

Package ‘moanin’

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Title An R Package for Time Course RNASeq Data Analysis

Version 1.12.0

Description Simple and efficient workflow for time-course gene expression data, built on publicly available open-source projects hosted on CRAN and bioconductor. moanin provides helper functions for all the steps required for analysing time-course data using functional data analysis: (1) functional modeling of the timecourse data; (2) differential expression analysis; (3) clustering; (4) downstream analysis.

Depends R (>= 4.0), SummarizedExperiment, topGO, stats

Imports S4Vectors, MASS (>= 1.0.0), limma, viridis, edgeR, graphics, methods, grDevices, reshape2, NMI, zoo, ClusterR, splines, matrixStats

Suggests testthat (>= 1.0.0), timecoursedata, knitr, rmarkdown, markdown, covr, BiocStyle

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consensus_matrix	<i>Compute consensus matrix from labels</i>
------------------	---

Description

Compute consensus matrix from labels

Usage

```
consensus_matrix(labels, scale = TRUE)
```

Arguments

labels	a matrix with each column corresponding to a the results of a single clustering routine. Each column should give the cluster assignment FIXME : What is the required format of entries??
scale	boolean, optional, default: TRUE. Whether to rescale the resulting consensus matrix so that entries correspond to proportions.

Value

a symmetric matrix of size $N \times N$, where N is the number of rows of the input matrix labels. Each i,j entry of the matrix corresponds the number of times the two rows were in the same cluster across the clusterings (`scale=FALSE`) or the proportion of clustering that the two rows are in the same cluster (`scale=TRUE`).

Examples

```
data(exampleData)
moanin <- create_moanin_model(data=testData,meta=testMeta)
#small function to run splines_kmeans on subsample of 50 genes
subsampleCluster<-function(){
  ind<-sample(1:nrow(moanin),size=50,replace=FALSE)
  km<-splines_kmeans(moanin[ind,],n_clusters=3)
  assign<-splines_kmeans_predict(moanin,km,
    method="distance")
}
kmClusters=replicate(10,subsampleCluster())
cm<-consensus_matrix(kmClusters)
heatmap(cm)
```

`create_meta_prediction`*Create prediction meta data from splines model*

Description

Create prediction meta data from splines model

Usage

```
create_meta_prediction(moanin_model, num_timepoints = 100)
```

Arguments

`num_timepoints` integer, optional, default: 100. Number of timepoints to use for the prediction metadata

Value

`moanin_model`

DE_timecourse

*Run spline models and test for DE of contrasts.***Description**

Run spline models and test for DE of contrasts.

Usage

```
## S4 method for signature 'Moanin'
DE_timecourse(
  object,
  contrasts,
  center = FALSE,
  statistic = c("ftest", "lrt"),
  use_voom_weights = TRUE
)
```

Arguments

object	An object of class Moanin , an object containing all related information for time course data and the splines model that will be used (if applicable). See create_moanin_model for more details.
contrasts	Contrasts, either provided as a vector of strings, or a matrix of contrasts coefficients obtained using makeContrasts from the package <code>limma</code> . If given as a character string, will be passed to makeContrasts to be converted into such a matrix.
center	boolean, whether to center the data matrix
statistic	Which test statistic to use, a likelihood ratio statistic or a F-test.
use_voom_weights	boolean, optional, default: TRUE. Whether to use voom weights. See details.

Details

The implementation of the spline fit and the calculation of p-values was based on code from [edge](#), and expanded to enable handling of comparisons of groups via contrasts. The code assumes that the `Moanin` object was created via either a formula or a basis where a different spline was fit for each `group_variable` and thus the contrasts are comparisons of those spline fits. If the `Moanin` object was created via user-provided basis matrix or formula, then the user should take a great deal of caution in using this code, as the degrees of freedom for the tests of significance cannot be verified to be correct.

If `use_voom_weights=TRUE`, then before fitting splines to each gene, voom weights are calculated from `assay(object)`:

```

y <- edgeR::DGEList(counts=assay(object))
y <- edgeR::calcNormFactors(y, method="upperquartile")
v <- limma::voom(y, design, plot=FALSE)
weights <- v$weights

```

The design matrix for the voom weights is based on the formula \sim Group + Timepoint + 0 where Group and Timepoint are replaced with the user-defined values where appropriate. These weights are given to the `lm.fit` which fits the spline coefficients. This workflow assumes that the input to the Moanin object were counts.

If the user set `log_transform=TRUE` in the creation of the Moanin object, the splines will be fit to the log of the input data, and not directly to the input data. This is independent of whether the user chooses `use_voom_weights`.

Value

A data frame with two columns for each of the contrasts given in `contrasts`, corresponding to the raw p-value of the contrast for that gene (`_pval`) and the adjusted p-value (`_qval`). The adjusted p-values are FDR-adjusted based on the Benjamini-Hochberg method, as implemented in [p.adjust](#). The adjustment is done across all p-values for all contrasts calculated.

