Package ‘oppar’

April 4, 2024

Type Package
Title Outlier profile and pathway analysis in R
Version 1.30.0
Date 2016
Description The R implementation of mCOPA package published by Wang et al. (2012). Oppar provides methods for Cancer Outlier profile Analysis. Although initially developed to detect outlier genes in cancer studies, methods presented in oppar can be used for outlier profile analysis in general. In addition, tools are provided for gene set enrichment and pathway analysis.

biocViews Pathways, GeneSetEnrichment, SystemsBiology, GeneExpression

Software

Depends R (>= 3.3)
Imports Biobase, methods, GSEABase, GSVA
License GPL-2
LazyData true

Collate 'class_OPPARList.R' 'class_nuchar.R' 'getSampleOutlier.R'
'getSubtypeProbes.R' 'gsva.R' 'opa-generic-methods.R' 'oppar.R'
'utils.R' 'zzz.R'

RoxygenNote 5.0.1

Suggests knitr, rmarkdown, limma, org.Hs.eg.db, GO.db, snow, parallel

VignetteBuilder knitr

git_url https://git.bioconductor.org/packages/oppar

git_branch RELEASE_3_18

git_last_commit 3acac4f

git_last_commit_date 2023-10-24

Repository Bioconductor 3.18

Date/Publication 2024-04-03

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An ExpressionSet object containing trimmed GSE46141 data. The object contains gene expression measurements on local breast tumours and liver, lymph node, skin local-regional etc metastatic tumours. Contains 9503 features and 80 samples (ascite, bone, lung and skin samples were removed).

Usage

data(GSE46141)

Format

Contains gene expression matrix, phenotype (pData) and feature (fData) data:

- **ID**  In fData(e) – probe IDs
- **EntrezGeneID**  In fData(e) – Entrez Ids
- **GeneSymbol**  In fData(e) – Gene Symbols
- **geo_accession**  In pData(e) – GEO accession IDs
- **source_name_ch1**  In pData(e) – Sample information

Value

An ExpressionSet object
eset

Source


eset Tomlins et al. Prostate Cancer data (GEO: GSE6099)

Description

An ExpressionSet object containing microarray gene expression measurements on normal tissue and metastatic prostate cancer tumours, and the corresponding feature and phenotypic meta data.

Usage

data(tomlins)

Format

An ExpressionSet object with 10945 features and 86 samples.

- **title** In pData(eset) – Sample names
- **geo_accession** In pData(eset) – GEO accession numbers
- **characteristics.ch1** In pData(eset) – sample description
- **ID** In fData(eset) – probes IDs
  - **Gene.title** ...
  - **Gene.symbol** In fData(eset) – Gene Symbol
  - **Gene.ID** In fData(eset) – Entrez Gene IDs

Value

An ExpressionSet object

Source

getSampleOutlier

Retrieving outlier genes in samples

Description

Returns a list of outlier genes for each given sample

Usage

getSampleOutlier(profileMatrix, sample.name)

## S4 method for signature 'matrix,nuchar'
getSampleOutlier(profileMatrix, sample.name)

## S4 method for signature 'OPPARList,nuchar'
getSampleOutlier(profileMatrix, sample.name)

Arguments

profileMatrix A matrix of 0, -1 and 1 representing outlier genes in samples. Also an object of type OPPARList.
sample.name A character vector containing one or more sample names, or a numeric vector containing sample indices.

Value

A list of lists. For each given sample, the function return up-regulated and down-regulated outlier genes.

