# Package 'timeOmics'

April 18, 2025

Title Time-Course Multi-Omics data integration

**Version** 1.21.0

**Description** timeOmics is a generic data-driven framework to integrate multi-Omics longitudinal data measured on the same biological samples and select key temporal features with strong associations within the same sample group. The main steps of timeOmics are:

- 1. Plaform and time-specific normalization and filtering steps;
- 2. Modelling each biological into one time expression profile;
- 3. Clustering features with the same expression profile over time;
- 4. Post-hoc validation step.

**License** GPL-3 **Encoding** UTF-8

LazyData true

**Imports** dplyr, tidyr, tibble, purrr, magrittr, ggplot2, stringr, ggrepel, lmtest, plyr, checkmate

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 ${\it dmatrix.spearman.dissimilarity} \\ {\it dmatrix.spearman.dissimilarity}$ 

# Description

Compute the spearman dissimilarity distance.

# Usage

dmatrix.spearman.dissimilarity(X)

# Arguments

X A numeric matrix with feature in colnames

# Value

Return a dissimilarity matrix of size PxP.

getCluster 3

# **Description**

This function returns the cluster associated to each feature from a mixOmics object.

# Usage

```
getCluster(X, user.block = NULL, user.cluster = NULL)
```

#### **Arguments**

```
X an object of the class: pca, spca, pls, spls, block.pls or block.spls
user.block a vector to filter the result and return the features of the specified blocks.
user.cluster a vector to filter the result and return only the features of the specified clusters
```

#### **Details**

For each feature, the cluster is assigned according to the maximum contribution on a component and the sign of that contribution.

#### Value

A data frame containing the name of feature, its assigned cluster and other information such as selected component, contribution, sign, ...

#### See Also

```
selectVar
```

```
demo <- suppressWarnings(get_demo_cluster())
pca.cluster <- getCluster(demo$pca)
spca.cluster <- getCluster(demo$spca)
pls.cluster <- getCluster(demo$pls)
spls.cluster <- getCluster(demo$spls)
block.pls.cluster <- getCluster(demo$block.pls)
block.spls.cluster <- getCluster(demo$block.spls)</pre>
```

getNcomp

getNcomp	Get optimal number of components	
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# **Description**

Compute the average silhouette coefficient for a given set of components on a mixOmics result. Foreach given ncomp, the mixOmics method is performed with the sames arguments and the given 'ncomp'. Longitudinal clustering is performed and average silhouette coefficient is computed.

#### Usage

```
getNcomp(object, max.ncomp = NULL, X, Y = NULL, indY = NULL, ...)
```

# **Arguments**

object	A mixOmics object of the class 'pca', 'spca', 'mixo_pls', 'mixo_spls', 'block.pls', 'block.spls'
max.ncomp	integer, maximum number of component to include. If no argument is given, 'max.ncomp=object\$ncomp'
Χ	a numeric matrix/data.frame or a list of data.frame for block.pls
Υ	(only for pls, optional for block.spls) a numeric matrix, with the same nrow as X
indY	(optional and only for block.pls, if $Y$ is not provided), an integer which indicates the position of the matrix response in the list $X$
	Other arguments to be passed to methods (pca, pls, block.pls)

# Value

getNcomp returns a list with class "ncomp.tune.silhouette" containing the following components:

```
ncomp a vector containing the tested ncomp
silhouette a vector containing the average silhouette coefficient by ncomp
dmatrix the distance matrix used to compute silhouette coefficient
```

# See Also

```
getCluster, silhouette, pca, pls, block.pls
```

```
# random input data
demo <- suppressWarnings(get_demo_cluster())

# pca
pca.res <- mixOmics::pca(X=demo$X, ncomp = 5)
res.ncomp <- getNcomp(pca.res, max.ncomp = 4, X = demo$X)
plot(res.ncomp)

# pls
pls.res <- mixOmics::pls(X=demo$X, Y=demo$Y)</pre>
```

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getSilhouette

Get Silhouette Coefficient from (s)PCA, (s)PLS or block.(s)PLS clusters

#### **Description**

getSilhouette is a generic function that compute silhouette coefficient for an object of the type pca, spca, pls, spls, block.pls, block.spls.