See Also

[makeContrasts](#), [create_moanin_model](#), [DE_timepoints](#), [edge](#)

Examples

```

data(exampleData)
moanin <- create_moanin_model(data=testData, meta=testMeta)
deTimecourse=DE_timecourse(moanin,
  contrasts="K-C", use_voom_weights=FALSE)
head(deTimecourse)

```

DE_timepoints

Fit weekly differential expression analysis

Description

Fit weekly differential expression analysis
 Creates pairwise contrasts for all timepoints

Usage

```

## S4 method for signature 'Moanin'
DE_timepoints(object, contrasts, add_factors = NULL, use_voom_weights = TRUE)

## S4 method for signature 'Moanin'
create_timepoints_contrasts(

```

```

object,
group1,
group2 = NULL,
type = c("per_timepoint_group_diff", "per_group_timepoint_diff",
         "group_and_timepoint_diff"),
timepoints = sort(unique(time_variable(object))),
timepoints_before = head(sort(timepoints), -1),
timepoints_after = tail(sort(timepoints), -1),
format = c("vector", "data.frame")
)

```

Arguments

object	An object of class <code>Moanin</code> , an object containing all related information for time course data and the splines model that will be used (if applicable). See <code>create_moanin_model</code> for more details.
contrasts	Contrasts, either provided as a vector of strings, or a matrix of contrasts coefficients obtained using <code>makeContrasts</code> from the package <code>limma</code> . If given as a character string, will be passed to <code>makeContrasts</code> to be converted into such a matrix.
add_factors	A character vector of additional variables to add to the design. See details.
use_voom_weights	boolean, optional, default: <code>TRUE</code> . Whether to use voom weights. See details.
group1	First group to consider in making contrasts, character value that must match a value of the grouping variable contained in <code>moanin_model</code> .
group2	Second group to consider in making contrasts, character value that must match a value of the grouping variable contained in <code>moanin_model</code> , unless <code>type=="per_group_timepoint_diff"</code> , in which case should be <code>NULL</code> (only <code>group1</code> is used in comparison)
type	the type of contrasts that should be created. See details.
timepoints	vector of timepoints to compare. Must be contained in the <code>time_variable</code> of the <code>moanin</code> object.
timepoints_before	for type equal to <code>"per_group_timepoint_diff"</code> or <code>"group_and_timepoint_diff"</code> , the set of timepoints to compare, see details. By default, taken from the <code>timepoints</code> variable.
timepoints_after	for type equal to <code>"per_group_timepoint_diff"</code> or <code>"group_and_timepoint_diff"</code> , the set of timepoints to compare, see details. By default, taken from the <code>timepoints</code> variable.
format	the choice of <code>"vector"</code> (the default) for <code>create_timepoints_contrasts</code> returns just the character vector of contrasts. If instead <code>format="data.frame"</code> then a <code>data.frame</code> is return that identifies the timepoint and group comparisons involved in each contrast. If this is the desired output, then the input to <code>DE_timepoints</code> should be the column corresponding to the contrast. See examples.

Details

By default the formula fitted for each gene is

```
~ Group*Timepoint +0
```

If the user gives values to `add_factors`, then the vector of character values given in `add_factors` will be *added* to the default formula. So that `add_factors="Replicate"` will change the formula to

```
~ Group*Timepoint +0 + Replicate
```

This allows for a small amount of additional complexity to control for other variables. Users should work directly with `limma` for more complex models.

If `use_voom_weights=TRUE`, the data is given directly to `limma` via `assay(object)`. The specific series of calls is:

```
y <- edgeR::DGEList(counts=assay(object))
y <- edgeR::calcNormFactors(y, method="upperquartile")
v <- limma::voom(y, design, plot=FALSE)
v <- limma::lmFit(v)
```

If the user set `log_transform=TRUE` in the creation of the `Moanin` object, this will not have an impact in the analysis if `use_voom_weights=TRUE`. Only if `use_voom_weights=FALSE` will this matter, in which case the log of the input data will be given to a regular call to `limma`:

```
y<-get_log_data(object)
v <- limma::lmFit(y, design)
```

`create_timepoints_contrasts` creates the needed contrasts for comparing groups or timepoints in the format needed for `DE_timepoints` (i.e. `makeContrasts`), to which the contrasts are ultimately passed. The time points and groups are determined by the levels of the `grouping_variable` and the values of `time_variable` in the `moanin_object` provided by the user.

Three different types of contrasts are created:

- "per_timepoint_group_diff" Contrasts that compare the groups within a timepoint
- "per_group_timepoint_diff" Contrasts that compare two timepoints within a group
- "group_and_timepoint_diff" Contrasts that compare the difference between two timepoints between two levels of the `group_variable` of the `Moanin` object. These are contrasts in the form $(TP_i - TP_{(i-1)})[Group1] - (TP_i - TP_{(i-1)})[Group2]$.

Value

`create_timepoints_contrasts`: a character vector with each element of the vector corresponding to a contrast to be compared.

See Also

[makeContrasts](#)

