factor_matrices = [fm0, fm1]
>>> K = ttb.ktensor.from_data(weights, factor_matrices)
>>> print(K)
ktensor of shape 2 x 2
weights=[1. 2.]
factor_matrices[0] =
[[1. 2.]
 [3. 4.]]
factor_matrices[1] =
[[5. 6.]
 [7. 8.]]
```

Make the factor matrices of the *pyttb.ktensor* symmetric with respect to any permutation of the factor matrices:

```
>>> K1 = K.symmetrize()
>>> print(K1)
ktensor of shape 2 x 2
weights=[1. 1.]
factor_matrices[0] =
[[2.3404... 4.9519...]
 [4.5960... 8.0124...]]
factor_matrices[1] =
[[2.3404... 4.9519...]
 [4.5960... 8.0124...]]
```

tolist(mode=None)

Convert *pyttb.ktensor* to a list of factor matrices, evenly distributing the weights across factors. Optionally absorb the weights into a single mode.

Parameters

mode (*int, optional*) – Index of factor matrix to absorb all of the weights.

Returnslist of `numpy.ndarray`**Examples**Create a `pyttb.ktensor` of all ones:

```

>>> weights = np.array([1., 2.])
>>> fm0 = np.array([[1., 2.], [3., 4.]])
>>> fm1 = np.array([[5., 6.], [7., 8.]])
>>> factor_matrices = [fm0, fm1]
>>> K = ttb.ktensor.from_data(weights, factor_matrices)
>>> print(K)
ktensor of shape 2 x 2
weights=[1. 2.]
factor_matrices[0] =
[[1. 2.]
 [3. 4.]]
factor_matrices[1] =
[[5. 6.]
 [7. 8.]]

```

Spread weights equally to all factors and return list of factor matrices:

```

>>> fm_list = K.tolist()
>>> for fm in fm_list: print(fm)
[[1. 2.8284...]
 [3. 5.6568...]]
[[ 5. 8.4852...]
 [ 7. 11.313...]]

```

Shift weight to single factor matrix and return list of factor matrices:

```

>>> fm_list = K.tolist(0)
>>> for fm in fm_list: print(fm)
[[ 8.6023... 40. ]
 [25.8069... 80. ]]
[[0.5812... 0.6...]
 [0.8137... 0.8...]]

```

tovec(*include_weights=True*)Convert `pyttb.ktensor` to column vector. Optionally include or exclude the weights.**Parameters****include_weights** (*bool, optional*) – Flag to specify whether or not to include weights in output.**Returns**`numpy.ndarray` – The length of the column vector is $(\text{sum}(\text{self.shape})+1)*\text{self.ncomponents}$. The vector contains the weights (if requested) stacked on top of each of the columns of the `factor_matrices` in order.

Examples

Create a *pyttb.ktensor* from a vector:

```
>>> rank = 2
>>> shape = np.array([2, 3, 4])
>>> data = np.arange(1, rank*sum(shape)+1)
>>> weights = 2 * np.ones(rank)
>>> weights_and_data = np.concatenate((weights, data), axis=0)
>>> K = ttb.ktensor.from_vector(weights_and_data[:], shape, True)
>>> print(K)
ktensor of shape 2 x 3 x 4
weights=[2. 2.]
factor_matrices[0] =
[[1. 3.]
 [2. 4.]]
factor_matrices[1] =
[[ 5.  8.]
 [ 6.  9.]
 [ 7. 10.]]
factor_matrices[2] =
[[11. 15.]
 [12. 16.]
 [13. 17.]
 [14. 18.]]
```

Create a *pyttb.ktensor* from a vector of data extracted from another *pyttb.ktensor*:

```
>>> K2 = ttb.ktensor.from_vector(K.tovec(), shape, True)
>>> print(K2)
ktensor of shape 2 x 3 x 4
weights=[2. 2.]
factor_matrices[0] =
[[1. 3.]
 [2. 4.]]
factor_matrices[1] =
[[ 5.  8.]
 [ 6.  9.]
 [ 7. 10.]]
factor_matrices[2] =
[[11. 15.]
 [12. 16.]
 [13. 17.]
 [14. 18.]]
```

ttv(*vector*, *dims=None*, *exclude_dims=None*)

Tensor times vector for a *pyttb.ktensor*.

Computes the product of a *pyttb.ktensor* with a vector (i.e., *np.array*). If *dims* is an integer, it specifies the dimension in the *pyttb.ktensor* along which the vector is multiplied. If the shape of the vector is = (I,1), then the length of dimension *dims* of the *pyttb.ktensor* must be I. Note that the number of dimensions of the returned *pyttb.ktensor* is 1 less than the dimension of the *pyttb.ktensor* used in the multiplication because dimension *dims* is removed.

If *vector* is a *list* of *np.array* instances, the *pyttb.ktensor* is multiplied with each vector in the list. The

products are computed sequentially along all dimensions (or modes) of the `pyttb.ktensor`, and thus the list must contain `self.ndims` vectors.

When `dims` is not `None`, compute the products along the dimensions specified by `dims`. In this case, the number of products can be less than `self.ndims` and the order of the sequence does not need to match the order of the dimensions in the `pyttb.ktensor`. Note that the number of vectors must match the number of dimensions provided, and the length of each vector must match the size of each dimension of the `pyttb.ktensor` specified in `dims`.

Parameters

- **vector** (`numpy.ndarray` or `list[numpy.ndarray]`, required)
- **dims** (int, `numpy.ndarray`, optional)
- **exclude_dims**

Returns

float or `pyttb.ktensor` – The number of dimensions of the returned `pyttb.ktensor` is `n-k`, where `n = self.ndims` and `k = number of vectors provided as input`. If `k == n`, a scalar is returned.

Examples

Compute the product of a `pyttb.ktensor` and a single vector (results in a `pyttb.ktensor`):

```
>>> rank = 2
>>> shape = np.array([2, 3, 4])
>>> data = np.arange(1, rank*sum(shape)+1)
>>> weights = 2 * np.ones(rank)
>>> weights_and_data = np.concatenate((weights, data), axis=0)
>>> K = ttb.ktensor.from_vector(weights_and_data[:], shape, True)
>>> K0 = K.ttv(np.array([1, 1, 1]), dims=1) # compute along a single dimension
>>> print(K0)
ktensor of shape 2 x 4
weights=[36. 54.]
factor_matrices[0] =
[[1. 3.]
 [2. 4.]]
factor_matrices[1] =
[[11. 15.]
 [12. 16.]
 [13. 17.]
 [14. 18.]
```

Compute the product of a `pyttb.ktensor` and a vector for each dimension (results in a `float`):

```
>>> vec2 = np.array([1, 1])
>>> vec3 = np.array([1, 1, 1])
>>> vec4 = np.array([1, 1, 1, 1])
>>> K1 = K.ttv([vec2, vec3, vec4])
>>> print(K1)
30348.0
```

Compute the product of a `pyttb.ktensor` and multiple vectors out of order (results in a `pyttb.ktensor`):

```
>>> K2 = K.ttv([vec4, vec3], np.array([2, 1]))
>>> print(K2)
ktensor of shape 2
weights=[1800. 3564.]
factor_matrices[0] =
[[1. 3.]
 [2. 4.]]
```

update(*modes*, *data*)

Updates a *pyttb.ktensor* in the specific dimensions with the values in *data* (in vector or matrix form). The value of *modes* must be a value in [-1,...,self.ndoms]. If the Further, the number of elements in *data* must equal self.shape[modes] * self.ncomponents. The update is performed in place.

Parameters

- **modes** (int or *list* of int, required) – List of dimensions to update; values must be in ascending order. If the first element of the list is -1, then update the weights. All other integer values values must be sorted and in [0,...,self.ndims].
- **data** (*numpy.ndarray*, required) – Data values to use in the update.
- **Results**
- **_____**
- **:class: `pyttb.ktensor`**

Examples

Create a *pyttb.ktensor* of all ones:

```
>>> K = ttb.ktensor.from_function(np.ones, (2, 3, 4), 2)
```

Create vectors for updating various factor matrices of the *pyttb.ktensor*:

```
>>> vec0 = 2 * np.ones(K.shape[0] * K.ncomponents)
>>> vec1 = 3 * np.ones(K.shape[1] * K.ncomponents)
>>> vec2 = 4 * np.ones(K.shape[2] * K.ncomponents)
```

Update a single factor matrix:

```
>>> K1 = K.copy()
>>> K1 = K1.update(0, vec0)
>>> print(K1)
ktensor of shape 2 x 3 x 4
weights=[1. 1.]
factor_matrices[0] =
[[2. 2.]
 [2. 2.]]
factor_matrices[1] =
[[1. 1.]
 [1. 1.]
 [1. 1.]]
factor_matrices[2] =
[[1. 1.]
```

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```
[1. 1.]
[1. 1.]
[1. 1.]
```

Update all factor matrices:

```
>>> K2 = K.copy()
>>> vec_all = np.concatenate((vec0, vec1, vec2))
>>> K2 = K2.update([0, 1, 2], vec_all)
>>> print(K2)
ktensor of shape 2 x 3 x 4
weights=[1. 1.]
factor_matrices[0] =
[[2. 2.]
 [2. 2.]]
factor_matrices[1] =
[[3. 3.]
 [3. 3.]
 [3. 3.]]
factor_matrices[2] =
[[4. 4.]
 [4. 4.]
 [4. 4.]
 [4. 4.]]
```

Update some but not all factor matrices:

```
>>> K3 = K.copy()
>>> vec_some = np.concatenate((vec0, vec2))
>>> K3 = K3.update([0, 2], vec_some)
>>> print(K3)
ktensor of shape 2 x 3 x 4
weights=[1. 1.]
factor_matrices[0] =
[[2. 2.]
 [2. 2.]]
factor_matrices[1] =
[[1. 1.]
 [1. 1.]
 [1. 1.]]
factor_matrices[2] =
[[4. 4.]
 [4. 4.]
 [4. 4.]
 [4. 4.]]
```

__add__(*other*)

Binary addition for *pyttb.ktensor*.

Parameters

other (*pyttb.ktensor*, required) – *pyttb.ktensor* to add to *self*.

Returns

pyttb.ktensor

__getitem__(item)

Subscripted reference for a *pyttb.ktensor*.

Subscripted reference is used to query the components of a *pyttb.ktensor*.

Parameters

item (*tuple(int) or int, required*)

Examples

```
>>> K = ttb.ktensor.from_function(np.ones, (2, 3, 4), 2)
>>> K.weights
array([[1., 1.]])
>>> K.factor_matrices
[array([[1., 1.],
        [1., 1.]]), array([[1., 1.],
        [1., 1.],
        [1., 1.]]), array([[1., 1.],
        [1., 1.],
        [1., 1.],
        [1., 1.]])]
>>> K.factor_matrices[0]
array([[1., 1.],
        [1., 1.]])
>>> K[0]
array([[1., 1.],
        [1., 1.]])
>>> K[1, 2, 0]
2.0
>>> K[0][:, [0]]
array([[1.],
        [1.]])
```

__neg__()

Unary minus (negative) for *pyttb.ktensor* instances.