Methods (by class)

- profileMatrix = matrix, sample.name = nuchar: A method for getSampleOutlier
- profileMatrix = OPPARList, sample.name = nuchar: A method for getSampleOutlier

Examples

data(GSE46141)
library(Biobase)
group <- sapply(pData(bcm)$source_name_ch1, function(x) ifelse(x == "breast", 0, 1))
group <- factor(group)
bcm.opa <- opa(bcm, group=group)
# Outlier profile for sample "GSM1124929"
getSampleOutlier(bcm.opa, "GSM1124929")

# Also can use sample index, instead of sample name
getSampleOutlier(bcm.opa, 11)

# A vector of sample names/indices are accepted as well
getSubtypeProbes

getSampleOutlier(bcm.opa, c(1,2))
getSampleOutlier(bcm.opa, c("GSM1124929","GSM1124941"))

getSubtypeProbes

Retrieving outlier genes from a group of related samples

Description

Returns a list of genes that are outlier in a group of samples, such as samples from the same subtype.

Usage

getSubtypeProbes(profileMatrix, sample.names)

## S4 method for signature 'matrix,nuchar'
getSubtypeProbes(profileMatrix, sample.names)

## S4 method for signature 'OPPARList,nuchar'
getSubtypeProbes(profileMatrix, sample.names)

Arguments

profileMatrix A matrix of 0,1 and -1, representing outlier genes in samples. Also an object of type OPPARList.
sample.names A character vector containing sample names, or a numeric vector containing the indices of the samples.

Value

A list of lists. The sub-lists are up-regulated outlier genes, and down-regulated outlier genes.

Methods (by class)

- profileMatrix = matrix, sample.names = nuchar: A method for getSubtypeProbes with signature profileMatrix = matrix and sample.names = nuchar
- profileMatrix = OPPARList, sample.names = nuchar: A method for getSubtypeProbes with signature profileMatrix = OPPARList and sample.names = nuchar

Examples

data(GSE46141)
library(BiocBase)
group <- sapply(pData(bcm)$source_name_ch1, function(x) { ifelse(x == "breast", 0, 1)})
group <- factor(group)
bcm.opa <- opa(bcm, group=group)
# extracting liver samples
index <- which(pData(bcm)$source_name_ch1 == "liver")
samples <- rownames(pData(bcm)[index,])
samples <- match(samples, colnames(bcm.opa$profileMatrix))
samples <- Reduce(c, samples)
# liver subtype outlier profile
liver.subtype.outlier <- getSubtypeProbes(bcm.opa, samples)

---

**Description**

Gene Set Variation Analysis

**Usage**

gsva(expr, gset.idx.list, ...)  
## S4 method for signature 'ExpressionSet,list'
gsva(expr, gset.idx.list, annotation,  
method = c("gsva", "ssgsea", "zscore", "plage"), rnaseq = FALSE,  
asr.rank = FALSE, min.sz = 1, max.sz = Inf, no.bootstraps = 0,  
bootstrap.percent = 0.632, parallel.sz = 0, parallel.type = "SOCK",  
mx.diff = TRUE, tau = switch(method, gsva = 1, ssgsea = 0.25, NA),  
kernel = TRUE, ssgsea.norm = TRUE, verbose = TRUE,  
is.gset.list.up.down = FALSE)

## S4 method for signature 'ExpressionSet,GeneSetCollection'
gsva(expr, gset.idx.list, annotation, method = c("gsva", "ssgsea", "zscore", "plage"),  
rnaseq = FALSE, abs.ranking = FALSE, min.sz = 1, max.sz = Inf,  
no.bootstraps = 0, bootstrap.percent = 0.632, parallel.sz = 0,  
parallel.type = "SOCK", mx.diff = TRUE, tau = switch(method, gsva = 1,  
ssgsea = 0.25, NA), kernel = TRUE, ssgsea.norm = TRUE, verbose = TRUE,  
is.gset.list.up.down = FALSE)

## S4 method for signature 'matrix,GeneSetCollection'
gsva(expr, gset.idx.list, annotation,  
method = c("gsva", "ssgsea", "zscore", "plage"), rnaseq = FALSE,  
asr.rank = FALSE, min.sz = 1, max.sz = Inf, no.bootstraps = 0,  
bootstrap.percent = 0.632, parallel.sz = 0, parallel.type = "SOCK",  
mx.diff = TRUE, tau = switch(method, gsva = 1, ssgsea = 0.25, NA),  
kernel = TRUE, ssgsea.norm = TRUE, verbose = TRUE,  
is.gset.list.up.down = FALSE)

