

# Package ‘OGSA’

April 15, 2017

**Version** 1.4.0

**Date** 2015-06-25

**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

**LazyData** true

**License** GPL (== 2)

**biocViews** GeneExpression, Microarray, CopyNumberVariation

**NeedsCompilation** no

## R topics documented:

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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> |
|-----|-----------------------------------|

---

### 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> |
|---------|----------------|

---

**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> |
|------|-------------|

---

**Description**

Matrix of expression data

**Usage**

expr

**Format**

Matrix of 2000 rows by 69 columns with expression data

---

|      |             |
|------|-------------|
| meth | <i>meth</i> |
|------|-------------|

---

**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

*outCallRankE*

---

**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 | <i>outCallTib</i> |
|------------|-------------------|

---

### 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 | <i>outCallTibE</i> |
|-------------|--------------------|

---

**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 | <i>outCount</i> |
|----------|-----------------|

---

### 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=phenotype,
                       names=c('Expr', 'Meth', 'CNV'), tail=tailLRL)
```

---

|        |               |
|--------|---------------|
| outMap | <i>outMap</i> |
|--------|---------------|

---

### 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 | <i>outRank</i> |
|---------|----------------|

---

## 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 | <i>Gene set defined by BioCarta pathway</i> |
|--------|---|

---

### 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> |
|-------------|--------------------|

---

**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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