Examples

```

data(exampleData)
moanin <- create_moanin_model(data=testData, meta=testMeta)
# compare groups within each timepoint
contrasts <- create_timepoints_contrasts(moanin,"C", "K",
  type="per_timepoint_group_diff")
head(contrasts)
deTimepoints=DE_timepoints(moanin,
  contrasts=contrasts, use_voom_weights=FALSE)
head(deTimepoints)
# Control for replicate variable:
deTimepoints=DE_timepoints(moanin,
  contrasts=contrasts, add_factors="Replicate",
  use_voom_weights=FALSE)
head(deTimepoints)

# compare adjacent timepoints within each group
contrastsDiff <- create_timepoints_contrasts(moanin,"C",
  type="per_group_timepoint_diff")
deDiffTimepoints=DE_timepoints(moanin,
  contrasts=contrastsDiff,
  use_voom_weights=FALSE)
# provide the sets of timepoints to compare:
contrastsDiff2<-create_timepoints_contrasts(moanin,"C",
  timepoints_before=c(72,120),timepoints_after=c(168,168),
  type="per_group_timepoint_diff")
deDiffTimepoints2=DE_timepoints(moanin,
  contrasts=contrastsDiff2,
  use_voom_weights=FALSE)

# Compare selected timepoints across groups.
# This time we also return format="data.frame" which helps us keep track of
# the meaning of each contrast.
contrastsGroupDiff<-create_timepoints_contrasts(moanin,"C", "K",
  timepoints_before=c(72,120),timepoints_after=c(168,168),
  type="group_and_timepoint_diff",format="data.frame")
head(contrastsGroupDiff)
deGroupDiffTimepoints=DE_timepoints(moanin,
  contrasts=contrastsGroupDiff$contrasts,
  use_voom_weights=FALSE)

```

discont_basis

Provides set of basis functions on either side of a time point, allowing for a discontinuity in the fitted functions

Description

Provides set of basis functions on either side of a time point, allowing for a discontinuity in the fitted functions

Usage

```
discont_basis(
  timepoints,
  discont_point,
  knots = NULL,
  dfPre = NULL,
  dfPost = dfPre,
  degree = 3,
  intercept = TRUE,
  type = c("ns", "bs")
)
```

Arguments

timepoints	vector of numeric timepoints for which the splines basis will be evaluated
discont_point	a single numeric value that represents where the discontinuity should be
knots	passed to ns or bs. If not NULL, should give knots on either side of discon_point as single vector – they will be separated in the call to discon_point
dfPre	the df for the basis functions defined before the discontinuity point
dfPost	the df for the basis functions defined after the discontinuity point
degree	passed to bs (if applicable)
intercept	Whether to include an intercept (vector of all 1s) for each side of the discontinuity. Note this is different than the argument intercept of either bs or ns, which is set to FALSE.
type	either "ns" or "bs" indicating which splines basis function should be used.

Examples

```
x<-seq(0,10,length=100)
basis<-discont_basis(x,discont_point=3, dfPre=3, dfPost=4, intercept=TRUE)
# Plot of the basis functions
par(mfrow=c(3,3))
for(i in 1:ncol(basis)){
  plot(x,basis[,i],type="l")
  abline(v=3,lty=2)
}
# Use it in a moanin_model object instead of ns/bs:
data(exampleData)
moanin <- create_moanin_model(data=testData, meta=testMeta,
  spline_formula=~Group:discont_basis(Timepoint,dfPre=3,
    dfPost=3,discont=20,intercept=TRUE)+0,
  degrees_of_freedom=6)
```

```
estimate_log_fold_change
```

Estimates log fold change

Description

Estimates log fold change

Usage

```
## S4 method for signature 'Moanin'
estimate_log_fold_change(
  object,
  contrasts,
  method = c("timecourse", "sum", "max", "timely", "abs_sum", "abs_squared_sum", "min")
)
```

Arguments

object	An object of class Moanin , an object containing all related information for time course data and the splines model that will be used (if applicable). See create_moanin_model for more details.
contrasts	The contrasts to consider
method	method for calculating the log-fold change. See details.

Details

The following methods exist for calculating the log-fold change between conditions over time (default is "timecourse"):

- **timely**The log-fold change for each individual timepoint ($lfc(t)$)
- **timecourse**The average absolute per-week fold-change, multiplied by the sign of the average per-week fold-change.
- **sum**Sum of per-week log fold change, over all timepoints
- **max**Max of per-week log fold change, over all timepoints
- **abs_sum**Sum of the absolute value of the per-week log fold change, over all timepoints
- **abs_squared_sum**Sum of the square value of the per-week log fold change, over all timepoint
- **min**Min of per-week log fold change, over all timepoints

If the user set `log_transform=TRUE` in the creation of the `Moanin` object, the data will be log transformed before calculating the fold-change.

Value

A `data.frame` giving the estimated log-fold change for each gene (row). For all methods except for "timely", the data frame will consist of one column for each value of the argument contrasts. For "timely" there will be one column for each timepoint and contrast combination.

Examples

```

data(exampleData)
moanin <- create_moanin_model(data=testData,meta=testMeta)
estsTimely <- estimate_log_fold_change(moanin,
  contrasts=c("K-C"), method="timely")
head(estsTimely)
estsTimecourse <- estimate_log_fold_change(moanin,
  contrasts=c("K-C"),method="timecourse")
head(estsTimecourse)

```

exampleData

*Small data set for running examples***Description**

Small data set for running examples

Format

Three objects are loaded, a data frame of expression of 500 genes by 84 samples (testData), a data frame with meta information on those 84 samples (testMeta), and a data frame giving the GOID of the genes in testData.

Details

This data is a subset of the full time course data available as [shoemaker2015](#) and is only provided for the purpose of running examples, and not for biological meaning. Users should refer to the full data set.

The rownames of testData are RefSeq.

