

Package ‘OGSA’

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Title Outlier Gene Set Analysis

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Description OGSA provides a global estimate of pathway deregulation in cancer subtypes by integrating the estimates of significance for individual pathway members that have been identified by outlier analysis.

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Depends R (>= 3.2.0)

Imports gplots(>= 2.8.0), limma(>= 3.18.13), Biobase

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License GPL (== 2)

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OGSA-package	<i>This package uses outlier statistics and gene set analysis to identify deregulated pathways.</i>
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Description

The package applies three versions of outlier statistics across multiple molecular data types to create a single estimate at the gene level for the number of outliers relative to normal controls. These gene estimates are used in gene set analysis to determine deregulated pathways. Visualization of outlier calls provide sample specific information on potential drivers of the gene set statistic.

Details

Package: OGSA
 Type: Package
 Version: 1.0
 Date: 2015-01-01
 License: Gnu Public License

Author(s)

Michael Ochs

Maintainer: Michael Ochs <ochsm@tcnj.edu>

References

Ochs MF, Farrar JE, Considine M, Wei Y, Meschinchi S, Arceci RJ. Outlier analysis and top scoring pair for integrated data analysis and biomarker discovery. *IEEE/ACM Trans Comput Biol and Bioinfo.* 2014; 11: 520-32.

cnv	<i>Copy number variation data</i>
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Description

Matrix of copy number variation data.

Usage

```
cnv
```

Format

Matrix of 2000 rows by 69 columns with copy number variation data

copaInt	<i>copaInt</i>
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Description

Counts outliers by Tibshirani-Hastie method by calling outCount after setting up list or by rank outlier method by calling outRank

Usage

```
copaInt(dataSet, phenotype, tails, thres = 0.05, method='Tibshirani',
corr=FALSE, offsets=NULL)
```

Arguments

dataSet	Set of matrices of molecular data
phenotype	Vector of 1 for case, 0 for control
tails	Vector equal to number of matrices with values left or right for where to find outliers
thres	alpha value
method	Tibshirani , Rank
corr	Whether to correct for normal outliers
offsets	A vector equal to the number of matrices which sets the minimum value relative to normal to call outlier (corrected rank only)

Value

A vector with outlier counts by gene

References

Ochs, M. F., Farrar, J. E., Considine, M., Wei, Y., Meshinchi, S., & Arceci, R. J. (n.d.). Outlier Analysis and Top Scoring Pair for Integrated Data Analysis and Biomarker Discovery. *IEEE/ACM Transactions on Computational Biology and Bioinformatics*, 1-1. doi:10.1109/tcbb.2013.153

Examples

```

data(ExampleData)

#Set up phenotype
phenotype <- pheno
names(phenotype) <- colnames(cnv)

#set up values for expr-meth-cnv in that order
tailLRL <- c('left', 'right', 'left')

#setup dataSet
dataSet <- list(expr, meth, cnv)

tibLRL <- copaInt(dataSet, phenotype, tails=tailLRL)

```

copaIntE

copaIntE

Description

Counts outliers by Tibshirani-Hastie method by calling outCount after setting up list or by rank outlier method by calling outRank

Usage

```

copaIntE(expressionSet, tails, thres = 0.05, method='Tibshirani',
corr=FALSE, offsets=NULL)

```

Arguments

expressionSet	object containing Set of matrices of molecular data and phenotype data (1 for case, 0 for control)
tails	Vector equal to number of matrices with values left or right for where to find outliers
thres	alpha value
method	Tibshirani , Rank
corr	Whether to correct for normal outliers
offsets	A vector equal to the number of matrices which sets the minimum value relative to normal to call outlier (corrected rank only)

Value

A vector with outlier counts by gene

References

Ochs, M. F., Farrar, J. E., Considine, M., Wei, Y., Meshinchi, S., & Arceci, R. J. (n.d.). Outlier Analysis and Top Scoring Pair for Integrated Data Analysis and Biomarker Discovery. *IEEE/ACM Transactions on Computational Biology and Bioinformatics*, 1-1. doi:10.1109/tcbb.2013.153