## S4 method for signature 'matrix,list'
gsva(expr, gset.idx.list, annotation,  
method = c("gsva", "ssgsea", "zscore", "plage"), rnaseq = FALSE,  
asr.rank = FALSE, min.sz = 1, max.sz = Inf, no.bootstraps = 0,
```r
gsva

bootstrap.percent = 0.632, parallel.sz = 0, parallel.type = "SOCK",
mx.diff = TRUE, tau = switch(method, gsva = 1, ssgsea = 0.25, NA),
kernel = TRUE, ssgsea.norm = TRUE, verbose = TRUE,
is.gset.list.up.down = FALSE)
```

**Arguments**

- `expr`  
  Gene expression data which can be given either as an `ExpressionSet` object or as a matrix of expression values where rows correspond to genes and columns correspond to samples.

- `gset.idx.list`  
  Gene sets provided either as a `list` object or a `GeneSetCollection` object.

- `...`  
  Other optional arguments.

- `annotation`  
  In the case of calling `gsva()` with expression data in a matrix and gene sets as a `GeneSetCollection` object, the `annotation` argument can be used to supply the name of the Bioconductor package that contains annotations for the class of gene identifiers occurring in the row names of the expression data matrix. By default `gsva()` will try to match the identifiers in `expr` to the identifiers in `gset.idx.list` just as they are, unless the `annotation` argument is set.

- `method`  
  Method to employ in the estimation of gene-set enrichment scores per sample. By default this is set to `gsva` (Hanzelmann et al, 2013) and other options are `ssgsea` (Barbie et al, 2009), `zscore` (Lee et al, 2008) or `plage` (Tomfohr et al, 2005). The latter two standardize first expression profiles into z-scores over the samples and, in the case of `zscore`, it combines them together as their sum divided by the square-root of the size of the gene set, while in the case of `plage` they are used to calculate the singular value decomposition (SVD) over the genes in the gene set and use the coefficients of the first right-singular vector as pathway activity profile.

- `rnaseq`  
  Flag to inform whether the input gene expression data comes from microarray (`rnaseq=FALSE`, default) or RNA-Seq (`rnaseq=TRUE`) experiments.

- `abs.ranking`  
  Flag to determine whether genes should be ranked according to their sign (`abs.ranking=FALSE`) or by absolute value (`abs.ranking=TRUE`). In the latter, pathways with genes enriched on either extreme (high or low) will be regarded as ‘highly’ activated.

- `min.sz`  
  Minimum size of the resulting gene sets.

- `max.sz`  
  Maximum size of the resulting gene sets.

- `no.bootstraps`  
  Number of bootstrap iterations to perform.

- `bootstrap.percent`  
  .632 is the ideal percent samples bootstrapped.

- `parallel.sz`  
  Number of processors to use when doing the calculations in parallel. This requires to previously load either the `parallel` or the `snow` library. If `parallel` is loaded and this argument is left with its default value (`parallel.sz=0`) then it will use all available core processors unless we set this argument with a smaller number. If `snow` is loaded then we must set this argument to a positive integer number that specifies the number of processors to employ in the parallel calculation.

- `parallel.type`  
  Type of cluster architecture when using `snow`.
mx.diff

Offers two approaches to calculate the enrichment statistic (ES) from the KS random walk statistic. \texttt{mx.diff=FALSE}: ES is calculated as the maximum distance of the random walk from 0. \texttt{mx.diff=TRUE} (default): ES is calculated as the magnitude difference between the largest positive and negative random walk deviations.

tau

Exponent defining the weight of the tail in the random walk performed by both the gsVA (Hanzelmann et al., 2013) and the ssgsea (Barbie et al., 2009) methods. By default, this \( \tau=1 \) when \texttt{method="gsVA"} and \( \tau=0.25 \) when \texttt{method="ssgsea"} just as specified by Barbie et al. (2009) where this parameter is called \text{alpha}.