Returns

pyttb.ktensor

__pos__()

Unary plus (positive) for *pyttb.ktensor* instances.

Returns

pyttb.ktensor

__setitem__(key, value)

Subscripted assignment for *pyttb.ktensor*.

Subscripted assignment cannot be used to update individual elements of a *pyttb.ktensor*. You can update the weights vector or the factor matrices of a *pyttb.ktensor*.

Example

```

>>> K = ttb.ktensor.from_data(np.ones((4,1)), [np.random.random((2,4)), np.
↳ random.random((3,4)), np.random.random((4,4))])
>>> K.weights = 2 * np.ones((4,1))
>>> K.factor_matrices[0] = np.zeros((2, 4))
>>> K.factor_matrices = [np.zeros((2, 4)), np.zeros((3, 4)), np.zeros((4, 4))]
>>> print(K)
ktensor of shape 2 x 3 x 4
weights=[[2.]
 [2.]
 [2.]]
factor_matrices[0] =
[[0. 0. 0. 0.]
 [0. 0. 0. 0.]]
factor_matrices[1] =
[[0. 0. 0. 0.]
 [0. 0. 0. 0.]
 [0. 0. 0. 0.]]
factor_matrices[2] =
[[0. 0. 0. 0.]
 [0. 0. 0. 0.]
 [0. 0. 0. 0.]
 [0. 0. 0. 0.]]

```

__sub__(*other*)Binary subtraction for *pyttb.ktensor*.**Parameters****other** (*pyttb.ktensor*)**Returns***pyttb.ktensor***__mul__**(*other*)Elementwise (including scalar) multiplication for *pyttb.ktensor* instances.**Parameters****other** (*pyttb.tensor*, *pyttb.sptensor*, float, int)**Returns***pyttb.ktensor***__rmul__**(*other*)Elementwise (including scalar) multiplication for *pyttb.ktensor* instances.**Parameters****other** (*pyttb.tensor*, *pyttb.sptensor*, float, int)**Returns***pyttb.ktensor***__repr__**()String representation of a *pyttb.ktensor*.**Returns***str*

`__str__()`

String representation of a `pyttb.ktensor`.

Returns

`str`

`__module__ = 'pyttb.ktensor'`

pyttb.sptensor

Sparse Tensor Implementation

`pyttb.sptensor.tt_to_sparse_matrix(sptensorInstance: sptensor, mode: int, transpose: bool = False) → coo_matrix`

Helper function to unwrap sptensor into sparse matrix, should replace the core need for sptenmat

Parameters

- **sptensorInstance** (*sparse tensor to unwrap*)
- **mode** (*Mode around which to unwrap tensor*)
- **transpose** (*Whether or not to tranpose unwrapped tensor*)

Returns

`spmatrix` (*unwrapped tensor*)

`pyttb.sptensor.tt_from_sparse_matrix(spmatrix: coo_matrix, shape: Any, mode: int, idx: int) → sptensor`

Helper function to wrap sparse matrix into sptensor. Inverse of `pyttb.tt_to_sparse_matrix`

Parameters

- **spmatrix** (`Scipy.sparse.coo_matrix`)
- **mode** (`int`) – Mode around which tensor was unwrapped
- **idx** (`int`) – in `{0,1}`, idx of mode in spmatrix, s.b. 0 for tranpose=True

Returns

`sptensorInstance` (`pyttb.sptensor`)

class `pyttb.sptensor.sptensor`

Bases: `object`

SPTENSOR Class for sparse tensors.

`__init__()`

Create an empty sparse tensor

Returns

`pyttb.sptensor`

classmethod `from_data(subs: ndarray, vals: ndarray, shape: Tuple[int, ...]) → sptensor`

Construct an sptensor from fully defined SUB, VAL and SIZE matrices. This does no validation to optimize for speed when components are known. For default initializer with error checking see `from_aggregator()`.

Parameters

- **subs** (*location of non-zero entries*)
- **vals** (*values for non-zero entries*)

- **shape** (*shape of sparse tensor*)

Examples

Import required modules:

```
>>> import pyttb as ttb
>>> import numpy as np
```

Set up input data # Create sptensor with explicit data description

```
>>> subs = np.array([[1, 2], [1, 3]])
>>> vals = np.array([[6], [7]])
>>> shape = (4, 4, 4)
>>> K0 = ttb.sptensor.from_data(subs,vals, shape)
```

classmethod from_tensor_type(*source: Union[sptensor, tensor, coo_matrix]*) → *sptensor*

Construct an *pyttb.sptensor* from compatible tensor types

Parameters

source (*Source tensor to create sptensor from*)

Returns

Generated Sparse Tensor

classmethod from_function(*function_handle: Callable[[Tuple[int, ...]], ndarray], shape: Tuple[int, ...], nonzeros: float*) → *sptensor*

Creates a sparse tensor of the specified shape with NZ nonzeros created from the specified function handle

Parameters

- **function_handle** (*function that accepts 2 arguments and generates*) – *numpy.ndarray* of length nonzeros
- **shape** (*tuple*)
- **nonzeros** (*int or float*)

Returns

Generated Sparse Tensor

classmethod from_aggregator(*subs: ndarray, vals: ndarray, shape: Optional[Tuple[int, ...]] = None, function_handle: Union[str, Callable[[Any], Union[float, ndarray]]] = 'sum'*) → *sptensor*

Construct an sptensor from fully defined SUB, VAL and shape matrices, after an aggregation is applied

Parameters

- **subs** (*location of non-zero entries*)
- **vals** (*values for non-zero entries*)
- **shape** (*shape of sparse tensor*)
- **function_handle** (*Aggregation function, or name of supported*) – aggregation function from *numpy_groupies*

Returns

Generated Sparse Tensor

Examples

```
>>> subs = np.array([[1, 2], [1, 3]])
>>> vals = np.array([[6], [7]])
>>> shape = np.array([4, 4])
>>> K0 = ttb.sptensor.from_aggregator(subs,vals)
>>> K1 = ttb.sptensor.from_aggregator(subs,vals,shape)
>>> function_handle = sum
>>> K2 = ttb.sptensor.from_aggregator(subs,vals,shape,function_handle)
```

allsubs() → ndarray

Generate all possible subscripts for sparse tensor

Returns

s (All possible subscripts for sptensor)

collapse(*dims*: ~typing.Optional[~numpy.ndarray] = None, *fun*: ~typing.Callable[[~numpy.ndarray], ~typing.Union[float, ~numpy.ndarray]] = <function sum>) → Union[float, ndarray, sptensor]

Collapse sparse tensor along specified dimensions.

Parameters

- **dims** (*Dimensions to collapse*)
- **fun** (*Method used to collapse dimensions*)

Returns

Collapsed value

Example

```
>>> subs = np.array([[1, 2], [1, 3]])
>>> vals = np.array([[1], [1]])
>>> shape = np.array([4, 4])
>>> X = ttb.sptensor.from_data(subs, vals, shape)
>>> X.collapse()
2
>>> X.collapse(np.arange(X.ndims), sum)
2
```

contract(*i*: int, *j*: int) → Union[ndarray, sptensor, tensor]

Contract tensor along two dimensions (array trace).

Parameters

- **i** (*First dimension*)
- **j** (*Second dimension*)

Returns

Contracted sptensor, converted to tensor if sufficiently dense

Example

```
>>> X = ttb.tensor.from_data(np.ones((2,2)))
>>> Y = sptensor.from_tensor_type(X)
>>> Y.contract(0, 1)
2.0
```

double() → ndarray

Convert sptensor to dense multidimensional array

elemfun(function_handle: Callable[[ndarray], ndarray]) → sptensor

Manipulate the non-zero elements of a sparse tensor

Parameters

function_handle (Function that updates all values.)

Returns

Updated sptensor

Example

```
>>> X = ttb.tensor.from_data(np.ones((2,2)))
>>> Y = sptensor.from_tensor_type(X)
>>> Z = Y.elemfun(lambda values: values*2)
>>> Z.isequal(Y*2)
True
```

end(k: Optional[int] = None) → int

Last index of indexing expression for sparse tensor

Parameters

k (int Dimension for subscript indexing)

extract(searchsubs: ndarray) → ndarray

Extract value for a sptensor.

Parameters

searchsubs (subscripts to find in sptensor)

See also:

[`__getitem__\(\)`](#)

find() → Tuple[ndarray, ndarray]

FIND Find subscripts of nonzero elements in a sparse tensor.

Returns

- **subs** (Subscripts of nonzero elements)
- **vals** (Values at corresponding subscripts)

full() → tensor

FULL Convert a sparse tensor to a (dense) tensor.

innerprod(other: Union[sptensor, tensor, ktensor, ttensor]) → float

Efficient inner product with a sparse tensor

Parameters**other** (*Other tensor to take innerproduct with*)**isequal**(*other*: *Union[sptensor, tensor]*) → bool

Exact equality for sptensors

Parameters**other** (*Other tensor to compare against*)**logical_and**(*B*: *Union[float, sptensor, tensor]*) → *sptensor*

Logical and with self and another object

Parameters**B** (*Other value to compare with*)**Returns***Indicator tensor***logical_not**() → *sptensor*

Logical NOT for sptensors

Returns

- *Sparse tensor with all zero-values marked from original*
- *sparse tensor*

logical_or(*B*: *Union[float, tensor]*) → *tensor***logical_or**(*B*: *sptensor*) → *sptensor*

Logical OR for sptensor and another value

Returns*Indicator tensor***logical_xor**(*other*: *Union[float, tensor]*) → *tensor***logical_xor**(*other*: *sptensor*) → *sptensor*

Logical XOR for sptensors

Parameters**other** (*Other value to xor against*)**Returns***Indicator tensor***mask**(*W*: *sptensor*) → ndarray

Extract values as specified by a mask tensor

Parameters**W** (*Mask tensor*)**Returns***Extracted values***mttkrp**(*U*: *Union[ktensor, List[ndarray]]*, *n*: *int*) → ndarray

Matricized tensor times Khatri-Rao product for sparse tensor.