Examples

```

#code used to create data:
## Not run:
library(timecoursedata)
data(shoemaker2015)
testData<-shoemaker2015$data[1:500,]
whSamples<-which(shoemaker2015$meta$Group %in% c("C","K","M"))
testData<-testData[,whSamples]
testMeta<-droplevels(shoemaker2015$meta[whSamples,])
library(biomaRt)
ensembl = useMart("ensembl")
ensembl = useMart("ensembl")
ensembl = useDataset("mmusculus_gene_ensembl", mart=ensembl)
testGenesGO = getBM(attributes=c("go_id", "refseq_mrna"),
  values=rownames(testData),
  filters="refseq_mrna",
  mart=ensembl)
save(list=c("testData","testMeta","testGenesGO"),file="data/exampleData.rda")

```

```
## End(Not run)
```

expand_contrast *Internal Validation Checks*

Description

Will check that the contrasts provided are indeed contrasts. contrasts are expected to be either a vector of string or a matrix containing the contrasts coefficients.

Usage

```
expand_contrast(moanin_model, contrast_vector)

check_data_meta(data, object)

check_is_2d(X)

is_contrasts(contrasts, moanin_model)
```

Details

If a vector of string is provided, the function will call `limma::makeContrast` in order to obtain the contrasts coefficients.

If a contrasts matrix is provided, it will perform a number of checks on the contrasts matrix to make sure it contains the number of rows expected, and that each contrast indeed sums to 0.

Value

Does not return anything. Only hits errors if there are problems.

`is_contrasts` returns the contrasts, with any corrections.

expression_filtering *Utility function to filter out low-expressed genes*

Description

Utility function to filter out low-expressed genes

Usage

```
expression_filtering(counts, min_counts = 20, min_samples = 3)
```

Arguments

counts	n by p matrix containing the count data.
min_counts	integer, corresponding to the minimum number of counts for the gene to be considered expressed in a sample. Default is 20.
min_samples	integer, corresponding to the minimum of samples for a gene to be expressed to be included in downstream analyses.

Value

The filtered counts matrix

find_enriched_go_terms

Find enriched GO terms

Description

Find enriched GO terms

Create the Gene to GO Term mapping

Usage

```
find_enriched_go_terms(
  assignments,
  gene_id_to_go,
  ontology = "BP",
  weighted = FALSE,
  node_size = 10
)

create_go_term_mapping(genes, gene_col = "refseq_mrna")
```

Arguments

assignments	boolean named vector determining the gene subset to be tested for enrichment of GO terms. The names of the vector should be the gene names. Elements with TRUE will consist of the gene cluster.
gene_id_to_go	List giving the Gene ID to GO object required for topGO (see topGOdata-class). <code>create_go_term_mapping</code> can construct such a list from a data-frame.
ontology	string, optional, default: BP. specifies which ontology to use (passed to ontology argument in creating a new <code>topGOdata</code> object). Can be 'BP', 'CC', or 'NF'. See topGOdata-class .
weighted	boolean, optional, default: FALSE. Whether to use the weighted algorithm or not in runTest .

node_size	integer, optional, default: 10. Consider only GO terms with node_size number of genes, passed to nodeSize argument of topGOdata-class
genes	dataframe, with two required columns. The first gives the gene names, with column name by the argument gene_col. The other column must be named "go_id" and give the genes GO id. Genes will have multiple GO id that they map to, and each go mapping of a gene is a separate row. Thus genes will be in multiple rows of the input.
gene_col	the name of the column of the genes data frame that contains the correct gene reference. By default, is "refseq_mrna".

Details

find_enriched_go_terms is a wrapper for running a GO enrichment analysis via the package topGO. This function creates a [topGOdata-class](#) object, runs the function [runTest](#) to test for enrichment using the statistic="fisher" option, and then runs [GenTable](#). This function then does some post-processing of the results, returning only GO terms that satisfy:

1. BH adjusted p-values less than 0.05 using [p.adjust](#)
2. GO terms are *enriched*, i.e. the number of genes from the GO term found in the subset is greater than expected

Value

Returns results in the format of [GenTable](#).

create_go_term_mapping returns a list giving the gene to GO id in the format required by [topGOdata-class](#).

See Also

[create_go_term_mapping](#), [find_enriched_pathway](#), [GenTable](#), [runTest](#), [topGOdata-class](#), [p.adjust](#)

Examples

```
data(exampleData)
head(testGenesGO) #gives the mapping of genes to GO
geneId2Go <- create_go_term_mapping(testGenesGO)
#create fake assignment of genes to group based on TRUE/FALSE values
inGroup=rep(FALSE,nrow(testData))
inGroup[1:10]=TRUE
names(inGroup) <- row.names(testData)
find_enriched_go_terms(inGroup, geneId2Go)
```

 Moanin-class

 Class *Moanin*

Description

Moanin is a class that extends `SummarizedExperiment` and is used to store the additional spline basis and meta data for timecourse analysis.

In addition to the slots of the `SummarizedExperiment` class, the `Moanin` object has the additional slots described in the Slots section.

There are several methods implemented for this class. The most important methods have their own help page. Simple helper methods are described in the Methods section below. For a comprehensive list of methods specific to this class see the Reference Manual.

The constructor `create_moanin_model` creates an object of the class `Moanin`.

Usage

```
## S4 method for signature 'DataFrame'
create_moanin_model(data, meta, ...)

## S4 method for signature 'data.frame'
create_moanin_model(data, ...)

## S4 method for signature 'matrix'
create_moanin_model(data, meta, ...)

## S4 method for signature 'SummarizedExperiment'
create_moanin_model(
  data,
  spline_formula = NULL,
  basis_matrix = NULL,
  group_variable_name = "Group",
  time_variable_name = "Timepoint",
  degrees_of_freedom = NULL,
  log_transform = FALSE,
  drop_levels = TRUE
)
```

Arguments

<code>data</code>	The input data. Can be a SummarizedExperiment class, or matrix/data.frame. If the input data is a matrix or data.frame, then the user must also provide input to the meta argument, which will be transformed into colData of the resulting Moanin object
<code>meta</code>	Meta data on the samples (columns) of the data argument. Must be given if input data is a matrix or data.frame. If input is SummarizedExperiment, this argument is ignored.