kernel

Logical, set to \texttt{TRUE} when the GSVA method employs a kernel non-parametric estimation of the empirical cumulative distribution function (default) and \texttt{FALSE} when this function is directly estimated from the observed data. This last option is justified in the limit of the size of the sample by the so-called Glivenko-Cantelli theorem.

ssgsea.norm

Logical, set to \texttt{TRUE} (default) with \texttt{method="ssgsea"} runs the SSGSEA method from Barbie et al. (2009) normalizing the scores by the absolute difference between the minimum and the maximum, as described in their paper. When \texttt{ssgsea.norm=FALSE} this last normalization step is skipped.

verbose

Gives information about each calculation step. Default: \texttt{FALSE}.

is.gset.list.up.down

logical. Is the gene list divided into up/down sublists? Please note that it is important to name the up-regulated gene set list 'up', and the down-regulated gene set list to 'down', if this argument is used (e.g \texttt{gset = list(up = up_gset, down = down_gset)})

Value

returns gene set enrichment scores for each sample and gene set

Methods (by class)

- \texttt{expr = ExpressionSet, gset.idx.list = list}: Method for ExpressionSet and list
- \texttt{expr = ExpressionSet, gset.idx.list = GeneSetCollection}: Method for ExpressionSet and GeneSetCollection
- \texttt{expr = matrix, gset.idx.list = GeneSetCollection}: Method for matrix and GeneSetCollection
- \texttt{expr = matrix, gset.idx.list = list}: Method for matrix and list

See Also

maupin

Examples

```r
data("Maupin")
names(maupin)
geneSet<- maupin$sig$EntrezID  #Symbol  #EntrezID # both up and down genes:
up_sig<- maupin$sig[maupin$sig$upDown == "up",]
d_sig<- maupin$sig[maupin$sig$upDown == "down",]
u_geneSet<- up_sig$EntrezID  #Symbol # up_sig$Symbol  ## EntrezID
d_geneSet<- d_sig$EntrezID
es.dif <- gsva(maupin$data, list(up = u_geneSet, down= d_geneSet), mx.diff=1,
                  verbose=TRUE, abs.ranking=FALSE, is.gset.list.up.down=TRUE, parallel.sz = 1)$es.obs
```

### Description

A list consisting of two components: data: A matrix consisting of gene expression values on 3 control and 3 TGFb treated samples. sig: A TGFb gene signature. A dataframe containing a gene signature and information on whether genes in the signature are up or down regulated.

### Usage

```r
data(Maupin)
```

### Format

A list of two components:

- **M_Ctrl_R1** In data; Control Sample - Replicate 1
- **M_Ctrl_R2** In data; Control Sample - Replicate 2
- **M_Ctrl_R3** In data; Control Sample - Replicate 3
- **M_TGFb_R1** In data; Control Sample - Replicate 1
- **M_TGFb_R2** In data; Control Sample - Replicate 2
- **M_TGFb_R3** In data; Control Sample - Replicate 3
- **EntrezID** In sig; The Entrez IDs
- **Symbol** In sig; Gene symbols
- **upDown** In sig; Direction of regulation e.g. up, down

### Value

A list of 2

### Source

Outlier profile Analysis

Description

Returns a matrix with 0, -1 and 1 entries that describe outlier profiles in samples. The rows represent genes and the columns represent samples. -1 implies that the gene is a down-regulated outlier, 1 indicates an up-regulate outlier and 0 means that the gene is not an outlier in a sample.

Usage

opa(exprs.matrix, ...)