Parameters

- **U** (*Matrices to create the Khatri-Rao product*)
- **n** (*Mode to matricize sptensor in*)

Returns*Matrix product***Examples**

```

>>> matrix = np.ones((4, 4))
>>> subs = np.array([[1, 1, 1], [1, 1, 3], [2, 2, 2], [3, 3, 3]])
>>> vals = np.array([[0.5], [1.5], [2.5], [3.5]])
>>> shape = (4, 4, 4)
>>> sptensorInstance = sptensor.from_data(subs, vals, shape)
>>> sptensorInstance.mttkrp(np.array([matrix, matrix, matrix]), 0)
array([[0. , 0. , 0. , 0. ],
       [2. , 2. , 2. , 2. ],
       [2.5, 2.5, 2.5, 2.5],
       [3.5, 3.5, 3.5, 3.5]])

```

property ndims: `int`

NDIMS Number of dimensions of a sparse tensor.

property nnz: `int`

Number of nonzeros in sparse tensor

norm() → floating

Compute the Frobenius norm of a sparse tensor.

nvecs(*n*: `int`, *r*: `int`, *flipsign*: `bool = True`) → `ndarray`Compute the leading mode-*n* vectors for a sparse tensor.**Parameters**

- **n** (*Mode to unfold*)
- **r** (*Number of eigenvectors to compute*)
- **flipsign** (*Make each eigenvector's largest element positive*)

ones() → *sptensor*

Replace nonzero elements of sparse tensor with ones

permute(*order*: `ndarray`) → *sptensor*

Rearrange the dimensions of a sparse tensor

Parameters

- **order** (*Updated order of dimensions*)

reshape(*new_shape*: `Tuple[int, ...]`, *old_modes*: `Optional[Union[ndarray, int]] = None`) → *sptensor*

Reshape specified modes of sparse tensor

Parameters

- **new_shape** (*tuple*)
- **old_modes** (`Numpy.ndarray`)

Returns*pyttb.sptensor*

scale(factor: *ndarray*, dims: *Union[float, ndarray]*) → *sptensor*

Scale along specified dimensions for sparse tensors

Parameters

- **factor** (*numpy.ndarray*)
- **dims** (int or *numpy.ndarray*)

Returns

pyttb.sptensor

spmatrix() → *coo_matrix*

Converts a two-way sparse tensor to a sparse matrix in *scipy.sparse.coo_matrix* format

squeeze() → *Union[sptensor, float]*

Remove singleton dimensions from a sparse tensor

Returns

pyttb.sptensor or float if sptensor is only singleton dimensions

subdims(region: *Sequence[Union[int, np.ndarray, slice]]*) → *np.ndarray*

SUBDIMS Compute the locations of subscripts within a subdimension.

Parameters

region (*numpy.ndarray* or tuple denoting indexing) – Subset of total sptensor shape in which to find non-zero values

Returns

numpy.ndarray – Index into subs for non-zero values in region

Examples

```
>>> subs = np.array([[1, 1, 1], [1, 1, 3], [2, 2, 2], [3, 3, 3]])
>>> vals = np.array([[0.5], [1.5], [2.5], [3.5]])
>>> shape = (4, 4, 4)
>>> sp = sptensor.from_data(subs,vals,shape)
>>> region = [np.array([1]), np.array([1]), np.array([1,3])]
>>> loc = sp.subdims(region)
>>> print(loc)
[0 1]
>>> region = (1, 1, slice(None, None, None))
>>> loc = sp.subdims(region)
>>> print(loc)
[0 1]
```

ttv(vector: *Union[ndarray, List[ndarray]]*, dims: *Optional[Union[ndarray, int]] = None*, exclude_dims: *Optional[Union[ndarray, int]] = None*) → *Union[sptensor, tensor]*

Sparse tensor times vector

Parameters

- **vector** (*Vector(s) to multiply against*)
- **dims** (*Dimensions to multiply with vector(s)*)
- **exclude_dims** (*Use all dimensions but these*)

__getitem__(item)

Subscripted reference for a sparse tensor.

We can extract elements or subtensors from a sparse tensor in the following ways.

Case 1a: $y = X(i_1, i_2, \dots, i_N)$, where each i_n is an index, returns a scalar.

Case 1b: $Y = X(R_1, R_2, \dots, R_N)$, where one or more R_n is a range and the rest are indices, returns a sparse tensor. The elements are renumbered here as appropriate.

Case 2a: $V = X(S)$ or $V = X(S, \text{'extract'})$, where S is a $p \times n$ array of subscripts, returns a vector of p values.

Case 2b: $V = X(I)$ or $V = X(I, \text{'extract'})$, where I is a set of p linear indices, returns a vector of p values.

Any ambiguity results in executing the first valid case. This is particularly an issue if $\text{ndims}(X)=1$.

Parameters

item (tuple(int), tuple(slice), :class:numpy.ndarray)

Returns

numpy.ndarray or pyttb.sptensor

Examples

```
>>> subs = np.array([[3,3,3],[1,1,0],[1,2,1]])
>>> vals = np.array([3,5,1])
>>> shape = (4,4,4)
>>> X = sptensor.from_data(subs,vals,shape)
>>> _ = X[0,1,0] #<-- returns zero
>>> _ = X[3,3,3] #<-- returns 3
>>> _ = X[2:3, :, :] #<-- returns 1 x 4 x 4 sptensor
```

__setitem__(key, value)

Subscripted assignment for sparse tensor.

We can assign elements to a sptensor in three ways.

Case 1: $X(R_1, R_2, \dots, R_N) = Y$, in which case we replace the rectangular subtensor (or single element) specified by the ranges R_1, \dots, R_N with Y . The right-hand-side can be a scalar or an sptensor.

Case 2: $X(S) = V$, where S is a $p \times n$ array of subscripts and V is a scalar value or a vector containing p values.

Linear indexing is not supported for sparse tensors.

Examples

$X = \text{sptensor}([30 \ 40 \ 20])$ <- Create an empty $30 \times 40 \times 20$ sptensor $X(30,40,20) = 7$ <- Assign a single element to be 7 $X([1,1,1;2,2,2]) = 1$ <- Assign a list of elements to the same value $X(11:20,11:20,11:20) = \text{sptenrand}([10,10,10],10)$ <- subtensor! $X(31,41,21) = 7$ <- grows the size of the tensor $X(111:120,111:120,111:120) = \text{sptenrand}([10,10,10],10)$ <- grows $X(1,1,1,1) = 4$ <- increases the number of dimensions from 3 to 4

$X = \text{sptensor}([30])$ <- empty one-dimensional tensor $X([4:6]) = 1$ <- set subtensor to ones (does not increase dimension) $X([10;12;14]) = (4:6)$ <- set three elements $X(31) = 7$ <- grow the first dimension $X(1,1) = 0$ <- add a dimension, but no nonzeros

Note regarding singleton dimensions: It is not possible to do, for instance, $X(1,1:10,1:10) = \text{sptenrand}([1 \ 10 \ 10],5)$. However, it is okay to do $X(1,1:10,1:10) = \text{squeeze}(\text{sptenrand}([1 \ 10 \ 10],5))$.

Parameters

- **key** (tuple(int),tuple(slice),:class:numpy.ndarray)
- **value** (int,float, numpy.ndarray, pyttb.sptensor)

__eq__(other)

Equal comparator for sptensors

Parameters**other** (compare equality of sptensor to other)**Returns**

pyttb.sptensor

__ne__(other)

Not equal comparator (~=) for sptensors

Parameters**other** (compare equality of sptensor to other)**Returns**

pyttb.sptensor

__sub__(other)

MINUS Binary subtraction for sparse tensors.

Parameters**other** (pyttb.tensor, pyttb.sptensor)**Returns**

pyttb.sptensor

__add__(other)

MINUS Binary addition for sparse tensors.

Parameters**other** (pyttb.tensor, pyttb.sptensor)**Returns**

pyttb.sptensor

__pos__()

Unary plus (+) for sptensors

Returns

pyttb.sptensor, copy of tensor

__neg__()

Unary minus (-) for sptensors

Returns

pyttb.sptensor, copy of tensor

__mul__(other)

Element wise multiplication (*) for sptensors

Parameters**other** (pyttb.sptensor, pyttb.tensor, float, int)**Returns**

pyttb.sptensor

`__rmul__(other)`

Element wise right multiplication (*) for sptensors

Parameters

other (*float, int*)

Returns

pyttb.sptensor

`__le__(other)`

Less than or equal (<=) for sptensor

Parameters

other (*pyttb.sptensor, pyttb.tensor, float, int*)

Returns

pyttb.sptensor

`__lt__(other)`

Less than (<) for sptensor

Parameters

other (*pyttb.sptensor, pyttb.tensor, float, int*)

Returns

pyttb.sptensor

`__ge__(other)`

Greater than or equal (>=) to for sptensor

Parameters

other (*pyttb.sptensor, pyttb.tensor, float, int*)

Returns

pyttb.sptensor

`__gt__(other)`

Greater than (>) to for sptensor

Parameters

other (*pyttb.sptensor, pyttb.tensor, float, int*)

Returns

pyttb.sptensor

`__truediv__(other)`

Division for sparse tensors (sptensor/other).

Parameters

other

`__rtruediv__(other)`

Right Division for sparse tensors (other/sptensor).

Parameters

other

`__repr__()`

String representation of a sparse tensor.

Returns

str – Contains the shape, subs and vals as strings on different lines.

`__hash__ = None`

`__module__ = 'pyttb.sptensor'`

`__str__()`

String representation of a sparse tensor.

Returns

str – Contains the shape, subs and vals as strings on different lines.

ttm(*matrices*: *Union[ndarray, List[ndarray]]*, *dims*: *Optional[Union[float, ndarray]] = None*, *exclude_dims*: *Optional[Union[float, ndarray]] = None*, *transpose*: *bool = False*)

Sparse tensor times matrix.

Parameters

- **matrices** (*A matrix or list of matrices*)
- **dims** (*Dimensions to multiply against*)
- **exclude_dims** (*Use all dimensions but these*)
- **transpose** (*Transpose matrices to be multiplied*)

`pyttb.sptensor.sptenrand`(*shape*: *Tuple[int, ...]*, *density*: *Optional[float] = None*, *nonzeros*: *Optional[float] = None*) → *sptensor*

Create sptensor with entries drawn from a uniform distribution on the unit interval

Parameters

- **shape** (*Shape of resulting tensor*)
- **density** (*Density of resulting sparse tensor*)
- **nonzeros** (*Number of nonzero entries in resulting sparse tensor*)

Returns

Constructed tensor

Example

```
>>> X = ttb.sptenrand((2,2), nonzeros=1)
>>> Y = ttb.sptenrand((2,2), density=0.25)
```

`pyttb.sptensor.sptendiag`(*elements*: *ndarray*, *shape*: *Optional[Tuple[int, ...]] = None*) → *sptensor*

Creates a sparse tensor with elements along super diagonal If provided shape is too small the tensor will be enlarged to accomodate

Parameters

- **elements** (*Elements to set along the diagonal*)
- **shape** (*Shape of resulting tensor*)

Returns

Constructed tensor

Example

```

>>> shape = (2,)
>>> values = np.ones(shape)
>>> X = ttb.sptendiag(values)
>>> Y = ttb.sptendiag(values, (2, 2))
>>> X.isequal(Y)
True

```

pyttb.tensor

Dense Tensor Implementation

class pyttb.tensor.tensor

Bases: `object`

TENSOR Class for dense tensors.

