## Package 'gsean'

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Type Package

Title Gene Set Enrichment Analysis with Networks

**Description** Biological molecules in a living organism seldom work individually. They usually interact each other in a cooperative way. Biological process is too complicated to understand without considering such interactions. Thus, network-based procedures can be seen as powerful methods for studying complex process. However, many methods are devised for analyzing individual genes. It is said that techniques based on biological networks such as gene coexpression are more precise ways to represent information than those using lists of genes only. This package is aimed to integrate the gene expression and biological network. A biological network is constructed from gene expression data and it is used for Gene Set Enrichment Analysis.

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Suggests SummarizedExperiment, knitr, plotly, WGCNA, rmarkdown

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gsean-package

Gene Set Enrichment Analysis with Networks

#### **Description**

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Biological molecules in a living organism seldom work individually. They usually interact each other in a cooperative way. Biological process is too complicated to understand without considering such interactions. Thus, network-based procedures can be seen as powerful methods for studying complex process. However, many methods are devised for analyzing individual genes. It is said that techniques based on biological networks such as gene co-expression are more precise ways to represent information than those using lists of genes only. This package is aimed to integrate the gene expression and biological network. A biological network is constructed from gene expression data and it is used for Gene Set Enrichment Analysis.

#### **Details**

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#### Author(s)

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Gene Set Enrichment Analysis with centrality measure

#### **Description**

GSEA is performed with centrality measure

#### Usage

#### **Arguments**

geneset list of gene sets

x Named vector of gene-level statistics. Names should be the same as in gene sets.

adjacency adjacency matrix

pseudo pseudo number for log2 transformation (default: 1)

nperm number of permutations (default: 1000)

centrality centrality measure, degree centrality or node strength is default

weightParam weight parameter value for the centrality measure, equally weight if weight-

Param = 0 (default: 1)

minSize minimal size of a gene set (default: 1)
maxSize maximal size of a gene set (default: Inf)
gseaParam GSEA parameter value (default: 1)

nproc see fgsea::fgsea BPPARAM see fgsea::fgsea

#### Value

GSEA result

#### Author(s)

Dongmin Jung

#### See Also

fgsea::fgsea

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#### **Examples**

```
data(examplePathways)
data(exampleRanks)
exampleRanks <- exampleRanks[1:100]
adjacency <- diag(length(exampleRanks))
rownames(adjacency) <- names(exampleRanks)
set.seed(1)
result.GSEA <- centrality_gsea(examplePathways, exampleRanks, adjacency)</pre>
```

exprs2adj

Convert gene expression data to adjacency matrix by using correlation coefficients

## Description

A biological network is constructed from gene expression data and it is used for Gene Set Enrichment Analysis.

#### Usage

```
exprs2adj(x, pseudo = 1, ...)
```

## Arguments

x gene expression data
 pseudo pseudo number for log2 transformation (default: 1)
 ... additional parameters for correlation; see WGCNA::cor

#### Value

adjacency matrix

## Author(s)

Dongmin Jung

#### See Also

```
fgsea::fgsea, WGCNA::cor
```

## **Examples**

```
data(exampleRanks)
Names <- names(exampleRanks)
exprs <- matrix(rnorm(10*length(exampleRanks)), ncol = 10)
adjacency <- exprs2adj(exprs)</pre>
```

GO\_dme 5

GO\_dme

Gene Ontology terms with gene ID for Drosophila melanogaster

#### Description

The data set contains all Gene Ontology terms for Drosophila melanogaster and genes are identified by gene ID. There are 2823 categories.

#### Usage

GO\_dme

#### **Format**

a list of gene sets

#### Value

GO gene sets

#### Author(s)

Dongmin Jung

#### Source

http://www.go2msig.org/cgi-bin/prebuilt.cgi?taxid=7227

#### **Examples**

```
load(system.file("data", "GO_dme.rda", package = "gsean"))
```

gsean

Gene Set Enrichment Analysis with Networks

#### **Description**

GSEA or ORA is performed with networks from gene expression data

#### Usage

```
gsean(geneset, x, exprs, pseudo = 1, threshold = 0.99, nperm = 1000,
    centrality = function(x) rowSums(abs(x)), weightParam = 1,
    minSize = 1, maxSize = Inf, gseaParam = 1, nproc = 0,
    BPPARAM = NULL, corParam = list(), tmax = 10, ...)
```

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#### **Arguments**

geneset list of gene sets

x Named vector of gene-level statistics for GSEA or set of genes for ORA. Names

should be the same as in gene sets.

exprs gene expression data

pseudo pseudo number for log2 transformation (default: 1)

threshold threshold of correlation for nodes to be considered neighbors for ORA (default:

0.99)

nperm number of permutations (default: 1000)

centrality centrality measure, degree centrality or node strength is default

weightParam weight parameter value for the centrality measure, equally weight if weight-

Param = 0 (default: 1)

minSize minimal size of a gene set (default: 1)
maxSize maximal size of a gene set (default: Inf)
gseaParam GSEA parameter value (default: 1)

nproc see fgsea::fgsea
BPPARAM see fgsea::fgsea

corParam additional parameters for correlation; see WGCNA::cor

tmax maximum number of iterations for label propagtion (default: 10)
... additional parameters for label propagation; see RANKS::label.prop

#### Value

GSEA result

#### Author(s)

Dongmin Jung

#### See Also

```
exprs2adj, label_prop_gsea, centrality_gsea
```

#### **Examples**

```
data(examplePathways)
data(exampleRanks)
exampleRanks <- exampleRanks[1:100]
Names <- names(exampleRanks)
exprs <- matrix(rnorm(10*length(exampleRanks)), ncol = 10)
rownames(exprs) <- names(exampleRanks)
set.seed(1)
result.GSEA <- gsean(examplePathways, exampleRanks, exprs)</pre>
```

KEGG\_hsa 7

KEGG\_hsa

KEGG pathways with gene symbol for human

## Description

The data set contains 186 KEGG pathways for Drosophila melanogaster and genes are identified by gene symbol.

#### Usage

KEGG\_hsa

#### **Format**

a list of gene sets

#### Value

KEGG gene sets

#### Author(s)

Dongmin Jung

#### Source

http://software.broadinstitute.org/gsea/msigdb/collections.jsp

## Examples

```
load(system.file("data", "KEGG_hsa.rda", package = "gsean"))
```

label\_prop\_gsea

Over-representaion analysis with the label propagation algorithm

## Description

ORA is performed by GSEA with the label propagation algorithm

#### Usage

8 label\_prop\_gsea

### **Arguments**

geneset list of gene sets

x set of genes

adjacency adjacency matrix

threshold threshold of correlation for nodes to be considered neighbors (default: 0.99)

nperm number of permutations (default: 1000)
minSize minimal size of a gene set (default: 1)
maxSize maximal size of a gene set (default: Inf)
gseaParam GSEA parameter value (default: 1)

nproc see fgsea::fgsea BPPARAM see fgsea::fgsea

... additional parameters for label propagation; see RANKS::label.prop

#### Value

GSEA result

#### Author(s)

Dongmin Jung

#### See Also

fgsea::fgsea

## **Examples**

```
data(examplePathways)
data(exampleRanks)
exampleRanks <- exampleRanks[1:100]
geneNames <- names(exampleRanks)
set.seed(1)
x <- sample(geneNames, 10)
adjacency <- diag(length(exampleRanks))
rownames(adjacency) <- geneNames
result.GSEA <- label_prop_gsea(examplePathways, x, adjacency)</pre>
```

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