...	arguments passed from methods to the SummarizedExperiment method.
spline_formula	formula object, optional, default: NULL. Used to construct splines from the data in meta. See details.
basis_matrix	matrix, optional, default: NULL. A basis matrix, where each row corresponds to the evaluation of a sample on the basis function (thus one column for each basis function).
group_variable_name	A character value giving the column that corresponds to the grouping variable to test for DE. By default "Group"
time_variable_name	A character value giving the column that corresponds to the time variable. By default "Timepoint".
degrees_of_freedom	int, optional. Number of degrees of freedom to use if neither the basis_matrix nor the spline_formula is provided. If not provided by the user, internally will be set to 4
log_transform	whether the data should be log-transformed by certain methods (see splines_kmeans)
drop_levels	Logical, whether to perform droplevels on the grouping variable (i.e. remove empty levels)

Details

If neither `spline_formula` nor `basis_matrix` is given, then by default, the function will create a basis matrix based on the formula:

```
spline_formula = ~Group:ns(Timepoint, df=4) + Group +
0
```

Note that the meta data will have levels dropped (via `droplevels`).

Input to data that is given as a class `DataFrame` or `data.frame` will be converted to class `matrix`. The reason for this is that use of a `data.frame` creates errors in taking duplicate rows/columns of `SummarizedExperiment`, as in bootstrapping. Users who absolutely want the object to hold a object that is not a matrix can construct a `SummarizedExperiment` object (which will not convert the input into a matrix), and use this as input to `create_moanin_model`.

Value

An object of class `Moanin`

Slots

<code>time_variable_name</code>	character value giving the column in <code>colData</code> that defines the time variable (must be of class <code>numeric</code>)
<code>group_variable_name</code>	character value giving the column in <code>colData</code> that defines the grouping variable (must be of class <code>factor</code>)
<code>basis_matrix</code>	A basis matrix, where each row corresponds to the evaluation of a sample on the basis function (thus one column for each basis function).

spline_formula a formula. The formula used in creating the basis matrix

degrees_of_freedom a numeric integer. Number of degrees of freedom used in creating basis matrix. If NULL, degrees of freedom is not known (usually if user provided basis without degrees of freedom)

log_transform logical, whether to log-transform the data for certain methods

Examples

```
# Load some data
data(exampleData)

# Use the default options
moanin = create_moanin_model(data=testData,meta=testMeta)
moanin

# Change the number of degrees of freedom
moanin = create_moanin_model(data=testData,meta=testMeta,
  degrees_of_freedom=6)
moanin
```

Moanin-methods

Helper methods for the Moanin class

Description

This is a collection of helper methods for the Moanin class.

Usage

```
## S4 method for signature 'Moanin'
group_variable_name(object)

## S4 replacement method for signature 'Moanin'
group_variable_name(object) <- value

## S4 method for signature 'Moanin'
time_by_group_variable(object)

## S4 method for signature 'Moanin'
group_variable(object)

## S4 replacement method for signature 'Moanin'
group_variable(object) <- value

## S4 method for signature 'Moanin'
time_variable_name(object)
```

```

## S4 replacement method for signature 'Moanin'
time_variable_name(object) <- value

## S4 method for signature 'Moanin'
time_variable(object)

## S4 replacement method for signature 'Moanin'
time_variable(object) <- value

## S4 method for signature 'Moanin'
degrees_of_freedom(object)

## S4 method for signature 'Moanin'
basis_matrix(object)

## S4 method for signature 'Moanin'
spline_formula(object)

## S4 method for signature 'Moanin'
show(object)

## S4 method for signature 'Moanin,ANY,character,ANY'
x[i, j, ..., drop = TRUE]

## S4 method for signature 'Moanin,ANY,logical,ANY'
x[i, j, ..., drop = TRUE]

## S4 method for signature 'Moanin,ANY,numeric,ANY'
x[i, j, ..., drop = TRUE]

## S4 method for signature 'Moanin'
log_transform(object)

## S4 method for signature 'Moanin'
get_log_data(object)

```

Arguments

object	An object of class Moanin , an object containing all related information for time course data and the splines model that will be used (if applicable). See create_moanin_model for more details.
value	replacement value
x	Moanin object
i, j	A vector of logical or integer subscripts, indicating the rows and columns to be subsetted for i and j, respectively.
...	arguments passed to subsetting
drop	A logical scalar that is ignored.

Details

Note that when subsetting the data, the dendrogram information and the co-clustering matrix are lost.

Value

group_variable_name and time_variable_name return the name of the column containing the variable. group_variable and time_variable return the actual variable.

