Package 'TDbasedUFEadv'

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Type Package

Title Advanced package of tensor decomposition based unsupervised feature extraction

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Description This is an advanced version of TDbasedUFE, which is a comprehensive package to perform Tensor decomposition based unsupervised feature extraction. In contrast to TDbasedUFE which can perform simple the feature selection and the multiomics analyses, this package can perform more complicated and advanced features, but they are not so popularly required. Only users who require more specific features can make use of its functionality.

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`TDbasedUFEadv-package` *TDbasedUFEadv: Advanced package of tensor decomposition based unsupervised feature extraction*

Description

This is an advanced version of TDbasedUFE, which is a comprehensive package to perform Tensor decomposition based unsupervised feature extraction. In contrast to TDbasedUFE which can perform simple the feature selection and the multiomics analyses, this package can perform more complicated and advanced features, but they are not so popularly required. Only users who require more specific features can make use of its functionality.

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See Also

Useful links:

- https://github.com/tagtag/TDbasedUFEadv
- Report bugs at https://github.com/tagtag/TDbasedUFEadv/issues

`computeSVD` *Title Perform SVD toward reduced matrix generated from a tensor with partial summation*

Description

Title Perform SVD toward reduced matrix generated from a tensor with partial summation

Usage

```
computeSVD(matrix1, matrix2, dim = 10L, scale = TRUE)
```

Arguments

`matrix1`	The first original matrix that generates a tensor
`matrix2`	The second original matrix that generates a tensor
`dim`	The number of singular value vectors to be computed
`scale`	If matrix should be scaled or not

Value

Singular value vectors attributed to two sets of objects associated with singular value vectors attributed to features, by multiplying

Examples

```
matrix1 <- matrix(runif(200),20)
matrix2 <- matrix(runif(400),20)
SVD <- computeSVD(matrix1,matrix2)
```

`prepareCondDrugandDisease` *Prepare condition matrix for expDrug*

Description

Prepare condition matrix for expDrug

Usage

```
prepareCondDrugandDisease(expDrug)
```

Arguments

`expDrug`	input gene expression profile

Value

Condition matrix for expDrug

Examples

```
library(RTCGA.rnaseq)
Cancer_cell_lines <- list(ACC.rnaseq,BLCA.rnaseq,BRCA.rnaseq)
Drug_and_Disease <- prepareexpDrugandDisease(Cancer_cell_lines)
Cond <- prepareCondDrugandDisease(Drug_and_Disease$expDrug)
```

---

prepareCondTCGA *Prepare Sample label for TCGA data*

Description

Prepare Sample label for TCGA data

Usage

```
prepareCondTCGA(
  Multi_sample,
  Clinical,
  ID_column_of_Multi_sample,
  ID_column_of_Clinical
)
```

Arguments

Multi_sample	list of sample ids
Clinical	List of clinical data matrix from RTCGA.clinical
ID_column_of_Multi_sample	Column numbers used for conditions
ID_column_of_Clinical	Column numbers that include corresponding sample ids in clinical data

Value

list of sample labels

Examples

```
library(RTCGA.clinical)
library(RTCGA.rnaseq)
Clinical <- list(BLCA.clinical, BRCA.clinical, CESC.clinical, COAD.clinical)
Multi_sample <- list(
  BLCA.rnaseq[seq_len(100), 1, drop = FALSE],
  BRCA.rnaseq[seq_len(100), 1, drop = FALSE],
```

```
  CESC.rnaseq[seq_len(100), 1, drop = FALSE],
  COAD.rnaseq[seq_len(100), 1, drop = FALSE]
)
ID_column_of_Multi_sample <- c(770, 1482, 773, 791)
ID_column_of_Clinical <- c(20, 20, 12, 14)
cond <- prepareCondTCGA(
  Multi_sample, Clinical,
  ID_column_of_Multi_sample, ID_column_of_Clinical
)
```

`prepareexpDrugandDisease` *Generating gene expression of drug treated cell lines and a disease cell line*

Description

Generating gene expression of drug treated cell lines and a disease cell line

Usage

```
prepareexpDrugandDisease(Cancer_cell_lines)
```

Arguments

`Cancer_cell_lines` <- list(ACC.rnaseq,BLCA.rnaseq,BRCA.rnaseq) list that includes individual data set from RTCGA.rnaseq

Value

list of expDrug and expDisease

Examples

```
library(RTCGA.rnaseq)
Cancer_cell_lines <- list(ACC.rnaseq,BLCA.rnaseq,BRCA.rnaseq)
Drug_and_Disease <- prepareexpDrugandDisease(Cancer_cell_lines)
```

prepareTensorfromList *Prepare tensor from a list that includes multiple profiles*

Description

Prepare tensor from a list that includes multiple profiles

Usage

```
prepareTensorfromList(Multi, proj_dim)
```

Arguments

Multi	a list that includes multiple profiles
proj_dim	the number of projection dimensions

Value

a tensor as a bundle of singular value vectors obtained by applying SVD to individual omics

Examples

```
library(MOFAdata)
data("CLL_data")
data("CLL_covariates")
Z <- prepareTensorfromList(CLL_data,10L)
```

prepareTensorfromMatrix *Generate tensor from two matrices*

Description

Generate tensor from two matrices

Usage

```
prepareTensorfromMatrix(matrix1, matrix2)
```

Arguments

matrix1	the first input matrix
matrix2	the second input matrix

Value

A tensor generated from the first and second matrices

Examples

```
Z <- prepareTensorfromMatrix(matrix(runif(100),10),matrix(runif(100),10))
```