__init__()

TENSOR Create empty tensor.

data: `ndarray`

shape: `Tuple`

classmethod `from_data`(data: `ndarray`, shape: *Optional*[`Tuple`[`int`, ...]] = `None`) → *tensor*

Creates a tensor from explicit description. Note that 1D tensors (i.e., when `len(shape)==1`) contains a data array that follow the Numpy convention of being a row vector, which is different than in the Matlab Tensor Toolbox.

Parameters

- **data** (*Tensor source data*)
- **shape** (*Shape of resulting tensor if not the same as data shape*)

Returns

Constructed tensor

Example

```

>>> X = ttb.tensor.from_data(np.ones((2,2)))
>>> Y = ttb.tensor.from_data(np.ones((2,2)), shape=(4,1))

```

classmethod `from_tensor_type`(source: *Union*[`sptensor`, `tensor`, `ktensor`, `tenmat`]) → *tensor*

Converts other tensor types into a dense tensor

Parameters

source (*Tensor type to create dense tensor from*)

Returns

Constructed tensor

Example

```
>>> X = ttb.tensor.from_data(np.ones((2,2)))
>>> Y = ttb.tensor.from_tensor_type(X)
```

classmethod **from_function**(*function_handle: Callable[[Tuple[int, ...]], ndarray], shape: Tuple[int, ...]*)
→ *tensor*

Creates a tensor from a function handle and size

Parameters

- **function_handle** (*Function to generate data to construct tensor*)
- **shape** (*Shape of resulting tensor*)

Returns

Constructed tensor

Example

```
>>> X = ttb.tensor.from_function(lambda a_shape: np.ones(a_shape), (2,2))
```

collapse(*dims: ~typing.Optional[~numpy.ndarray] = None, fun: ~typing.Callable[[~numpy.ndarray], ~typing.Union[float, ~numpy.ndarray]] = <function sum>*) → *Union[float, ndarray, tensor]*

Collapse tensor along specified dimensions.

Parameters

- **dims** (*Dimensions to collapse*)
- **fun** (*Method used to collapse dimensions*)

Returns

Collapsed value

Example

```
>>> X = ttb.tensor.from_data(np.ones((2,2)))
>>> X.collapse()
4.0
>>> X.collapse(np.arange(X.ndims), sum)
4.0
```

contract(*i: int, j: int*) → *Union[ndarray, tensor]*

Contract tensor along two dimensions (array trace).

Parameters

- **i** (*First dimension*)
- **j** (*Second dimension*)

Returns

Contracted tensor

Example

```
>>> X = ttb.tensor.from_data(np.ones((2,2)))
>>> X.contract(0, 1)
2.0
```

double() → ndarray

Convert tensor to an array of doubles

Returns

Copy of tensor data

Example

```
>>> X = ttb.tensor.from_data(np.ones((2,2)))
>>> X.double()
array([[1., 1.],
       [1., 1.]])
```

exp() → tensor

Exponential of the elements of tensor

Returns

Copy of tensor data element-wise raised to exponential

Examples

```
>>> tensor1 = ttb.tensor.from_data(np.array([[1, 2], [3, 4]]))
>>> tensor1.exp().data
array([[ 2.7182...,  7.3890... ],
       [20.0855..., 54.5981...]])
```

end(k: Optional[int] = None) → int

Last index of indexing expression for tensor

Parameters

k (*dimension for subscripted indexing*)

Examples

```
>>> X = ttb.tensor.from_data(np.ones((2,2)))
>>> X.end() # linear indexing
3
>>> X.end(0)
1
```

find() → Tuple[ndarray, ndarray]

FIND Find subscripts of nonzero elements in a tensor.

S, V = FIND(X) returns the subscripts of the nonzero values in X and a column vector of the values.

Examples

```
>>> X = ttb.tensor.from_data(np.zeros((3,4,2)))
>>> larger_entries = X > 0.5
>>> subs, vals = larger_entries.find()
```

See also:

TENSOR, TENSOR

Returns

Subscripts and values for non-zero entries

full() → *tensor*

Convert dense tensor to dense tensor.

Returns

Deep copy

innerprod(other: Union[tensor, sptensor, ktensor]) → float

Efficient inner product with a tensor

Parameters

other (*Tensor type to take an innerproduct with*)

Examples

```
>>> tensor1 = ttb.tensor.from_data(np.array([[1, 2], [3, 4]]))
>>> tensor1.innerprod(tensor1)
30
```

isequal(other: Union[tensor, sptensor]) → bool

Exact equality for tensors

Parameters

other (*Tensor to compare against*)

Examples

```
>>> X = ttb.tensor.from_data(np.ones((2,2)))
>>> Y = ttb.tensor.from_data(np.zeros((2,2)))
>>> X.isequal(Y)
False
```

issymmetric(grps: Optional[ndarray] = None, version: Optional[Any] = None, return_details: bool = False) → Union[bool, Tuple[bool, ndarray, ndarray]]

Determine if a dense tensor is symmetric in specified modes.

Parameters

- **grps** (*Modes to check for symmetry*)
- **version** (*Flag*) – Any non-None value will call the non-default old version
- **return_details** (*Flag to return symmetry details in addition to bool*)

Returns*If symmetric in modes, optionally all differences and permutations***Examples**

```

>>> X = ttb.tensor.from_data(np.ones((2,2)))
>>> X.issymmetric()
True
>>> X.issymmetric(grps=np.arange(X.ndims))
True
>>> is_sym, diffs, perms = X.issymmetric(grps=np.arange(X.ndims),
↳ version=1, return_details=True)
>>> print(f"Tensor is symmetric: {is_sym}")
Tensor is symmetric: True
>>> print(f"Differences in modes: {diffs}")
Differences in modes: [[0.]
[0.]]
>>> print(f"Permutations: {perms}")
Permutations: [[0. 1.]
[1. 0.]]

```

logical_and(*B*: *Union[float, tensor]*) → *tensor*

Logical and for tensors

Parameters**B** (*Value to and against self*)**Examples**

```

>>> X = ttb.tensor.from_data(np.ones((2,2), dtype=bool))
>>> X.logical_and(X).collapse() # All true
4

```

logical_not() → *tensor*

Logical Not For Tensors

Returns*Negated tensor***Examples**

```

>>> X = ttb.tensor.from_data(np.ones((2,2), dtype=bool))
>>> X.logical_not().collapse() # All false
0

```

logical_or(*other*: *Union[float, tensor]*) → *tensor*

Logical or for tensors

Parameters**other** (*Value to perform or against*)

Examples

```
>>> X = ttb.tensor.from_data(np.ones((2,2), dtype=bool))
>>> X.logical_or(X.logical_not()).collapse() # All true
4
```

logical_xor(*other*: *Union[float, tensor]*) → *tensor*

Logical xor for tensors

Parameters

other (*Value to perform xor against*)

Examples

```
>>> X = ttb.tensor.from_data(np.ones((2,2), dtype=bool))
>>> X.logical_xor(X.logical_not()).collapse() # All true
4
```

mask(*W*: *tensor*) → *ndarray*

Extract non-zero values at locations specified by mask tensor

Parameters

W (*Mask tensor*)

Returns

Extracted values

Examples

```
>>> W = ttb.tensor.from_data(np.ones((2,2)))
>>> tensor1 = ttb.tensor.from_data(np.array([[1, 2], [3, 4]]))
>>> tensor1.mask(W)
array([1, 3, 2, 4])
```

mttkrp(*U*: *Union[ktensor, List[ndarray]]*, *n*: *int*) → *ndarray*

Matricized tensor times Khatri-Rao product

Parameters

- **U** (*Matrices to create the Khatri-Rao product*)
- **n** (*Mode to matricize tensor in*)

Returns

Matrix product

Example

```
>>> tensor1 = ttb.tensor.from_data(np.ones((2,2,2)))
>>> matrices = [np.ones((2,2))] * 3
>>> tensor1.mttkrp(matrices, 2)
array([[4., 4.],
       [4., 4.]])
```

property ndims: `int`

Return the number of dimensions of a tensor

Examples

```
>>> X = ttb.tensor.from_data(np.ones((2,2)))
>>> X.ndims
2
```

property nnz: `int`

Number of non-zero elements in tensor

Examples

```
>>> X = ttb.tensor.from_data(np.ones((2,2)))
>>> X.nnz
4
```

norm() → `floating`

Frobenius Norm of Tensor

Examples

```
>>> X = ttb.tensor.from_data(np.ones((2,2)))
>>> X.norm()
2.0
```

nvecs(*n*: `int`, *r*: `int`, *flipsign*: `bool = True`) → `ndarray`

Compute the leading mode-*n* eigenvectors for a tensor

Parameters

- **n** (*Mode to unfold*)
- **r** (*Number of eigenvectors to compute*)
- **flipsign** (*Make each eigenvector's largest element positive*)

Examples

```
>>> tensor1 = ttb.tensor.from_data(np.array([[1, 2], [3, 4]]))
>>> tensor1.nvecs(0,1)
array([[0.4045...],
       [0.9145...]])
>>> tensor1.nvecs(0,2)
array([[ 0.4045...,  0.9145...],
       [ 0.9145..., -0.4045...]])
```

permute(order: *ndarray*) → *tensor*

Permute tensor dimensions.

Parameters

order (*New order of tensor dimensions*)

Returns

Updated tensor with shapeNew == shapePrevious[order]

Examples

```
>>> X = ttb.tensor.from_data(np.ones((2,2)))
>>> Y = X.permute(np.array((1,0)))
>>> X.isequal(Y)
True
```

reshape(shape: *Tuple[int, ...]*) → *tensor*

Reshapes a tensor

Parameters

shape (*New shape*)

Examples

```
>>> X = ttb.tensor.from_data(np.ones((2,2)))
>>> Y = X.reshape((4,1))
>>> Y.shape
(4, 1)
```

squeeze() → *Union[tensor, ndarray, float]*

Removes singleton dimensions from a tensor

Returns

Tensor or scalar if all dims squeezed

Examples

```
>>> tensor1 = ttb.tensor.from_data(np.array([[[[4]]]])
>>> tensor1.squeeze()
4
>>> tensor2 = ttb.tensor.from_data(np.array([[1, 2, 3]]))
>>> tensor2.squeeze().data
array([1, 2, 3])
```

symmetrize(*grps*: *Optional*[*ndarray*] = *None*, *version*: *Optional*[*Any*] = *None*) → *tensor*

Symmetrize a tensor in the specified modes .. rubric:: Notes

It is *the same or less* work to just call `X = symmetrize(X)` then to first check if `X` is symmetric and then symmetrize it, even if `X` is already symmetric.