Examples

```
# Load some data
data(exampleData)
moanin = create_moanin_model(data=testData,meta=testMeta)
group_variable_name(moanin)
time_variable_name(moanin)
```

perWeek_barplot	<i>Creates barplot of results of per-timepoint comparison</i>
-----------------	---

Description

Creates barplot of results of per-timepoint comparison

Usage

```
perWeek_barplot(
  de_results,
  type = c("qval", "pval"),
  labels = NULL,
  threshold = 0.05,
  xlab = "Timepoint",
  ylab = "Number of DE genes",
  main = "",
  ...
)
```

Arguments

de_results	results from DE_timepoints
type	type of p-value to count ("qval" or "pval")
labels	labels to give each bar
threshold	cutoff for counting gene as DE
xlab	x-axis label
ylab	y-axis label
main	title of plot
...	arguments passed to barplot

Details

`create_timepoints_contrasts` creates the needed contrasts for comparing two groups for every timepoint in the format needed for `DE_timepoints` (i.e. `makeContrasts`, to which the contrasts are ultimately passed). The time points are determined by the meta data in the `moanin_object` provided by the user.

Value

This is a plotting function, and returns (invisibly) the results of `barplot`

Examples

```
data(exampleData)
moanin <- create_moanin_model(data=testData, meta=testMeta)
contrasts <- create_timepoints_contrasts(moanin, "C", "K")
deTimepoints <- DE_timepoints(moanin,
  contrasts=contrasts, use_voom_weights=FALSE)
perWeek_barplot(deTimepoints)
```

`plot_cdf_consensus` *Evaluate the consensus between sets of clusterings*

Description

Methods for evaluating the consensus between sets of clusterings, usually in the context of subsetting of the data or different numbers of clusters.

Usage

```
plot_cdf_consensus(labels)

get_auc_similarity_scores(labels, method = c("consensus", "nmi"))

plot_model_explorer(labels, colors = rainbow(length(labels)))
```

Arguments

<code>labels</code>	a list. Each element of the list is a matrix that gives the results of a clustering routine in each column (see <code>consensus_matrix</code>). Usually each column would be the result of running the clustering on a subsample or bootstrap resample of the data.
<code>method</code>	method for calculation of similarity for the AUC measure, one of "consensus" or "nmi". See details.
<code>colors</code>	a vector of colors, of length equal to the length of labels

Details

For each element of the list labels, plot_cdf_consensus calculates the consensus between the clusterings in the matrix, i.e. the number of times that pairs of rows are in the same cluster for different clusterings (columns) of the matrix using the [consensus_matrix](#) function. Then the set of values (the $N(N-1)$ values in the upper triangle of the matrix), are converted into a cdf function and plotted.

For each set of clusterings given by labels (i.e. for each matrix M which is an element of the list labels) get_auc_similarity_scores calculates a pairwise measure of similarity between the columns of M. These pairwise scores are plotted against their rank, and the final AUC measure is the area under this curve.

For method "consensus", the pairwise measure is given by calculating the consensus matrix using [consensus_matrix](#) with scale=FALSE. The consensus matrix is divided by the max of M.

For method "nmi", the pairwise value is the NMI value between each pair of columns of the matrix of clusterings using the [NMI](#) function.

Value

plot_cdf_consensus invisibly returns list of the upper triangle values, with the list of same length as that of labels.

get_auc_similarity_scores returns a vector, equal to length of the list labels, giving the AUC value for each element of labels.

This function is a plotting function does not return anything

See Also

[consensus_matrix](#), [NMI](#), [plot_cdf_consensus](#)

Examples

```
data(exampleData)
moanin <- create_moanin_model(data=testData,meta=testMeta)
#small function to run splines_kmeans on subsample of 50 genes
subsampleCluster<-function(){
  ind<-sample(1:nrow(moanin),size=50)
  km<-splines_kmeans(moanin[ind,],n_clusters=3)
  assign<-splines_kmeans_score_and_label(moanin, km,
    proportion_genes_to_label=1.0)$label
}
kmClusters1=replicate(10,subsampleCluster())
kmClusters2=replicate(10,subsampleCluster())
# Note, because of the small number of replicates (10),
# these plots are not representative of what to expect.
out<-plot_cdf_consensus(labels=list(kmClusters1,kmClusters2))
get_auc_similarity_scores(list(kmClusters1,kmClusters2))
plot_model_explorer(list(kmClusters1,kmClusters2))
```

plot_splines_data *Plotting splines*

Description

Plotting splines

Usage

```
## S4 method for signature 'Moanin,matrix'
plot_splines_data(
  object,
  data,
  colors = NULL,
  smooth = FALSE,
  legend = TRUE,
  legendArgs = NULL,
  subset_conditions = NULL,
  subset_data = NULL,
  simpleY = TRUE,
  centroid = NULL,
  scale_centroid = c("toData", "toCentroid", "none"),
  mar = c(2.5, 2.5, 3, 1),
  mfrow = NULL,
  addToPlot = NULL,
  ylab = "",
  xaxis = TRUE,
  yaxis = TRUE,
  xlab = "Time",
  ...
)

## S4 method for signature 'Moanin,numeric'
plot_splines_data(object, data, ...)

## S4 method for signature 'Moanin,data.frame'
plot_splines_data(object, data, ...)

## S4 method for signature 'Moanin,DataFrame'
plot_splines_data(object, data, ...)

## S4 method for signature 'Moanin,missing'
plot_splines_data(object, data, ...)
```