## S4 method for signature 'matrix'
opa(exprs.matrix, group, upper.quantile = 0.95, lower.quantile = 0.05)

## S4 method for signature 'ExpressionSet'
opa(exprs.matrix, group, upper.quantile = 0.95, lower.quantile = 0.05)

Arguments

exprs.matrix Gene expression data. Can be either a matrix or an object of type ExpressionSet.
...
Numeric. To supply values for upper.quantile and lower.quantile arguments if default values are going to be override.
group A vector of factors representing the groups to which each sample belong. This can be either a vector of 0s and 1s, or normal and cases.
upper.quantile Numeric. The cut-off for upper quantile when determining outliers. Default to 0.95
lower.quantile Numeric. The cut-off for lower quantile when determining outliers. Default to 0.05

Value

opa returns an object of type OPPARList. The outlier profiles are stored in profileMatrix and can be accessed using $. It it also possible to retrieve parameters used to run the outlier profile analysis, such as upper.quantile, lower.quantile via the $ operator.

Methods (by class)

- matrix: opa(exprs.matrix, group, lower.quantile = 0.05, upper.quantile = 0.95)
- ExpressionSet: opa(eset, group, lower.quantile = 0.05, upper.quantile = 0.95)
See Also


Examples

```r
# loading bcm object from GSE46141 dataset
data(GSE46141)
library(Biobase)

# defining the group variable. local breast tumors are the controls
# and the rest of the samples are the diseased samples
group <- sapply(pData(bcm)$source_name_ch1, function(x){ ifelse(x == "breast",0,1)})
group <- factor(group)
# running opa with default values (i.e upper.quantile = 0.95, lower.quantile = 0.05)
# the result is an object of type OPPARList
opa(bcm,group = group)
```

Description

The oppar package provides 3 main function for outlier profile analysis: `opa`, `getSampleOutlier` and `getSubtypeProbes`, and 1 function for pathway and gene set enrichment analysis, based on `gsva` function implemented in the GSV A package

`opa`

calculates the outlier profile matrix, using the method proposed in Wang et al. (2012) paper

`getSampleOutlier`

extracts outlier profile for individual samples

`getSubtypeProbes`

extracts outlier profile for a group of related samples, such as subtypes
Description

A S4 class for the output of OPPAR main function, opa.

Usage

## S4 method for signature 'OPPARList'

show(object)

## S4 method for signature 'OPPARList'

x$name

Arguments

object

An object of type OPPARList

x

Object of type OPPARList.

name

Name of the slot to access.

Value

returns the number of outlier features detected, the number of samples retained, and the parameters used to run the opa function

extracts slots from an object of type OPPARList.

Methods (by generic)

• show: A show method for objects of class OPPARList

• $: A method to extract slots in OPPARList

Slots

profileMatrix A matrix of 0, -1 and 1 representing outlier genes in samples

upper.quantile Numeric. The upper quantile cut-off for detection of outliers

lower.quantile Numeric. The lower quantile cut-off for detection of outliers

group A factor vector representing the group to which each sample belong

.Data matrix
## Index

* datasets
  - bcm, 2
  - eset, 3
  - maupin, 9
  - $,OPPARList-method (OPPARList-class), 12

  bcm, 2
  eset, 3
  getSampleOutlier, 4
  getSampleOutlier,matrix,nuchar-method
    (getSampleOutlier), 4
  getSampleOutlier,OPPARList,nuchar-method
    (getSampleOutlier), 4
  getSubtypeProbes, 5
  getSubtypeProbes,matrix,nuchar-method
    (getSubtypeProbes), 5
  getSubtypeProbes,OPPARList,nuchar-method
    (getSubtypeProbes), 5
  gsva, 6
  gsva,ExpressionSet,GeneSetCollection-method
    (gsva), 6
  gsva,ExpressionSet,list-method (gsva), 6
  gsva,matrix,GeneSetCollection-method
    (gsva), 6
  gsva,matrix,list-method (gsva), 6
  maupin, 9
  opa, 10
  opa,ExpressionSet-method (opa), 10
  opa,matrix-method (opa), 10
  oppar, 11
  oppar-package (oppar), 11
  OPPARList-class, 12

  show,OPPARList-method
    (OPPARList-class), 12