Parameters

- **grps** (*Modes to check for symmetry*)
- **version** (*Any non-None value will call the non-default old version*)

ttm(*matrix*: *Union*[*ndarray*, *List*[*ndarray*]], *dims*: *Optional*[*Union*[*float*, *ndarray*]] = *None*, *exclude_dims*: *Optional*[*Union*[*ndarray*, *int*]] = *None*, *transpose*: *bool* = *False*) → *tensor*

Tensor times matrix

Parameters

- **matrix** (*Matrix or matrices to multiply by*)
- **dims** (*Dimensions to multiply against*)
- **exclude_dims** (*Use all dimensions but these*)
- **transpose** (*Transpose matrices during multiplication*)

ttt(*other*: *tensor*, *selfdims*: *Optional*[*Union*[*ndarray*, *int*]] = *None*, *otherdims*: *Optional*[*Union*[*ndarray*, *int*]] = *None*) → *tensor*

Tensor multiplication (tensor times tensor)

Parameters

- **other** (*Tensor to multiply by*)
- **selfdims** (*Dimensions to contract this tensor by for multiplication*)
- **otherdims** (*Dimensions to contract other tensor by for multiplication*)

ttv(*vector*: *Union*[*ndarray*, *List*[*ndarray*]], *dims*: *Optional*[*Union*[*ndarray*, *int*]] = *None*, *exclude_dims*: *Optional*[*Union*[*ndarray*, *int*]] = *None*) → *tensor*

Tensor times vector

Parameters

- **vector** (*Vector(s) to multiply against*)
- **dims** (*Dimensions to multiply with vector(s)*)
- **exclude_dims** (*Use all dimensions but these*)

ttsv(*vector*: *Union*[*ndarray*, *List*[*ndarray*]], *skip_dim*: *Optional*[*int*] = *None*, *version*: *Optional*[*int*] = *None*) → *Union*[*ndarray*, *tensor*]

Tensor times same vector in multiple modes

Parameters

- **vector** (*Vector(s) to multiply against*)
- **skip_dim** (*Multiply tensor by vector in all dims except [0, skip_dim]*)

__setitem__(*key, value*)

SUBSASGN Subscripted assignment for a tensor.

We can assign elements to a tensor in three ways.

Case 1: $X(R1, R2, \dots, RN) = Y$, in which case we replace the rectangular subtensor (or single element) specified by the ranges $R1, \dots, RN$ with Y . The right-hand-side can be a scalar, a tensor, or an MDA.

Case 2a: $X(S) = V$, where S is a $p \times n$ array of subscripts and V is a scalar or a vector containing p values.

Case 2b: $X(I) = V$, where I is a set of p linear indices and V is a scalar or a vector containing p values. Resize is not allowed in this case.

Examples $X = \text{tensor}(\text{rand}(3,4,2))$ $X(1:2,1:2,1) = \text{ones}(2,2)$ \leftarrow replaces subtensor $X([1\ 1\ 1; 1\ 1\ 2]) = [5;7]$ \leftarrow replaces two elements $X([1;13]) = [5;7]$ \leftarrow does the same thing $X(1,1,2:3) = 1$ \leftarrow grows tensor $X(1,1,4) = 1$ \leftarrow grows the size of the tensor

__getitem__(*item*)

SUBSREF Subscripted reference for tensors.

We can extract elements or subtensors from a tensor in the following ways.

Case 1a: $y = X(i1, i2, \dots, iN)$, where each i is an index, returns a scalar.

Case 1b: $Y = X(R1, R2, \dots, RN)$, where one or more R_n is a range and the rest are indices, returns a sparse tensor.

Case 2a: $V = X(S)$ or $V = X(S, \text{'extract'})$, where S is a $p \times n$ array of subscripts, returns a vector of p values.

Case 2b: $V = X(I)$ or $V = X(I, \text{'extract'})$, where I is a set of p linear indices, returns a vector of p values.

Any ambiguity results in executing the first valid case. This is particularly an issue if $\text{ndims}(X) == 1$.

Examples $X = \text{tensor}(\text{rand}(3,4,2,1), [3\ 4\ 2\ 1])$; $X.\text{data} \leftarrow$ returns multidimensional array $X.\text{size} \leftarrow$ returns size $X(1,1,1,1) \leftarrow$ produces a scalar $X(1,1,1,:) \leftarrow$ produces a tensor of order 1 and size 1 $X(:,1,1,:) \leftarrow$ produces a tensor of size 3×1 $X(1:2,[2\ 4],1,:) \leftarrow$ produces a tensor of size $2 \times 2 \times 1$ $X(1:2,[2\ 4],1,1) \leftarrow$ produces a tensor of size 2×2 $X([1,1,1,1;3,4,2,1]) \leftarrow$ returns a vector of length 2 $X = \text{tensor}(\text{rand}(10,1),10)$; $X([1:6]) \leftarrow$ extracts a subtensor $X([1:6], \text{'extract'}) \leftarrow$ extracts a vector of 6 elements

Returns*pyttb.tensor* or *numpy.ndarray***__eq__**(*other*)

Equal for tensors

Parameters**other** (*pyttb.tensor*, float, int)**Returns***pyttb.tensor***__ne__**(*other*)

Not equal (!=) for tensors

Parameters**other** (*pyttb.tensor*, float, int)**Returns***pyttb.tensor*

`__ge__(other)`

Greater than or equal (\geq) for tensors

Parameters

other (*pyttb.tensor*, float, int)

Returns

pyttb.tensor

`__le__(other)`

Less than or equal (\leq) for tensors

Parameters

other (*pyttb.tensor*, float, int)

Returns

pyttb.tensor

`__gt__(other)`

Greater than ($>$) for tensors

Parameters

other (*pyttb.tensor*, float, int)

Returns

pyttb.tensor

`__lt__(other)`

Less than ($<$) for tensors

Parameters

other (*pyttb.tensor*, float, int)

Returns

pyttb.tensor

`__sub__(other)`

Binary subtraction ($-$) for tensors

Parameters

other (*pyttb.tensor*, float, int)

Returns

pyttb.tensor

`__add__(other)`

Binary addition ($+$) for tensors

Parameters

other (*pyttb.tensor*, float, int)

Returns

pyttb.tensor

`__radd__(other)`

Right binary addition ($+$) for tensors

Parameters

other (*pyttb.tensor*, float, int)

Returns

pyttb.tensor

__pow__(*power*)
Element Wise Power (**) for tensors

Parameters
other (*pyttb.tensor*, float, int)

Returns
pyttb.tensor

__mul__(*other*)
Element wise multiplication (*) for tensors, self*other

Parameters
other (*pyttb.tensor*, float, int)

Returns
pyttb.tensor

__rmul__(*other*)
Element wise right multiplication (*) for tensors, other*self

Parameters
other (*pyttb.tensor*, float, int)

Returns
pyttb.tensor

__truediv__(*other*)
Element wise left division (/) for tensors, self/other

Parameters
other (*pyttb.tensor*, float, int)

Returns
pyttb.tensor

__rtruediv__(*other*)
Element wise right division (/) for tensors, other/self

Parameters
other (*pyttb.tensor*, float, int)

Returns
pyttb.tensor

__pos__()
Unary plus (+) for tensors

Returns
pyttb.tensor – copy of tensor

__neg__()
Unary minus (-) for tensors

Returns
pyttb.tensor – copy of tensor

__repr__()
String representation of a tensor.

Returns
str – Contains the shape and data as strings on different lines.

`__str__()`

String representation of a tensor.

Returns

str – Contains the shape and data as strings on different lines.

`__annotations__ = {'data': 'np.ndarray', 'shape': 'Tuple'}`

`__hash__ = None`

`__module__ = 'pyttb.tensor'`

`pyttb.tensor.tenones(shape: Tuple[int, ...]) → tensor`

Creates a tensor of all ones

Parameters

shape (*Shape of resulting tensor*)

Returns

Constructed tensor

Example

```
>>> X = ttb.tenones((2,2))
```

`pyttb.tensor.tenzeros(shape: Tuple[int, ...]) → tensor`

Creates a tensor of all zeros

Parameters

shape (*Shape of resulting tensor*)

Returns

Constructed tensor

Example

```
>>> X = ttb.tenzeros((2,2))
```

`pyttb.tensor.tenrand(shape: Tuple[int, ...]) → tensor`

Creates a tensor with entries drawn from a uniform distribution on the unit interval

Parameters

shape (*Shape of resulting tensor*)

Returns

Constructed tensor

Example

```
>>> X = ttb.tenrand((2,2))
```

`pyttb.tensor.tendiag`(*elements: ndarray, shape: Optional[Tuple[int, ...]] = None*) → *tensor*

Creates a tensor with elements along super diagonal. If provided shape is too small the tensor will be enlarged to accomodate.

Parameters

- **elements** (*Elements to set along the diagonal*)
- **shape** (*Shape of resulting tensor*)

Returns

Constructed tensor

Example

```
>>> shape = (2,)
>>> values = np.ones(shape)
>>> X = ttb.tendiag(values)
>>> Y = ttb.tendiag(values, (2, 2))
>>> X.isequal(Y)
True
```

pyttb.ttensor

class `pyttb.ttensor.ttensor`

Bases: `object`

TTENSOR Class for Tucker tensors (decomposed).

__init__()

Create an empty decomposed tucker tensor

Returns

pyttb.ttensor

classmethod `from_data`(*core, factors*)

Construct an ttensor from fully defined core tensor and factor matrices.

Parameters

- **core** (:class: *ttb.tensor*)
- **factors** (list(*numpy.ndarray*))

Returns

pyttb.ttensor

Examples

Import required modules:

```
>>> import pyttb as ttb
>>> import numpy as np
```

Set up input data # Create tensor with explicit data description

```
>>> core_values = np.ones((2,2,2))
>>> core = ttb.tensor.from_data(core_values)
>>> factors = [np.ones((1,2))] * len(core_values.shape)
>>> K0 = ttb.ttensor.from_data(core, factors)
```

classmethod `from_tensor_type(source)`

Converts other tensor types into a tensor

Parameters

`source` (*pyttb.ttensor*)

Returns

pyttb.ttensor

property `shape`

Shape of the tensor this deconstruction represents.

Returns

tuple(int)

__repr__()

String representation of a tucker tensor.

Returns

str – Contains the core, and factor matrices as strings on different lines.

__str__()

String representation of a tucker tensor.

Returns

str – Contains the core, and factor matrices as strings on different lines.

full()

Convert a ttensor to a (dense) tensor.

Returns

pyttb.tensor

double()

Convert ttensor to an array of doubles

Returns

numpy.ndarray – copy of tensor data

property `ndims`

Number of dimensions of a ttensor.

Returns

int – Number of dimensions of tensor

isequal(*other*)

Component equality for tensors

Parameters**other** (*pyttb.ttensor*)**Returns****bool** (*True if ttensors decompositions are identical, false otherwise*)**__pos__**()

Unary plus (+) for tensors. Does nothing.