# Usage

```
getSilhouette(object)
```

# **Arguments**

object

a mixOmics object of the class pca, spca, pls, spls, block.pls, block.spls

# **Details**

This method extract the componant contribution depending on the object, perform the clustering step, and compute the silhouette coefficient.

# Value

silhouette coefficient

```
demo <- suppressWarnings(get_demo_cluster())
getSilhouette(object = demo$pca)
getSilhouette(object = demo$pca)
getSilhouette(object = demo$pls)
getSilhouette(object = demo$pls)
getSilhouette(object = demo$block.pls)
getSilhouette(object = demo$block.spls)</pre>
```

get\_demo\_cluster

getUpDownCluster  $Up ext{-}D$ 

Up-Down clustering

# **Description**

Performs a clustering based on the signs of variation between 2 timepoints. Optionally, if the difference between 2 timepoints is lower than a given threshold, the returned difference will be 0.

# Usage

```
getUpDownCluster(X, diff_threshold = 0)
```

# **Arguments**

X a dataframe or list of dataframe with the same number of rows.

diff\_threshold a number (optional, default 0), if the difference between 2 values is lower than the threshold, the returned sign will be 0 (no variation).

# **Examples**

```
demo <- suppressWarnings(get_demo_cluster())
X <- list(X = demo$X, Y = demo$Y, Z = demo$Z)
res <- getUpDownCluster(X)
class(res)
getCluster(res)

X <- demo$X
res <- getUpDownCluster(X)
res <- getUpDownCluster(X, diff_threshold = 15)
res_cluster <- getCluster(res)</pre>
```

get\_demo\_cluster

get\_demo\_cluster

### **Description**

Generates random data to be used in examples.

# Usage

```
get_demo_cluster()
```

# Value

a list containg:

X data.frame
Y data.frame
Z data.frame

pca a mixOmics pca result

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```
spca a mixOmics spca result

pls a mixOmics pls result

spls a mixOmics spls result

block.pls a mixOmics block.pls result

block.spls a mixOmics block.spls result
```

#### **Examples**

# Random data could lead to "The SGCCA algorithm did not converge" warning which is not important for a demo demo <- suppressWarnings(get\_demo\_cluster())

```
get_demo_silhouette Get data for silhouette demo
```

#### **Description**

Get data for silhouette demo

#### Usage

```
get_demo_silhouette()
```

## Value

A matrix of expression profile, sample in raws, time in columns.

# Examples

```
data <- get_demo_silhouette()</pre>
```

 ${\tt lmms.filter.lines}$ 

Filter Linear Profiles from Linear Mixed Model output

# **Description**

This function filters linear models with highly heterogeneous variability within residues. From an "lmms" output, 2 parameters are tested:

# Usage

```
lmms.filter.lines(
  data,
  lmms.obj,
  time,
  homoskedasticity = TRUE,
  MSE.filter = TRUE,
  homoskedasticity.cutoff = 0.05
)
```

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# **Arguments**

data a data.frame used in the lmms::lmmSpline command

lmms.obj a lmmspline object

time a numeric vector containing the sample time point information.

homoskedasticity

a logical whether or not to test for homoscedasticity with the Breusch-Pagan

test.

MSE.filter whether or not to test for low dispersion with a cutoff on the MSE.

homoskedasticity.cutoff

a numeric scalar between 0 and 1, p-value threshold for B-P test.

#### **Details**

\* homo-sedasticity of the residues with a Breusch-Pagan test \* low dispersion with a cutoff on the MSE (mean squared error)

#### Value

a list containing the following items

filtering.summary

a data.frame with the different tests per features (passed = TRUE, failed =

FALSE)

to.keep features which passed all the tests

filtered the filtered data.frame

#### See Also

**bptest** 

```
# data and lmms output
data(timeOmics.simdata)
data <- timeOmics.simdata$sim
lmms.output <- timeOmics.simdata$lmms.output
time <- timeOmics.simdata$time

# filter
filter.res <- lmms.filter.lines(data = data, lmms.obj = lmms.output, time = time)</pre>
```

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plotLong

Plot Longitudinal Profiles by Cluster

# Description

This function provides a expression profile representation over time and by cluster.

# Usage