Examples

```

data(ExampleData)

library(Biobase)
# building the Annotated Data Frame
phenoData <- AnnotatedDataFrame(
  data.frame(
    type = factor(x = pheno, labels = c("Control", "Case")),
    row.names = colnames(expr)
  )
)
# build environment
inputData <- list2env(list(exprs = expr, meth = meth, cnv = cnv))

# build expressionSet - other information can be added here
expressionSet <- ExpressionSet(inputData, phenoData)

# set up values for expr-meth-cnv in that order
tailLRL <- c('left', 'right', 'left')

tibLRL <- copaIntE(expressionSet, tails=tailLRL)

```

copaStat

copaStat

Description

Calculates outlier statistics by the Tibshirani-Hastie method

Usage

```
copaStat (data, phenotype, tail='right', perms=100, permType='array')
```

Arguments

data	A matrix of nGene by nSample
phenotype	A vector of 0s and 1s of length nSample, where 1 = case, 0 = control
tail	Indicates whether outliers are up (right) or down (left) outliers
perms	The number of permutations
permType	By all on array or by gene, if by gene increase perms significantly and plan on lots of time; in theory array should be fine as genes are rescaled

Value

A vector with outlier counts by gene

References

Ochs, M. F., Farrar, J. E., Considine, M., Wei, Y., Meshinchi, S., & Arceci, R. J. (n.d.). Outlier Analysis and Top Scoring Pair for Integrated Data Analysis and Biomarker Discovery. *IEEE/ACM Transactions on Computational Biology and Bioinformatics*, 1-1. doi:10.1109/tcbb.2013.153

Examples

```

data(ExampleData)

#Set up phenotype
phenotype <- pheno
names(phenotype) <- colnames(cnv)

#set up values for expr-meth-cnv in that order
tailLRL <- c('left', 'right', 'left')

#setup dataList
dataSet <- list(expr, meth, cnv)

data <- dataSet[[1]]

tibL <- copaStat(data, phenotype, tail='right', perms=100, permType='array')

```

expr	<i>expr</i>
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Description

Matrix of expression data

Usage

expr

Format

Matrix of 2000 rows by 69 columns with expression data

meth	<i>meth</i>
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Description

Matrix of methylation data

Usage

meth

Format

Matrix of 2000 rows by 69 columns with methylation data

outCallRank	<i>outCallRank</i>
-------------	--------------------

Description

Counts outliers by the Ghosh method and generates list objects with all outliers noted

Usage

```
outCallRank (dataSet, phenotype, thres= 0.05, tail='right', corr=FALSE,
offsets=NULL, names=NULL)
```

Arguments

dataSet	Set of matrices of molecular data
phenotype	A vector of 0s and 1s of length nSample, where 1 = case, 0 = control
thres	Alpha value
tail	A vector equal to the number of matrices with values left or right for where to find outliers
corr	Whether to correct for normal outliers
offsets	A vector equal to the number of matrices which sets the minimum value relative to normal to call outlier (corrected rank only)
names	A vector equal to the number of matrices to name molecular type of data (e.g., CNV)

Value

A list with all specific outlier calls for each molecular type in each case sample

References

Ochs, M. F., Farrar, J. E., Considine, M., Wei, Y., Meshinchi, S., & Arceci, R. J. (n.d.). Outlier Analysis and Top Scoring Pair for Integrated Data Analysis and Biomarker Discovery. *IEEE/ACM Transactions on Computational Biology and Bioinformatics*, 1-1. doi:10.1109/tcbb.2013.153

D. Ghosh. (2010). Discrete Nonparametric Algorithms for Outlier Detection with Genomic Data. *J. Biopharmaceutical Statistics*, 20(2), 193-208.