`prepareTensorRect` *Prepare tensor generated from two matrices that share samples*

Description

Prepare tensor generated from two matrices that share samples

Usage

```
prepareTensorRect(
  sample,
  feature,
  value,
  featureRange = GRanges(NULL),
  sampleData = list(NULL)
)
```

Arguments

`sample`	Character vector of sample names
`feature`	list of features from two matrices
`value`	array, contents of
`featureRange`	Genomic Ranges to be associated with features
`sampleData`	List of conditional labeling associated with samples

Value

Tensor generated from two matrices that share samples

Examples

```
matrix1 <- matrix(runif(1000),200) #row features, column samples
matrix2 <- matrix(runif(2000),400) #row features, column samples
Z <- prepareTensorfromMatrix(t(matrix1),t(matrix2))
Z <- prepareTensorRect(sample=as.character(seq_len(50)),
feature=list(as.character(seq_len(200)),as.character(seq_len(400))),
sampleData=list(rep(seq_len(2),each=25)),value=Z)
```

`selectFeatureProj` *Select feature when projection strategy is employed for the case where features are shared with multiple omics profiles*

Description

Select feature when projection strategy is employed for the case where features are shared with multiple omics profiles

Usage

```
selectFeatureProj(
  HOSVD,
  Multi,
  cond,
  de = 1e-04,
  p0 = 0.01,
  breaks = 100L,
  input_all = NULL
)
```

Arguments

`HOSVD`	HOSVD
`Multi`	list of omics profiles, row: sample, column: feature
`cond`	list of conditions for individual omics profiles
`de`	initial value for optimization of standard deviation
`p0`	Threshold P-value
`breaks`	The number of bins of histogram of P-values
`input_all`	The number of selected feature. if null, interactive mode is activated

Value

list composed of logical vector that represent which features are selected and p-values

Examples

```
library(TDbasedUFE)
Multi <- list(matrix(runif(1000),10),matrix(runif(1000),10),
matrix(runif(1000),10),matrix(runif(1000),10))
Z <- prepareTensorfromList(Multi,10L)
Z <- aperm(Z,c(2,1,3))
Z <- PrepareSummarizedExperimentTensor(feature =as.character(1:10),
                                       sample=array("",1),value=Z)
HOSVD <- computeHosvd(Z)
cond <- rep(list(rep(1:2,each=5)),4)
index <- selectFeatureProj(HOSVD,Multi,cond,de=0.1,input_all=2)
```

`selectFeatureRect` *Select features through the selection of singular value vectors*

Description

Select features through the selection of singular value vectors

Usage

```
selectFeatureRect(
  SVD,
  cond,
  de = rep(1e-04, 2),
  p0 = 0.01,
  breaks = 100L,
  input_all = NULL
)
```

Arguments

`SVD`	SVD computed from matrix generated by partial summation of a tensor
`cond`	Condition to select singular value vectors
`de`	Initial values to be used for optimization of standard deviation
`p0`	Threshold value for the significance
`breaks`	Number of bins of histogram of P-values
`input_all`	The ID of selected singular value vectors. If it is null, interactive mode is activated.

Value

List of lists that includes P-vales as well as if individual features selected.

Examples

```
set.seed(0)
matrix1 <- matrix(runif(2000),200)
matrix2 <- matrix(runif(4000),200)
SVD <- computeSVD(matrix1,matrix2)
index_all <- selectFeatureRect(SVD,
list(NULL,rep(seq_len(2),each=5),rep(seq_len(2),each=10)),de=rep(0.5,2),
input_all=1)
```

`selectFeatureTransRect` *Select features for a tensor generated from two matrices that share samples.*

Description

Select features for a tensor generated from two matrices that share samples.

Usage

```
selectFeatureTransRect(
  HOSVD,
  cond,
  de = rep(1e-04, 2),
  p0 = 0.01,
  breaks = 100L,
  input_all = NULL
)
```

Arguments

HOSVD	HOSVD
cond	list of conditions
de	initial values for optimization of standard deviation
p0	threshold value for the significance
breaks	number of bins of the histogram of P-values
input_all	The selected singular value vectors attributed to samples. if NULL, interactive mode

Value

list of logical vector that represent if the individual features are selected and P-values.

Examples

```
library(TDbasedUFE)
set.seed(0)
matrix1 <- matrix(runif(1000),20) #row features, column samples
matrix2 <- matrix(runif(2000),40) #row features, column samples
Z <- prepareTensorfromMatrix(t(matrix1),t(matrix2))
Z <- prepareTensorRect(sample=as.character(seq_len(50)),
feature=list(as.character(seq_len(20)),as.character(seq_len(40))),
sampleData=list(rep(seq_len(2),each=25)),value=Z)
HOSVD <- computeHosvd(Z)
cond <- list(attr(Z,"sampleData")[[1]],NULL,NULL)
index_all <- selectFeatureTransRect(HOSVD,cond,de=c(0.1,0.1),
input_all=2,p0=1e-10)
```

TensorRect-class *Class definitions*

Description

Class definitions

Slots

`sample` character.

`feature` list.

`value` array.

`featureRange` GRanges.

`sampleData` list.

transSVD *Convert SVD to that for the case where samples are shared between two matrices*

Description

Convert SVD to that for the case where samples are shared between two matrices

Usage

```
transSVD(SVD)
```

Arguments

`SVD`	input SVD object generated from computeSVD function

Value

converted SVD objects

Examples

```
matrix1 <- matrix(runif(200),20)
matrix2 <- matrix(runif(400),20)
SVD <- computeSVD(matrix1,matrix2)
SVD <- transSVD(SVD)
```

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