Arguments

object An object of class [Moanin](#), an object containing all related information for time course data and the splines model that will be used (if applicable). See [create_moanin_model](#)

	for more details.
data	matrix containing the data to be plotted, where each row of the data provided will be plotted as a separate plot. If missing, will rely on data in <code>assay(object)</code>
colors	vector, optional, default NULL. Vector of colors
smooth	boolean, optional, default: FALSE. Whether to smooth the centroids or not.
legend	boolean whether to include a legend (default:TRUE)
legendArgs	list of arguments to be passed to legend command (if legend=TRUE)
subset_conditions	list if provided, only plots the subset of conditions provided. Else, plots all conditions
subset_data	list if provided, only plots the subset of data (ie, the rows) provided. Can be any valid vector for subsetting a matrix. See details.
simpleY	boolean, if true, will plot all genes on same y-axis and minimize the annotation of the y axis to only label the axis in the exterior plots (the x-axis is always assumed to be the same across all plots and will always be simplified)
centroid	numeric vector (or matrix of 1 row) with data to use to fit the splines. If NULL, the splines plotted will be from the data.
scale_centroid	determines whether the centroid data given in centroid should be rescaled to match that of the data ("toData"), or the data scaled to match that of centroid ("toCentroid"), or simply plotted as is ("none").
mar	vector of margins to set the space around each plot (see par)
mfrow	a vector of integers of length 2 defining the grid of plots to be created (see par). If missing, the function will set a value.
addToPlot	A function that will be called after the plotting, allowing the user to add more to the plot.
ylab	label for the y-axis
xaxis	Logical, whether to add x-axis labels to plot (if FALSE can be manually created by user with call to <code>addToPlot</code>)
yaxis	Logical, whether to add y-axis labels to plot (if FALSE can be manually created by user with call to <code>addToPlot</code>)
xlab	label for the x-axis
...	arguments to be passed to the individual plot commands (Will be sent to all plot commands)

Details

If data is NULL, the data plotted will be from `assay(object)`, after log-transformation if `log_transform(object)=TRUE`.

If centroid is missing, then splines will be estimated (per group) for the the data in `data` – separately for each row of data. If centroid is provided, this data will be used to plot a spline function, and this same spline will be plotted for each row of data. This is useful, for example, in plotting cluster centroids over a series of genes.

If the user set `log_transform=TRUE` in the creation of the `Moanin` object, the data will be log transformed before plotting and calculating the spline fits.

Value

This function creates a plot and does not return anything to the user.

Examples

```
# First, load some data and create a moanin model
data(exampleData)
moanin <- create_moanin_model(data=testData,meta=testMeta,
  degrees_of_freedom=6)

# The moanin model contains all the information for plotting purposes. The
# plot_splines_data will automatically fit the splines from the
# information contained in the moanin model
genes <- c("NM_001042489", "NM_008725")
plot_splines_data(moanin, subset_data=genes,
mfrow=c(2, 2))
# By default, same axis for all genes. Can change with 'simpleY=FALSE'
plot_splines_data(moanin, subset_data=genes,
  smooth=TRUE, mfrow=c(2,2), simpleY=FALSE)

# The splines can also be smoothed
plot_splines_data(moanin, subset_data=genes,
  smooth=TRUE, mfrow=c(2, 2))
# You can provide different data (on same subjects),
# instead of data in moanin object
# (in which case moanin just provides grouping information)
plot_splines_data(moanin, data=1/assay(moanin), subset_data=genes,
  smooth=TRUE, mfrow=c(2, 2))

# You can also provide data to use for fitting splines to argument
# "centroid". This is helpful for overlaying centroids or predicted data
# Here we do a silly example, just to demonstrate syntax,
# where we use the data from the first gene as our centroid to fit a
# spline estimate, but plot data from genes 3-4
plot_splines_data(moanin, centroid=assay(moanin[1,]), subset_data=3:4,
  smooth=TRUE, mfrow=c(2,2))
```

pvalues_fisher_method *Fisher's method to combine pvalues*

Description

Combines all p-values per rows.

Usage

```
pvalues_fisher_method(pvalues)
```

Arguments

`pvalues` a matrix of pvalues, with columns corresponding to different tests or sources of p-values, and rows corresponding to the genes from which the p-values come.

Value

a vector of p-values, one for each row of `pvalues`, that is the result of Fisher's combined probability test applied to the p-values in that row.

Examples

```
data(exampleData)
moanin <- create_moanin_model(data=testData,meta=testMeta)
contrasts <- create_timepoints_contrasts(moanin,"C", "K")
deTimepoints=DE_timepoints(moanin,
  contrasts=contrasts, use_voom_weights=FALSE)
fisherPval=pvalues_fisher_method(
  deTimepoints[,grep("pval",colnames(deTimepoints))])
head(fisherPval)
```

rescale_values

Rescales rows of data to be between 0 and 1

Description

Rescales rows of data to be between 0 and 1

Usage

```
## S4 method for signature 'Moanin'
rescale_values(object, data = NULL, use_group = FALSE)

## S4 method for signature '`NULL`'
rescale_values(object, data)

## S4 method for signature 'missing'
rescale_values(object, ...)
```

Arguments

`object` a object of class `Moanin`, only needed if choose to rescale by grouping variable in the `moanin` object. If `NULL`, then data will be rescaled jointly across all observations.

`data` The matrix to rescale by row. If `NULL`, and `object` is given, data will be taken as `assay(object)` Each row should correspond to a gene or a centroid, and columns to samples.

`use_group` If true, then the data will be rescaled such that, for each row, all values associated to each group (defined by grouping variable of object) is between 0 and 1. For example, if column

... arguments passed to the matrix or Moanin method.