Returns*pyttb.ttensor*, copy of tensor**__neg__**()

Unary minus (-) for tensors

Returns*pyttb.ttensor*, copy of tensor**innerprod**(*other*)

Efficient inner product with a tensor

Parameters

- **other** (*pyttb.tensor*, *pyttb.sptensor*, *pyttb.ktensor*,)
- **:class:** *pyttb.ttensor*

Returns*float***__mul__**(*other*)

Element wise multiplication (*) for tensors (only scalars supported)

Parameters**other** (*float*, *int*)**Returns***pyttb.ttensor***__rmul__**(*other*)

Element wise right multiplication (*) for tensors (only scalars supported)

Parameters**other** (*float*, *int*)**Returns***pyttb.ttensor***ttv**(*vector*, *dims=None*, *exclude_dims=None*)

TTensor times vector

Parameters

- **vector** (*Numpy.ndarray*, *list[Numpy.ndarray]*)
- **dims** (*Numpy.ndarray*, *int*)

mttkrp(*U*, *n*)

Matricized tensor times Khatri-Rao product for tensors.

Parameters

- **U** (array of matrices or ktensor)
- **n** (multiplies by all modes except n)

Returns`numpy.ndarray`**norm()**

Compute the norm of a tensor. :returns: **norm** (float, Frobenius norm of Tensor)

permute(order)

Permute dimensions for a tensor

Parameters

order (Numpy.ndarray)

Returns`pyttb.ttensor`**ttm(matrix, dims=None, exclude_dims=None, transpose=False)**

Tensor times matrix for ttensor

Parameters

- **matrix** (Numpy.ndarray, list[Numpy.ndarray])
- **dims** (Numpy.ndarray, int)
- **transpose** (bool)

reconstruct(samples=None, modes=None)

Reconstruct or partially reconstruct tensor from ttensor.

Parameters

- **samples** (Numpy.ndarray, list[Numpy.ndarray])
- **modes** (Numpy.ndarray, list[Numpy.ndarray])

Returns`pyttb.ttensor`**nvecs(n, r, flipsign=True)**

Compute the leading mode-n vectors for a tensor.

Parameters

- **n** (mode for tensor matricization)
- **r** (number of eigenvalues)
- **flipsign** (Make each column's largest element positive if true)

Returns`numpy.ndarray`

`__module__ = 'pyttb.ttensor'`

pyttb.tenmat**class** pyttb.tenmat.tenmatBases: `object`

TENMAT Store tensor as a matrix.

__init__()

TENSOR Create empty tensor.

classmethod **from_data**(data, rdims, cdims=None, tshape=None)**classmethod** **from_tensor_type**(source, rdims=None, cdims=None, cdims_cyclic=None)**ctranspose**()

Complex conjugate transpose for tenmat.

Returns`pyttb.tenmat`**double**()

Convert tenmat to an array of doubles

Returns`numpy.ndarray` – copy of tenmat data**end**(k)

Last index of indexing expression for tenmat

Parameters**k** (*int*) – dimension for subscripted indexing**Returns****int** (*index*)**property** ndims

Return the number of dimensions of a tenmat

Returns*int***norm**()

Frobenius norm of a tenmat.

Returns*float***property** shape

Return the shape of a tenmat

Returns*tuple***__setitem__**(key, value)

SUBSASGN Subscripted assignment for a tensor.

__getitem__(item)

SUBSREF Subscripted reference for tenmat.

Parameters**item**

Returns`numpy.ndarray`, float, int`__mul__`(*other*)

Multiplies two tenmat objects.

Parameters**other** (`pyttb.tenmat`)**Returns**`pyttb.tenmat``__rmul__`(*other*)

Multiplies two tenmat objects.

Parameters**other** (`pyttb.tenmat`)**Returns**`pyttb.tenmat``__add__`(*other*)`__radd__`(*other*)

Reverse binary addition (+) for tenmats

Parameters**other** (`pyttb.tenmat`, float, int)**Returns**`pyttb.tenmat``__sub__`(*other*)`__rsub__`(*other*)

Reverse binary subtraction (-) for tenmats

Parameters**other** (`pyttb.tenmat`, float, int)**Returns**`pyttb.tenmat``__pos__`()

Unary plus (+) for tenmats

Returns`pyttb.tenmat` – copy of tenmat`__neg__`()

Unary minus (-) for tenmats

Returns`pyttb.tenmat` – copy of tenmat`__repr__`()

String representation of a tenmat.

Returns*str* – Contains the shape, row indices (rindices), column indices (cindices) and data as strings on different lines.

`__str__()`

String representation of a tenmat.

Returns

str – Contains the shape, row indices (rindices), column indices (cindices) and data as strings on different lines.

`__module__` = 'pyttb.tenmat'

2.1.2 Algorithms

pyttb.cp_als

`pyttb.cp_als.cp_als(input_tensor, rank, stoptol=0.0001, maxiters=1000, dimorder=None, init='random', printitn=1, fixsigns=True)`

Compute CP decomposition with alternating least squares

Parameters

- **input_tensor** (*pyttb.tensor* or *pyttb.sptensor* or *pyttb.ktensor*)
- **rank** (*int*) – Rank of the decomposition
- **stoptol** (*float*) – Tolerance used for termination - when the change in the fitness function in successive iterations drops below this value, the iterations terminate (default: 1e-4)
- **dimorder** (*list*) – Order to loop through dimensions (default: `[range(tensor.ndims)]`)
- **maxiters** (*int*) – Maximum number of iterations (default: 1000)
- **init** (*str* or *pyttb.ktensor*) –
Initial guess (default: “random”)
 - “random”: initialize using a *pyttb.ktensor* with values chosen from a Normal distribution with mean 0 and standard deviation 1
 - “nvecs”: initialize factor matrices of a *pyttb.ktensor* using the eigenvectors of the outer product of the matricized input tensor
 - *pyttb.ktensor*: initialize using a specific *pyttb.ktensor* as input - must be the same shape as the input tensor and have the same rank as the input rank
- **printitn** (*int*) – Number of iterations to perform before printing iteration status - 0 for no status printing (default: 1)
- **fixsigns** (*bool*) – Align the signs of the columns of the factorization to align with the input tensor data (default: True)

Returns

- **M** (*pyttb.ktensor*) – Resulting ktensor from CP-ALS factorization
- **Minit** (*pyttb.ktensor*) – Initial guess
- **output** (*dict*) –
Information about the computation. Dictionary keys:
 - *params* : tuple of (stoptol, maxiters, printitn, dimorder)
 - *iters*: number of iterations performed
 - *normresidual*: norm of the difference between the input tensor and ktensor factorization

– *fit*: value of the fitness function (fraction of tensor data explained by the model)

Example

Random initialization causes slight pertubation in intermediate results. ... is our place holder for these numeric values. Example using default values (“random” initialization):

```
>>> weights = np.array([1., 2.])
>>> fm0 = np.array([[1., 2.], [3., 4.]])
>>> fm1 = np.array([[5., 6.], [7., 8.]])
>>> K = ttb.ktensor.from_data(weights, [fm0, fm1])
>>> np.random.seed(1)
>>> M, Minit, output = ttb.cp_als(K.full(), 2)
CP_ALS:
  Iter 0: f = ... f-delta = ...
  Iter 1: f = ... f-delta = ...
  Final f = ...
>>> print(M)
ktensor of shape 2 x 2
weights=[108.4715... 8.6114...]
factor_matrices[0] =
[[0.4187... 0.3989...]
 [0.9080... 0.9169...]]
factor_matrices[1] =
[[0.6188... 0.2581...]
 [0.7854... 0.9661...]]
>>> print(Minit)
ktensor of shape 2 x 2
weights=[1. 1.]
factor_matrices[0] =
[[4.1702...e-01 7.2032...e-01]
 [1.1437...e-04 3.0233...e-01]]
factor_matrices[1] =
[[0.1467... 0.0923...]
 [0.1862... 0.3455...]]
>>> print(output)
{'params': (0.0001, 1000, 1, [0, 1]), 'iters': 1, 'normresidual': ..., 'fit': ...}
```

Example using “nvecs” initialization:

```
>>> M, Minit, output = ttb.cp_als(K.full(), 2, init="nvecs")
CP_ALS:
  Iter 0: f = ... f-delta = ...
  Iter 1: f = ... f-delta = ...
  Final f = ...
```

Example using `pyttb.ktensor` initialization:

```
>>> M, Minit, output = ttb.cp_als(K.full(), 2, init=K)
CP_ALS:
  Iter 0: f = ... f-delta = ...
  Iter 1: f = ... f-delta = ...
  Final f = ...
```

pyttb.cp_apr

```
pyttb.cp_apr.cp_apr(input_tensor, rank, algorithm='mu', stoptol=0.0001, stoptime=1000000.0, maxiters=1000,
                    init='random', maxinneriters=10, epsDivZero=1e-10, printitn=1, printinneritn=0,
                    kappa=0.01, kappatol=1e-10, epsActive=1e-08, mu0=1e-05, precompinds=True,
                    inexact=True, lbfgsMem=3)
```

Compute non-negative CP with alternating Poisson regression.

Parameters

- **input_tensor** (*pyttb.tensor* or *pyttb.sptensor*)
- **rank** (*int*) – Rank of the decomposition
- **algorithm** (*str*) – in { ‘mu’, ‘pdnr’, ‘pqr’ }
- **stoptol** (*float*) – Tolerance on overall KKT violation
- **stoptime** (*float*) – Maximum number of seconds to run
- **maxiters** (*int*) – Maximum number of iterations
- **init** (*str* or *pyttb.ktensor*) – Initial guess
- **maxinneriters** (*int*) – Maximum inner iterations per outer iteration
- **epsDivZero** (*float*) – Safeguard against divide by zero
- **printitn** (*int*) – Print every n outer iterations, 0 for none
- **printinneritn** (*int*) – Print every n inner iterations
- **kappa** (*int*) – MU ALGORITHM PARAMETER: Offset to fix complementary slackness
- **kappatol** – MU ALGORITHM PARAMETER: Tolerance on complementary slackness
- **epsActive** (*float*) – PDNR & PQNR ALGORITHM PARAMETER: Bertsekas tolerance for active set
- **mu0** (*float*) – PDNR ALGORITHM PARAMETER: Initial Damping Parameter
- **precompinds** (*bool*) – PDNR & PQNR ALGORITHM PARAMETER: Precompute sparse tensor indices
- **inexact** (*bool*) – PDNR ALGORITHM PARAMETER: Compute inexact Newton steps
- **lbfgsMem** (*int*) – PQNR ALGORITHM PARAMETER: Precompute sparse tensor indices

Returns

- **M** (*pyttb.ktensor*) – Resulting ktensor from CP APR
- **Minit** (*pyttb.ktensor*) – Initial Guess
- **output** (*dict*) – Additional output #TODO document this more appropriately

```
pyttb.cp_apr.tt_cp_apr_mu(input_tensor, rank, init, stoptol, stoptime, maxiters, maxinneriters, epsDivZero,
                          printitn, printinneritn, kappa, kappatol)
```

Compute nonnegative CP with alternating Poisson regression.