Examples

```
data(ExampleData)

#set up dataSet
dataSet <- list(expr, meth,cnv)

# Set up Phenotype
phenotype <- pheno
names(phenotype) <- colnames(cnv)

# set up values for expr-meth-cnv in that order
tailLRL <- c('left', 'right', 'left')
```

```
outRankLRL <- outCallRank(dataSet, phenotype, names=c('Expr',
  'Meth', 'CNV'), tail=tailLRL)
```

outCallRankE	<i>outCallRankE</i>
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Description

Counts outliers by the Ghosh method and generates list objects with all outliers noted

Usage

```
outCallRankE (expressionSet, thres= 0.05, tail='right', corr=FALSE,
  offsets=NULL, names=NULL)
```

Arguments

expressionSet	object containing Set of matrices of molecular data and phenotype data (1 for case, 0 for control)
thres	Alpha value
tail	A vector equal to the number of matrices with values left or right for where to find outliers
corr	Whether to correct for normal outliers
offsets	A vector equal to the number of matrices which sets the minimum value relative to normal to call outlier (corrected rank only)
names	A vector equal to the number of matrices to name molecular type of data (e.g., CNV)

Value

A list with all specific outlier calls for each molecular type in each case sample

References

Ochs, M. F., Farrar, J. E., Considine, M., Wei, Y., Meshinchi, S., & Arceci, R. J. (n.d.). Outlier Analysis and Top Scoring Pair for Integrated Data Analysis and Biomarker Discovery. *IEEE/ACM Transactions on Computational Biology and Bioinformatics*, 1-1. doi:10.1109/tcbb.2013.153

D. Ghosh. (2010). Discrete Nonparametric Algorithms for Outlier Detection with Genomic Data. *J. Biopharmaceutical Statistics*, 20(2), 193-208.

Examples

```
data(ExampleData)

library(Biobase)
# building the Annotated Data Frame
phenoData <- AnnotatedDataFrame(
  data.frame(
    type = factor(x = pheno, labels = c("Control", "Case")),
    row.names = colnames(expr)
```



```

    )
  )
  # build environment
  inputData <- list2env(list(exprs = expr, meth = meth, cnv = cnv))

  # build expressionSet - other information can be added here
  expressionSet <- ExpressionSet(inputData, phenoData)

  # set up values for for the tails in the order that they are exported,
  # for example:
  tailLRL <- c('left', 'right', 'left')

  outRankLRL <- outCallRankE(expressionSet, names=c('Expr', 'Meth', 'CNV'),
                             tail=tailLRL)

```

outCallTib

outCallTib

Description

Counts outliers by the Tibshirani and Hastie method and generates a list object with all outliers noted

Usage

```
outCallTib (dataSet, phenotype, tail='right', corr=FALSE, names=NULL)
```

Arguments

dataSet	Set of matrices of molecular data
phenotype	A vector of 0s and 1s of length nSample, where 1 = case, 0 = control
tail	Vector equal to number of matrices with values 'left' or 'right' for where to find outliers
corr	whether to correct for normal outliers ONLY for compatibility, since method does not allow determining specific changes in cases, it will just print message if corr = TRUE
names	Vector equal to number of matrices to name molecular type of data (e.g., 'CNV').

Value

A list with all specific outlier calls for each molecular type in each case sample

References

Ochs, M. F., Farrar, J. E., Considine, M., Wei, Y., Meshinchi, S., & Arceci, R. J. (n.d.). Outlier Analysis and Top Scoring Pair for Integrated Data Analysis and Biomarker Discovery. *IEEE/ACM Transactions on Computational Biology and Bioinformatics*, 1-1. doi:10.1109/tcbb.2013.153

D. Ghosh. (2010). Discrete Nonparametric Algorithms for Outlier Detection with Genomic Data. *J. Biopharmaceutical Statistics*, 20(2), 193-208.

Examples

```

data(ExampleData)
data('KEGG_BC_GS')

# Set up dataSet
dataSet <- list(expr, meth, cnv)

# Set up Phenotype
phenotype <- pheno
names(phenotype) <- colnames(cnv)

# set up values for expr-meth-cnv in that order
tailLRL <- c('left', 'right', 'left')

outTibLRL <- outCallTib(dataSet, phenotype, names=c('Expr', 'Meth', 'CNV'), tail=tailLRL)

```

outCallTibE

outCallTibE

Description

Counts outliers by the Tibshirani and Hastie method and generates a list object with all outliers noted

Usage

```
outCallTibE (expressionSet, tail='right', corr=FALSE, names=NULL)
```

Arguments

expressionSet	ExpressionSet object containing sets of data and phenotype information
tail	Vector equal to number of matrices with values 'left' or 'right' for where to find outliers
corr	whether to correct for normal outliers ONLY for compatibility, since method does not allow determining specific changes in cases, it will just print message if corr = TRUE
names	Vector equal to number of matrices to name molecular type of data (e.g., 'CNV').