Details

If the user set `log_transform=TRUE` in the creation of the Moanin object, the data will be log transformed before rescaling

Value

rescaled y, such that for each row, the values are comprised between 0 and 1. Note that if `use_group=TRUE` and object is not NULL, the values associated to the columns of unique values of the grouping variable of object will be rescaled separately.

Examples

```
data(exampleData)
moanin <- create_moanin_model(data=testData, meta=testMeta)
# Can rescale data in Moanin object
allData <- rescale_values(moanin)
# Or provide different data and/or rescale within grouping variable
smallData <- rescale_values(moanin, data=testData[1:10,], use_group=TRUE)
```

<code>splines_kmeans</code>	<i>Performs splines clustering using K-means</i>
-----------------------------	--

Description

Performs splines clustering using K-means

Usage

```
## S4 method for signature 'Moanin'
splines_kmeans(
  object,
  n_clusters = 10,
  init = "kmeans++",
  n_init = 10,
  max_iter = 300,
  random_seed = .Random.seed[1],
  fit_splines = TRUE,
  rescale = TRUE
)
```

Arguments

object	An object of class <code>Moanin</code> , an object containing all related information for time course data and the splines model that will be used (if applicable). See <code>create_moanin_model</code> for more details.
n_clusters	int optional, default: 10
init	["kmeans++", "random", "optimal_init"]
n_init	int, optional, default: 10 Number of initialization to perform.
max_iter	int, optional, default: 300 Maximum number of iteration to perform
random_seed	int, optional, default: NULL. Passed to argument seed in <code>KMeans_rcpp</code> . If NULL (default), set to <code>.Random.seed[1]</code> .
fit_splines	boolean, optional, default: TRUE Whether to fit splines or not.
rescale	boolean, optional, default: TRUE Whether to rescale the data or not.

Details

If `Moanin` object's slot has `log_transform=TRUE`, then the data will be transformed by the function `log(x+1)` before applying splines and clustering.

Value

A list in the format returned by `KMeans_rcpp`, with the following elements added or changed:

- `centroids` The centroids are rescaled so that they range from 0-1
- `fit_splines` Logical, the value of `fit_splines` given to the function
- `rescale` The value of `rescale` given to the function

Examples

```
data(exampleData)
# Use the default options
moanin <- create_moanin_model(data=testData, meta=testMeta)
out <- splines_kmeans( moanin, n_clusters=5)
table(out$clusters)
```

splines_kmeans_predict, Moanin-method

Assign score and labels from raw data

Description

Assign score and labels from raw data

Usage

```
## S4 method for signature 'Moanin'
splines_kmeans_predict(
  object,
  kmeans_clusters,
  data = NULL,
  method = c("distance", "goodnessOfFit"),
  ...
)

## S4 method for signature 'Moanin'
splines_kmeans_score_and_label(
  object,
  kmeans_clusters,
  data = NULL,
  proportion_genes_to_label = 0.5,
  max_score = NULL,
  previous_scores = NULL,
  rescale_separately = FALSE
)
```

Arguments

<code>object</code>	the Moanin object that contains the basis functions used in creating the clusters
<code>kmeans_clusters</code>	List returned by splines_kmeans
<code>data</code>	the data to predict. If not given, will use <code>assay(object)</code> . If given, the number of columns of data must match that of <code>object</code>
<code>method</code>	If "distance", predicts based on distance of data to kmeans centroids. If "goodnessOfFit", is a wrapper to <code>splines_kmeans_score_and_label</code> , assigning labels based on goodness of fit, including any filtering.
<code>...</code>	arguments passed to <code>splines_kmeans_score_and_label</code>
<code>proportion_genes_to_label</code>	float, optional, default: 0.5 Percentage of genes to label. If <code>max_score</code> is provided, will label genes that are either in the top 'proportion_genes_to_label' or with a score below 'max_score'.
<code>max_score</code>	optional, default: Null When provided, will only label genes below that score. If NULL, ignore this option.
<code>previous_scores</code>	matrix of scores, optional. Allows user to give the matrix scores results from a previous run of <code>splines_kmeans_score_and_label</code> , and only redo the filtering (i.e. if want to change <code>proportion_genes_to_label</code> without rerunning the calculation of scores)
<code>rescale_separately</code>	logical, whether to score separately within grouping variable

Value

`splines_kmeans_predict` returns a vector giving the labels for the given data.

A list consisting of

- `label` the label or cluster assigned to each gene based on the cluster with the best (i.e. lowest) score, with no label given to genes that do not have a score lower than a specified quantity
- `scores` the matrix of size `n_cluster` x `n_genes`, containing for each gene and each cluster, the goodness of fit score
- `score_cutoff` The required cutoff for a gene receiving an assignment

Examples

```
data(exampleData)
moanin <- create_moanin_model(data=testData, meta=testMeta)
# Cluster on a subset of genes
kmClusters=splines_kmeans(moanin[1:50,],n_clusters=3)
# get scores on all genes
scores_and_labels <- splines_kmeans_score_and_label(object=moanin, kmClusters)
head(scores_and_labels$scores)
head(scores_and_labels$labels)
# should be same as above, only just the assignments
predictLabels1 <- splines_kmeans_predict(object=moanin, kmClusters,
  method="goodnessOfFit")
# Instead use distance to centroid:
predictLabels2 <- splines_kmeans_predict(object=moanin, kmClusters,
  method="distance")
```

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