Parameters

- **input_tensor** (*pyttb.tensor* or *pyttb.sptensor*)
- **rank** (*int*) – Rank of the decomposition
- **init** (*pyttb.ktensor*) – Initial guess

- **stoptol** (*float*) – Tolerance on overall KKT violation
- **stoptime** (*float*) – Maximum number of seconds to run
- **maxiters** (*int*) – Maximum number of iterations
- **maxinneriters** (*int*) – Maximum inner iterations per outer iteration
- **epsDivZero** (*float*) – Safeguard against divide by zero
- **printitn** (*int*) – Print every n outer iterations, 0 for none
- **printinneritn** (*int*) – Print every n inner iterations
- **kappa** (*int*) – MU ALGORITHM PARAMETER: Offset to fix complementary slackness
- **kappatol** – MU ALGORITHM PARAMETER: Tolerance on complementary slackness

Notes

REFERENCE: E. C. Chi and T. G. Kolda. On Tensors, Sparsity, and Nonnegative Factorizations, arXiv:1112.2414 [math.NA], December 2011, URL: <http://arxiv.org/abs/1112.2414>. Submitted for publication.

`pyttb.cp_apr.tt_cp_apr_pdnr(input_tensor, rank, init, stoptol, stoptime, maxiters, maxinneriters, epsDivZero, printitn, printinneritn, epsActive, mu0, precompinds, inexact)`

Compute nonnegative CP with alternating Poisson regression computes an estimate of the best rank-R CP model of a tensor X using an alternating Poisson regression. The algorithm solves “row subproblems” in each alternating subproblem, using a Hessian of size R^2 .

Parameters

- **# TODO it looks like this method of define union helps the typ hinting better than or**
- **input_tensor** (Union[`pyttb.tensor`, `class:pyttb.sptensor`])
- **rank** (*int*) – Rank of the decomposition
- **init** (str or `pyttb.ktensor`) – Initial guess
- **stoptol** (*float*) – Tolerance on overall KKT violation
- **stoptime** (*float*) – Maximum number of seconds to run
- **maxiters** (*int*) – Maximum number of iterations
- **maxinneriters** (*int*) – Maximum inner iterations per outer iteration
- **epsDivZero** (*float*) – Safeguard against divide by zero
- **printitn** (*int*) – Print every n outer iterations, 0 for none
- **printinneritn** (*int*) – Print every n inner iterations
- **epsActive** (*float*) – PDNR & PQNR ALGORITHM PARAMETER: Bertsekas tolerance for active set
- **mu0** (*float*) – PDNR ALGORITHM PARAMETER: Initial Damping Parameter
- **precompinds** (*bool*) – PDNR & PQNR ALGORITHM PARAMETER: Precompute sparse tensor indices
- **inexact** (*bool*) – PDNR ALGORITHM PARAMETER: Compute inexact Newton steps

Returns

TODO detail return dictionary

Notes

REFERENCE: Samantha Hansen, Todd Plantenga, Tamara G. Kolda. Newton-Based Optimization for Non-negative Tensor Factorizations, arXiv:1304.4964 [math.NA], April 2013, URL: <http://arxiv.org/abs/1304.4964>. Submitted for publication.

`pyttb.cp_apr.tt_cp_apr_pqnr(input_tensor, rank, init, stoptol, stoptime, maxiters, maxinneriters, epsDivZero, printitn, printinneritn, epsActive, lbfgsMem, precompinds)`

Compute nonnegative CP with alternating Poisson regression.

`tt_cp_apr_pdnr` computes an estimate of the best rank-R CP model of a tensor X using an alternating Poisson regression. The algorithm solves “row subproblems” in each alternating subproblem, using a Hessian of size R^2 . The function is typically called by `cp_apr`.

The model is solved by nonlinear optimization, and the code literally minimizes the negative of log-likelihood. However, printouts to the console reverse the sign to show maximization of log-likelihood.

mu0: float

PDNR ALGORITHM PARAMETER: Initial Damping Parameter

precompinds: bool

PDNR & PQNR ALGORITHM PARAMETER: Precompute sparse tensor indices

inexact: bool

PDNR ALGORITHM PARAMETER: Compute inexact Newton steps

Parameters

- **input_tensor** (Union[`pyttb.tensor`, :class:pyttb.sptensor])
- **rank** (*int*) – Rank of the decomposition
- **init** (str or `pyttb.ktensor`) – Initial guess
- **stoptol** (*float*) – Tolerance on overall KKT violation
- **stoptime** (*float*) – Maximum number of seconds to run
- **maxiters** (*int*) – Maximum number of iterations
- **maxinneriters** (*int*) – Maximum inner iterations per outer iteration
- **epsDivZero** (*float*) – Safeguard against divide by zero
- **printitn** (*int*) – Print every n outer iterations, 0 for none
- **printinneritn** (*int*) – Print every n inner iterations
- **epsActive** (*float*) – PDNR & PQNR ALGORITHM PARAMETER: Bertsekas tolerance for active set
- **lbfgsMem** (*int*) – Number of vector pairs to store for L-BFGS
- **precompinds**

Returns

TODO detail return dictionary

Notes

REFERENCE: Samantha Hansen, Todd Plantenga, Tamara G. Kolda. Newton-Based Optimization for Non-negative Tensor Factorizations, arXiv:1304.4964 [math.NA], April 2013, URL: <http://arxiv.org/abs/1304.4964>. Submitted for publication.

`pyttb.cp_apr.tt_calcp_i_prowsubprob(Data, Model, rank, factorIndex, ndims, isSparse=False, sparse_indices=None)`

Compute Pi for a row subproblem.

Parameters

- **Data** :class:`pyttb.sptensor` or :class:`pyttb.tensor`
- **isSparse** (*bool*)
- **Model** :class:`pyttb.ktensor`
- **rank** (*int*)
- **factorIndex** (*int*)
- **ndims** (*int*)
- **sparse_indices** (*list*) – Indices of row subproblem nonzero elements

Returns

Pi (`numpy.ndarray`)

See also:

`pyttb.calculatePi`

`pyttb.cp_apr.calc_partials(isSparse, Pi, epsilon, data_row, model_row)`

Compute derivative quantities for a PDNR row subproblem.

Parameters

- **isSparse** (*bool*)
- **Pi** (`numpy.ndarray`)
- **epsilon** (*float*) – Prevent division by zero
- **data_row** (`numpy.ndarray`)
- **model_row** (`numpy.ndarray`)

Returns

- **phi_row** (`numpy.ndarray`) – gradient of row subproblem, except for a constant

$$\phi_{row}[r] = \sum_{j=1}^{J_n} \frac{x_j \pi_{rj}}{\sum_i^R b_i \pi_{ij}}$$
- **ups_row** (`numpy.ndarray`) – intermediate quantity (upsilon) used for second derivatives

$$\upsilon_{row}[j] = \frac{x_j}{(\sum_i^R b_i \pi_{ij})^2}$$

`pyttb.cp_apr.getSearchDirPdnr(Pi, ups_row, rank, gradModel, model_row, mu, epsActSet)`

Compute the search direction for PDNR using a two-metric projection with damped Hessian

Parameters

- **Pi** (`numpy.ndarray`)
- **ups_row** (`numpy.ndarray`) – intermediate quantity (upsilon) used for second derivatives

- **rank** (*int*) – number of variables for the row subproblem
- **gradModel** (`numpy.ndarray`) – gradient vector for the row subproblem
- **model_row** (`numpy.ndarray`) – vector of variables for the row subproblem
- **mu** (*float*) – damping parameter
- **epsActSet** (*float*) – Bertsekas tolerance for active set determination

Returns

- **search_dir** (`numpy.ndarray`) – search direction vector
- **pred_red** (`numpy.ndarray`) – predicted reduction in quadratic model

`pyttb.cp_apr.tt_linesearch_prowsubprob(direction, grad, model_old, step_len, step_red, max_steps, suff_decr, isSparse, data_row, Pi, phi_row, display_warning)`

Perform a line search on a row subproblem

Parameters

- **direction** (`numpy.ndarray`) – search direction
- **grad** (`numpy.ndarray`) – gradient vector a model_old
- **model_old** (`numpy.ndarray`) – current variable values
- **step_len** (*float*) – initial step length, which is the maximum possible step length
- **step_red** (*float*) – step reduction factor (suggest 1/2)
- **max_steps** (*int*) – maximum number of steps to try (suggest 10)
- **suff_decr** (*float*) – sufficient decrease for convergence (suggest 1.0e-4)
- **isSparse** (*bool*) – sparsity flag for computing the objective
- **data_row** (`numpy.ndarray`) – row subproblem data, for computing the objective
- **Pi** (`numpy.ndarray`) – Pi matrix, for computing the objective
- **phi_row** (`numpy.ndarray`) – 1-grad, more accurate if failing over to multiplicative update
- **display_warning** (*bool*) – Flag to display warning messages or not

Returns

- **m_new** (`numpy.ndarray`) – new (improved) model values
- **num_evals** (*int*) – number of times objective was evaluated
- **f_old** (*float*) – objective value at model_old
- **f_1** (*float*) – objective value at model_old + step_len*direction
- **f_new** (*float*) – objective value at model_new

`pyttb.cp_apr.getHessian(upsilon, Pi, free_indices)`

Return the Hessian for one PDNR row subproblem of Model[n], for just the rows and columns corresponding to the free variables

Parameters

- **upsilon** (`numpy.ndarray`) – intermediate quantity (upsilon) used for second derivatives
- **Pi** (`numpy.ndarray`)
- **free_indices** (*list*)

Returns

Hessian (`numpy.ndarray`) – Sub-block of full Hessian identified by free-indices

`pyttb.cp_apr.tt_loglikelihood_row(isSparse, data_row, model_row, Pi)`

Compute log-likelihood of one row subproblem

Parameters

- **isSparse** (*bool*) – Sparsity flag
- **data_row** (`numpy.ndarray`) – vector of data values
- **model_row** (`numpy.ndarray`) – vector of model values
- **Pi** (`numpy.ndarray`)

Notes

The row subproblem for a given mode includes one row of matricized tensor data (x) and one row of the model (m) in the same matricized mode. Then (dense case) m : R -length vector x : J -length vector P_i : $R \times J$ matrix (sparse case) m : R -length vector x : p -length vector, where $p = \text{nnz}$ in row of matricized data tensor P_i : $R \times p$ matrix $F = -(\sum_r m_r - \sum_j x_j * \log(m * P_{i,j}))$ where $P_{i,j}$ denotes the j^{th} column of P_i NOTE: Rows of P_i must sum to one

Returns

loglikelihood (*float*) – See notes for description

`pyttb.cp_apr.getSearchDirPqnr(model_row, gradModel, epsActSet, delta_model, delta_grad, rho, lbfgs_pos, iters, disp_warn)`

Compute the search direction by projecting with L-BFGS.