Value

A list with all specific outlier calls for each molecular type in each case sample

References

Ochs, M. F., Farrar, J. E., Considine, M., Wei, Y., Meshinchi, S., & Arceci, R. J. (n.d.). Outlier Analysis and Top Scoring Pair for Integrated Data Analysis and Biomarker Discovery. *IEEE/ACM Transactions on Computational Biology and Bioinformatics*, 1-1. doi:10.1109/tcbb.2013.153

D. Ghosh. (2010). Discrete Nonparametric Algorithms for Outlier Detection with Genomic Data. *J. Biopharmaceutical Statistics*, 20(2), 193-208.

Examples

```

data(ExampleData)
data('KEGG_BC_GS')

library(Biobase)
# building the Annotated Data Frame
phenoData <- AnnotatedDataFrame(
  data.frame(
    type = factor(x = pheno, labels = c("Control", "Case")),
    row.names = colnames(expr)
  )
)
# build environment
inputData <- list2env(list(exprs = expr, meth = meth, cnv = cnv))

# build expressionSet - other information can be added here
expressionSet <- ExpressionSet(inputData, phenoData)

# set up values for for the tails in the order that they are exported, for example:
tailLRL <- c('left', 'right', 'left')

outTibLRL <- outCallTibE(expressionSet, names=c('Expr', 'Meth', 'CNV'), tail=tailLRL)

```

outCount

outCount

Description

Counts outliers by the Tibshirani and Hastie method. Adds the ability to subtract for outliers in the normals using `corr = TRUE`

Usage

```
outCount (data, phenotype, tail='right', corr=FALSE)
```

Arguments

<code>data</code>	A matrix of nGene by nSample
<code>phenotype</code>	A vector of 0s and 1s of length nSample, where 1 = case, 0 = control
<code>tail</code>	Indicates whether outliers are up (right) or down (left) outliers
<code>corr</code>	Whether to correct for normal outliers

Value

A vector with outlier counts by gene

References

Ochs, M. F., Farrar, J. E., Considine, M., Wei, Y., Meshinchi, S., & Arceci, R. J. (n.d.). Outlier Analysis and Top Scoring Pair for Integrated Data Analysis and Biomarker Discovery. *IEEE/ACM Transactions on Computational Biology and Bioinformatics*, 1-1. doi:10.1109/tcbb.2013.153

Examples

```

data(ExampleData)
# Set up Phenotype
phenotype <- pheno
names(phenotype) <- colnames(cnv)

#set up datalist
dataSet <- list(expr,meth,cnv)

# set up values for expr-meth-cnv in that order
tailLRL <- c('left', 'right', 'left')

outTibLRL <- outCallTib(dataSet, phenotype=pheno,
                       names=c('Expr', 'Meth', 'CNV'), tail=tailLRL)

```

outMap

outMap

Description

Creates PDF color map of where outliers occur coded for molecular type

Usage

```

outMap (outList, geneList, hmName = 'PatSpecMap.pdf', plotName =
'Outliers', truncGene = FALSE, clust=FALSE)

```

Arguments

outList	List with all outliers generated by outCallRank or outCallTib
geneList	Gene set to compare against
hmName	Name for PDF output file
plotName	Header for plot
truncGene	if TRUE, only include genes that have outlier in the plot, default is all genes in gene set
clust	If TRUE, clusters data and produces dendrograms

Value

A matrix used for generating heatmap

References

Ochs, M. F., Farrar, J. E., Considine, M., Wei, Y., Meshinchi, S., & Arceci, R. J. (n.d.). Outlier Analysis and Top Scoring Pair for Integrated Data Analysis and Biomarker Discovery. *IEEE/ACM Transactions on Computational Biology and Bioinformatics*, 1-1. doi:10.1109/tcbb.2013.153