```
plotLong(
  object,
  time = NULL,
  plot = TRUE,
  center = TRUE,
  scale = TRUE,
  title = "Time-course Expression",
  X.label = NULL,
  Y.label = NULL,
  legend = FALSE,
  legend.title = NULL,
  legend.block.name = NULL
)
```

# Arguments

object	a mixOmics result of class (s)pca, (s)pls, block.(s)pls.
object	a mixormes result of class (s)pea, (s)pis, block.(s)pis.
time	(optional) a numeric vector, the same size as ncol(X), to change the time scale.
plot	a logical, if TRUE then a plot is produced. Otherwise, the data.frame on which the plot is based on is returned.
center	a logical value indicating whether the variables should be shifted to be zero centered.
scale	a logical value indicating whether the variables should be scaled to have unit variance before the analysis takes place.
title	character indicating the title plot.
X.label	x axis titles.
Y.label	y axis titles.
legend	a logical, to display or not the legend.
legend.title	if legend is provided, title of the legend.
legend.block.na	ame
	a character vector corresponding to the size of the number of blocks in the mixOmics object.

# Value

a data.frame (gathered form) containing the following columns:

```
 \begin{array}{ll} \text{time} & x \text{ axis values} \\ \text{molecule} & \text{names of features} \end{array}
```

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```
value y axis values
cluster assigned clusters
block name of 'blocks'
```

#### See Also

```
getCluster
```

### **Examples**

```
demo <- suppressWarnings(get_demo_cluster())</pre>
X <- demo$X
Y <- demo$Y
Z <- demo$Z
# (s)pca
pca.res <- mixOmics::pca(X, ncomp = 3)</pre>
plotLong(pca.res)
spca.res \leftarrow mixOmics::spca(X, ncomp = 2, keepX = c(15, 10))
plotLong(spca.res)
# (s)pls
pls.res <- mixOmics::pls(X,Y)</pre>
plotLong(pls.res)
spls.res \leftarrow mixOmics::spls(X,Y, keepX = c(15,10), keepY=c(5,6))
plotLong(spls.res)
# (s)block.spls
block.pls.res <- mixOmics::block.pls(X=list(X=X,Z=Z), Y=Y)</pre>
plotLong(block.pls.res)
block.spls.res <- mixOmics::block.spls(X=list(X=X,Z=Z), Y=Y,</pre>
                               keepX = list(X = c(15,10), Z = c(5,6)),
                               keepY = c(3,6))
plotLong(block.spls.res)
```

proportionality

Proportionality Distance

# **Description**

proportionality is a wrapper that compute proportionality distance for a clustering result (pca, spca, pls, spls, block.pls, block.spls). and it performs a u-test to compare the median within a cluster to the median of the entire background set.

# Usage

```
proportionality(X)
```

#### **Arguments**

Χ

an object of the class: pca, spca, pls, spls, block.pls or block.spls

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#### Value

Return a list containing the following components:

 $\label{lem:proprior} \mbox{propr.distance Square matrix with proportionality distance between pairs of features propr.distance.w.cluster$ 

distance between pairs with cluster label

pvalue Wilcoxon U-test p-value comparing the medians within clusters and with the

entire background set

#### References

Lovell, D., Pawlowsky-Glahn, V., Egozcue, J. J., Marguerat, S., Bähler, J. (2015). Proportionality: a valid alternative to correlation for relative data. PLoS Comput. Biol. 11, e1004075. doi: 10.1371/journal.pcbi.1004075

Quinn, T. P., Richardson, M. F., Lovell, D., Crowley, T. M. (2017). propr: an r-package for identifying proportionally abundant features using compositional data analysis. Sci. Rep. 7, 16252. doi: 10.1038/s41598-017-16520-0

#### **Examples**

```
demo <- suppressWarnings(get_demo_cluster())
# pca
X <- demo$pca
propr.res <- proportionality(X)
plot(propr.res)
# pls
X <- demo$spls
propr.res <- proportionality(X)
plot(propr.res)
# block.pls
X <- demo$block.spls
propr.res <- proportionality(X)
plot(propr.res)</pre>
```

remove.low.cv

Remove features with low variation

### **Description**

remove.low.cv that removes variables with low variation. From a matrix/data.frame (samples in rows, features in columns), it computes the coefficient of variation for every features (columns) and return a filtered data.frame with features for which the coefficient of variation is above a given threshold.