Parameters

- **model_row** (`numpy.ndarray`) – current variable values
- **gradModel** (`numpy.ndarray`) – gradient at model_row
- **epsActSet** (*float*) – Bertsekas tolerance for active set determination
- **delta_model** (`numpy.ndarray`) – L-BFGS array of variable deltas
- **delta_grad** (`numpy.ndarray`) – L-BFGS array of gradient deltas
- **rho**
- **lbfgs_pos** (*int*) – pointer into L-BFGS arrays
- **iters**
- **disp_warn** (*bool*)

Returns

direction (`numpy.ndarray`) – Search direction based on current L-BFGS and grad

Notes

Adapted from MATLAB code of Dongmin Kim and Suvrit Sra written in 2008. Modified extensively to solve row subproblems and use a better linesearch; for details see REFERENCE: Samantha Hansen, Todd Plantenga, Tamara G. Kolda. Newton-Based Optimization for Nonnegative Tensor Factorizations, arXiv:1304.4964 [math.NA], April 2013, URL: <http://arxiv.org/abs/1304.4964>. Submitted for publication.

`pyttb.cp_apr.calc_grad(isSparse, Pi, eps_div_zero, data_row, model_row)`

Compute the gradient for a PQNR row subproblem

Parameters

- **isSparse** (*bool*)
- **Pi** (`numpy.ndarray`)
- **eps_div_zero** (*float*)
- **data_row** (`numpy.ndarray`)
- **model_row** (`numpy.ndarray`)

Returns

- **phi_row** (`numpy.ndarray`)
- **grad_row** (`numpy.ndarray`)

`pyttb.cp_apr.calculatePi(Data, Model, rank, factorIndex, ndims)`

Helper function to calculate Pi matrix # TODO verify what pi is

Parameters

- **Data** (`pyttb.sptensor` or `pyttb.tensor`)
- **Model** (`pyttb.ktensor`)
- **rank** (*int*)
- **factorIndex** (*int*)
- **ndims** (*int*)

Returns

Pi (`numpy.ndarray`)

`pyttb.cp_apr.calculatePhi(Data, Model, rank, factorIndex, Pi, epsilon)`

Parameters

- **Data** (`pyttb.sptensor` or `pyttb.tensor`)
- **Model** (`pyttb.ktensor`)
- **rank** (*int*)
- **factorIndex** (*int*)
- **Pi** (`numpy.ndarray`)
- **epsilon** (*float*)

`pyttb.cp_apr.tt_loglikelihood(Data, Model)`

Compute log-likelihood of data with model.

Parameters

- **Data** (*pyttb.sptensor* or *pyttb.tensor*)
- **Model** (*pyttb.ktensor*)

Returns

loglikelihood (*float*) –

- $(\sum_i m_i - x_i * \log_i)$ where *i* is a multiindex across all tensor dimensions.

Notes

We define for any $x \geq 0, \log(x)=0$, such that if our true data is 0 the loglikelihood is the value of the model.

`pyttb.cp_apr.vectorizeForMu(matrix)`

Helper Function to unravel matrix into vector

Parameters

matrix (*numpy.ndarray*)

Returns

matrix (*numpy.ndarray*) – `len(matrix.shape)==1`

pyttb.hosvd

Higher Order SVD Implementation

`pyttb.hosvd.hosvd(input_tensor, tol: float, verbosity: float = 1, dimorder: Optional[List[int]] = None, sequential: bool = True, ranks: Optional[List[int]] = None)`

Compute sequentially-truncated higher-order SVD (Tucker).

Computes a Tucker decomposition with relative error specified by *tol*, i.e., it computes a tensor *T* such that $\|X-T\|/\|X\| \leq \text{tol}$.

Parameters

- **input_tensor** (*Tensor to factor*)
- **tol** (*Relative error to stop at*)
- **verbosity** (*Print level*)
- **dimorder** (*Order to loop through dimensions*)
- **sequential** (*Use sequentially-truncated version*)
- **ranks** (*Specify ranks to consider rather than computing*)

Example

```
>>> data = np.array([[29, 39.], [63., 85.]])
>>> tol = 1e-4
>>> disable_printing = -1
>>> tensorInstance = ttb.tensor().from_data(data)
>>> result = hosvd(tensorInstance, tol, verbosity=disable_printing)
>>> ((result.full() - tensorInstance).norm() / tensorInstance.norm()) < tol
True
```

pyttb.tucker_als

```
pyttb.tucker_als.tucker_als(input_tensor, rank, stoptol=0.0001, maxiters=1000, dimorder=None,
                             init='random', printitn=1)
```

Compute Tucker decomposition with alternating least squares

Parameters

- **input_tensor** (*pyttb.tensor*)
- **rank** (*int, list[int]*) – Rank of the decomposition(s)
- **stoptol** (*float*) – Tolerance used for termination - when the change in the fitness function in successive iterations drops below this value, the iterations terminate (default: 1e-4)
- **dimorder** (*list*) – Order to loop through dimensions (default: [range(tensor.ndims)])
- **maxiters** (*int*) – Maximum number of iterations (default: 1000)
- **init** (*str or list[np.ndarray]*) –
Initial guess (default: “random”)
 - “random”: initialize using a *pyttb.tensor* with values chosen from a Normal distribution with mean 1 and standard deviation 0
 - “nvecs”: initialize factor matrices of a *pyttb.tensor* using the eigenvectors of the outer product of the matricized input tensor
 - *pyttb.tensor*: initialize using a specific *pyttb.tensor* as input - must be the same shape as the input tensor and have the same rank as the input rank
- **printitn** (*int*) – Number of iterations to perform before printing iteration status - 0 for no status printing (default: 1)

Returns

- **M** (*pyttb.tensor*) – Resulting ttensor from Tucker-ALS factorization
- **Minit** (*pyttb.tensor*) – Initial guess
- **output** (*dict*) –
Information about the computation. Dictionary keys:
 - *params* : tuple of (stoptol, maxiters, printitn, dimorder)
 - *iters*: number of iterations performed
 - *normresidual*: norm of the difference between the input tensor and ktensor factorization
 - *fit*: value of the fitness function (fraction of tensor data explained by the model)

HOW TO CITE

Please see [references](#) for how to cite a variety of algorithms implemented in this project.

3.1 BibTex Entries: Tensor Toolbox for Python

```
@misc{TTB_Software,
  author = {Brett W. Bader and Tamara G. Kolda and others},
  title = {MATLAB Tensor Toolbox Version 3.0-dev},
  howpublished = {Available online},
  month = oct,
  year = {2017},
  url = {https://www.tensortoolbox.org}
}
@article{TTB_Dense,
  author = {Brett W. Bader and Tamara G. Kolda},
  title = {Algorithm 862: {MATLAB} tensor classes for fast algorithm prototyping},
  journal = {ACM Transactions on Mathematical Software},
  month = dec,
  year = {2006},
  volume = {32},
  number = {4},
  pages = {635-653},
  doi = {10.1145/1186785.1186794}
}
@article{TTB_Sparse,
  author = {Brett W. Bader and Tamara G. Kolda},
  title = {Efficient {MATLAB} computations with sparse and factored tensors},
  journal = {SIAM Journal on Scientific Computing},
  month = dec,
  year = {2007},
  volume = {30},
  number = {1},
  pages = {205-231},
  doi = {10.1137/060676489}
}
@article{TTB_CPOPT,
  author = {Evrin Acar and Daniel M. Dunlavy and Tamara G. Kolda},
  title = {A Scalable Optimization Approach for Fitting Canonical Tensor Decompositions},
  journal = {Journal of Chemometrics},
```

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month = feb,
year = {2011},
volume = {25},
number = {2},
pages = {67-86},
doi = {10.1002/cem.1335}
}
@Article{TTB_CPWOPT,
  author = {Evrin Acar and Daniel M. Dunlavy and Tamara G. Kolda and Morten M{\o}rup},
  title = {Scalable Tensor Factorizations for Incomplete Data},
  journal = {Chemometrics and Intelligent Laboratory Systems},
  month = mar,
  year = {2011},
  volume = {106},
  number = {1},
  pages = {41-56},
  doi = {10.1016/j.chemolab.2010.08.004}
}
@Article{TTB_SSHOPM,
  author = {Tamara G. Kolda and Jackson R. Mayo},
  title = {Shifted Power Method for Computing Tensor Eigenpairs},
  journal = {SIAM Journal on Matrix Analysis and Applications},
  month = oct,
  year = {2011},
  volume = {32},
  number = {4},
  pages = {1095-1124},
  doi = {10.1137/100801482}
}
@Article{TTB_EIGGEAP,
  title = {An Adaptive Shifted Power Method for Computing Generalized Tensor Eigenpairs},
  author = {Tamara G. Kolda and Jackson R. Mayo},
  doi = {10.1137/140951758},
  journal = {SIAM Journal on Matrix Analysis and Applications},
  number = {4},
  volume = {35},
  year = {2014},
  month = dec,
  pages = {1563-1581},
}
@Article{TTB_CPAPR,
  title = {On Tensors, Sparsity, and Nonnegative Factorizations},
  author = {Eric C. Chi and Tamara G. Kolda},
  doi = {10.1137/110859063},
  journal = {SIAM Journal on Matrix Analysis and Applications},
  number = {4},
  volume = {33},
  year = {2012},
  month = dec,
  pages = {1272-1299},
}
@Article{TTB_CPAPRB,

```

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    author = {Samantha Hansen and Todd Plantenga and Tamara G. Kolda},
    title = {Newton-Based Optimization for {Kullback-Leibler} Nonnegative Tensor_
↪Factorizations},
    journal = {Optimization Methods and Software},
    volume = {30},
    number = {5},
    pages = {1002-1029},
    month = {April},
    year = {2015},
    doi = {10.1080/10556788.2015.1009977},
}
@article{TTB_CPSYM,
    author = {Tamara G. Kolda},
    title = {Numerical Optimization for Symmetric Tensor Decomposition},
    journal = {Mathematical Programming B},
    volume = {151},
    number = {1},
    pages = {225-248},
    month = apr,
    year = {2015},
    doi = {10.1007/s10107-015-0895-0},
}
@misc{TTB_CPRALS,
    author = {Casey Battaglino and Grey Ballard and Tamara G. Kolda},
    title = {A Practical Randomized {CP} Tensor Decomposition},
    howpublished = {arXiv:1701.06600},
    month = jan,
    year = {2017},
    eprint = {1701.06600},
    eprintclass = {cs.NA},
}
@inproceedings{TTB_MET,
    author = {Tamara G. Kolda and Jimeng Sun},
    title = {Scalable Tensor Decompositions for Multi-aspect Data Mining},
    booktitle = {ICDM 2008: Proceedings of the 8th IEEE International Conference on Data_
↪Mining},
    month = dec,
    year = {2008},
    pages = {363-372},
    doi = {10.1109/ICDM.2008.89}
}

```


CONTACT

Please email dmdunla@sandia.gov with any questions about pyttb that cannot be resolved via issue reporting. Stories of its usefulness are especially welcome. We will try to respond to every email may not always be successful due to the volume of emails.

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