Examples

```

data(ExampleData)
data('KEGG_BC_GS')

# Set up Phenotype
phenotype <- pheno
names(phenotype) <- colnames(cnv)

#set up datalist
dataSet <- list(expr,meth,cnv)

# set up values for expr-meth-cnv in that order
tailLRL <- c('left', 'right', 'left')

outTibLRL <- outCallTib(dataSet, phenotype=pheno,
                       names=c('Expr', 'Meth', 'CNV'), tail=tailLRL)

# put in your pathways here
pdgfB <- pathGS$'BIOCARTA_PDGF_PATHWAY'
outMap(outTibLRL, pdgfB, hmName='BC_PDGF_TIB.pdf', plotName='PDGF
Outlier T-H LRL Calls')

```

outRank

outRank

Description

Counts outliers by the Ghosh method.

Usage

```
outRank (dataSet, phenotype, thres= 0.05, tail='right', corr=FALSE,
offsets=NULL)
```

Arguments

dataSet	Set of matrices of molecular data
phenotype	Vector of 1 for case, 0 for control
thres	Alpha value
tail	Vector equal to number of matrices with values 'left' or 'right' for where to find outliers
corr	Whether to correct for normal outliers
offsets	Vector equal to number of matrices which sets minimum value relative to normal to call outlier (corrected rank only)

Value

A vector with outlier counts by gene

References

- Ochs, M. F., Farrar, J. E., Considine, M., Wei, Y., Meshinchi, S., & Arceci, R. J. (n.d.). Outlier Analysis and Top Scoring Pair for Integrated Data Analysis and Biomarker Discovery. *IEEE/ACM Transactions on Computational Biology and Bioinformatics*, 1-1. doi:10.1109/tcbb.2013.153
- D. Ghosh. (2010). Discrete Nonparametric Algorithms for Outlier Detection with Genomic Data. *J. Biopharmaceutical Statistics*, 20(2), 193-208.

Examples

```
data(ExampleData)

# Set up Phenotype
phenotype <- pheno
names(phenotype) <- colnames(cnv)

#set up dataSet
dataSet <- list(expr, meth, cnv)

# set up values for expr-meth-cnv in that order
tailLRL <- c('left', 'right', 'left')

outRankLRL <- outRank(dataSet, phenotype, thres= 0.05, tail=tailLRL,
                      corr=FALSE, offsets=NULL)
```

pathGS

Gene set defined by BioCarta pathway

Description

A large list containing gene set from the BioCarta pathway

Usage

```
pathGS
```

Format

List of 403 elements

Details

Contained in KEGG_BC_GS data frame

Source

Kanehisa, M. (2002). The KEGG databases at GenomeNet. *Nucleic Acids Research*, 30(1), 42-46. doi:10.1093/nar/30.1.42

pheno	<i>pheno</i>
-------	--------------

Description

Vector of phenotype data

Usage

pheno

Format

A vector of 0s and 1s of length 69, where 1 = tumor, 0 = normal

testGScogps	<i>testGScogps</i>
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Description

Performs gene set test on outlier counts

Usage

testGScogps (outlierCts, geneSets)

Arguments

outlierCts	Vector with gene names and outlier counts
geneSets	List of gene sets

Value

A vector with rank sum gene set statistics

References

Ochs, M. F., Farrar, J. E., Considine, M., Wei, Y., Meshinchi, S., & Arceci, R. J. (n.d.). Outlier Analysis and Top Scoring Pair for Integrated Data Analysis and Biomarker Discovery. *IEEE/ACM Transactions on Computational Biology and Bioinformatics*, 1-1. doi:10.1109/tcbb.2013.153

Examples

```
## Not run:
data(ExampleData)
data('_BC_GS')

#Set up your phenotype
phenotype <- rep(0, 69)
phenotype[annot[, 3] == 'Event'] <- 1
names(phenotype) <- rownames(annot)

# set up values for expr-meth-cnv in that order
tailLRL <- c('left', 'right', 'left')

dataSet <- list(expr, meth, cnv)

tibLRLcorr <- copaInt(dataSet, phenotype, tails=tailLRL, corr=TRUE)
gsTibLRLcorr <- testGScogps(tibLRLcorr, pathGS)

## End(Not run)
```


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