#### Usage

```
remove.low.cv(X, cutoff = 0.5)
```

#### **Arguments**

```
X a matrix/data.frame cutoff a numeric value
```

#### Value

a data.frame/matrix

# **Examples**

```
mat <- matrix(sample(1:3, size = 200, replace = TRUE), ncol=20)
remove.low.cv(mat, 0.4)</pre>
```

```
tuneCluster.block.spls
```

Feature Selection Optimization for block (s)PLS method

# **Description**

This function identify the number of feautures to keep per component and thus by cluster in mixOmics::block.spls by optimizing the silhouette coefficient, which assesses the quality of clustering.

#### Usage

```
tuneCluster.block.spls(
   X,
   Y = NULL,
   indY = NULL,
   ncomp = 2,
   test.list.keepX = NULL,
   test.keepY = NULL,
   ...
)
```

# Arguments

V	list of numeric matrix	(am data fuama) with	factures in columns on	d commeles in
λ	usi or numeric mairix	tor data frame) with	-realities in collitions an	a sambles in

rows (with samples order matching in all data sets).

Y (optional) numeric matrix (or data.frame) with features in columns and samples

in rows (same rows as X).

indy integer, to supply if Y is missing, indicates the position of the matrix response

in the list X.

ncomp integer, number of component to include in the model

test.list.keepX

list of integers with the same size as X. Each entry corresponds to the different

keepX value to test for each block of X.

test.keepY only if Y is provideid. Vector of integer containing the different value of keepY

to test for block Y.

... other parameters to be included in the spls model (see mixOmics::block.spls)

#### **Details**

For each component and for each keepX/keepY value, a spls is done from these parameters. Then the clustering is performed and the silhouette coefficient is calculated for this clustering.

We then calculate "slopes" where keepX/keepY are the coordinates and the silhouette is the intensity. A z-score is assigned to each slope. We then identify the most significant slope which indicates a drop in the silhouette coefficient and thus a deterioration of the clustering.

#### Value

silhouette	silhouette coef. computed for every combinasion of keepX/keepY
ncomp	number of component included in the model
test.keepX	list of tested keepX
test.keepY	list of tested keepY
block	names of blocks
slopes	"slopes" computed from the silhouette coef. for each keep $\!$
choice.keepX	best keepX for each component
choice.keepY	best keepY for each component

#### See Also

```
block.spls, getCluster, plotLong
```

```
demo <- suppressWarnings(get_demo_cluster())</pre>
X \leftarrow list(X = demo$X, Z = demo$Z)
Y \leftarrow demo Y
test.list.keepX <- list("X" = c(5,10,15,20), "Z" = c(2,4,6,8))
test.keepY <- c(2:5)</pre>
tune.block.spls <- tuneCluster.block.spls(X= X, Y= Y,</pre>
                                              test.list.keepX= test.list.keepX,
                                              test.keepY= test.keepY,
                                              mode= "canonical")
keepX <- tune.block.spls$choice.keepX</pre>
keepY <- tune.block.spls$choice.keepY</pre>
# final model
block.spls.res <- mixOmics::block.spls(X= X, Y= Y, keepX = keepX,</pre>
                                keepY = keepY, ncomp = 2, mode = "canonical")
# get clusters and plot longitudinal profile by cluster
block.spls.cluster <- getCluster(block.spls.res)</pre>
```

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tuneCluster.spca

Feature Selection Optimization for sPCA method

# **Description**

This function identify the number of feautures to keep per component and thus by cluster in mixOmics::spca by optimizing the silhouette coefficient, which assesses the quality of clustering.

#### Usage

```
tuneCluster.spca(X, ncomp = 2, test.keepX = rep(ncol(X), ncomp), ...)
```

### **Arguments**

Χ	numeric matrix (or data.frame) with features in columns and samples in rows
ncomp	integer, number of component to include in the model
test.keepX	vector of integer containing the different value of keepX to test for block X.
	other parameters to be included in the spls model (see mixOmics::spca)

#### **Details**

For each component and for each keepX value, a spls is done from these parameters. Then the clustering is performed and the silhouette coefficient is calculated for this clustering.

We then calculate "slopes" where keepX are the coordinates and the silhouette is the intensity. A z-score is assigned to each slope. We then identify the most significant slope which indicates a drop in the silhouette coefficient and thus a deterioration of the clustering.

## Value

silhouette silhouette coef. computed for every combinasion of keepX/keepY

ncomp number of component included in the model

test.keepX list of tested keepX block names of blocks

slopes "slopes" computed from the silhouette coef. for each keepX and keepY, used to

determine the best keepX and keepY

choice.keepX best keepX for each component

```
demo <- suppressWarnings(get_demo_cluster())
X <- demo$X

# tuning
tune.spca.res <- tuneCluster.spca(X = X, ncomp = 2, test.keepX = c(2:10))
keepX <- tune.spca.res$choice.keepX
plot(tune.spca.res)

# final model
spca.res <- mixOmics::spca(X=X, ncomp = 2, keepX = keepX)
plotLong(spca.res)</pre>
```

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tuneCluster.spls	Feature Selection Optimization for sPLS method
------------------	--

# **Description**

This function identify the number of feautures to keep per component and thus by cluster in mixOmics::spls by optimizing the silhouette coefficient, which assesses the quality of clustering.

# Usage

```
tuneCluster.spls(
   X,
   Y,
   ncomp = 2,
   test.keepX = rep(ncol(X), ncomp),
   test.keepY = rep(ncol(Y), ncomp),
   ...
)
```

# **Arguments**

X	numeric matrix (or data.frame) with features in columns and samples in rows
Υ	numeric matrix (or data.frame) with features in columns and samples in rows (same rows as $X$ )
ncomp	integer, number of component to include in the model
test.keepX	vector of integer containing the different value of keepX to test for block X.
test.keepY	vector of integer containing the different value of keepY to test for block Y.
	other parameters to be included in the spls model (see mixOmics::spls)

# **Details**

For each component and for each keepX/keepY value, a spls is done from these parameters. Then the clustering is performed and the silhouette coefficient is calculated for this clustering.

We then calculate "slopes" where keepX/keepY are the coordinates and the silhouette is the intensity. A z-score is assigned to each slope. We then identify the most significant slope which indicates a drop in the silhouette coefficient and thus a deterioration of the clustering.

#### Value

silhouette	silhouette coef. computed for every combinasion of keepX/keepY
ncomp	number of component included in the model
test.keepX	list of tested keepX
test.keepY	list of tested keepY
block	names of blocks
slopes	"slopes" computed from the silhouette coef. for each keep $\!$
choice.keepX	best keepX for each component
choice.keepY	best keepY for each component

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#### See Also

```
spls, getCluster, plotLong
```

#### **Examples**

```
demo <- suppressWarnings(get_demo_cluster())
X <- demo$X
Y <- demo$Y

# tuning
tune.spls <- tuneCluster.spls(X, Y, ncomp= 2, test.keepX= c(5,10,15,20), test.keepY= c(2,4,6))
keepX <- tune.spls$choice.keepX
keepY <- tune.spls$choice.keepY

# final model
spls.res <- mixOmics::spls(X, Y, ncomp= 2, keepX= keepX, keepY= keepY)

# get clusters and plot longitudinal profile by cluster
spls.cluster <- getCluster(spls.res)
plotLong(spls.res)</pre>
```

unscale

Unscales a scaled data.frame

# **Description**

unscale is a generic function that unscale and/or uncenter the columns of a matrix generated by the scale base function

# Usage

```
unscale(x)
```

# Arguments

Χ

A numeric matrix.

#### **Details**

unscale uses attributes added by the scale function "scaled:scale" and "scaled:center" and use these scaling factor to retrieve the initial matrix. It first unscales and then uncenters.

## Value

Return a matrix, uncenterd and unscaled. Attributes "scaled:center" and "scaled:scale" are removed.

# See Also

scale

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```
X <- matrix(1:9, ncol = 3)
X.scale <- scale(X, center = TRUE, scale = TRUE)
X.unscale <- unscale(X.scale)
all(X == X.unscale)</pre>
```

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