

Package ‘miRSM’

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Type Package

Title Inferring miRNA sponge modules in heterogeneous data

Version 2.0.0

Description The package aims to identify miRNA sponge or ceRNA modules in heterogeneous data. It provides several functions to study miRNA sponge modules at single-sample and multi-sample levels, including popular methods for inferring gene modules (candidate miRNA sponge or ceRNA modules), and two functions to identify miRNA sponge modules at single-sample and multi-sample levels, as well as several functions to conduct modular analysis of miRNA sponge modules.

Depends R (>= 3.5.0)

License GPL-3

URL <https://github.com/zhangjunpeng411/miRSM>

Encoding UTF-8

biocViews GeneExpression, BiomedicalInformatics, Clustering, GeneSetEnrichment, Microarray, Software, GeneRegulation, GeneTarget

RoxygenNote 7.3.1

Imports WGCNA, flashClust, dynamicTreeCut, GFA, igraph, linkcomm, MCL, fabia, NMF, biclust, iBBiG, BicARE, isa2, s4vd, BiBitR, rqubic, Biobase, PMA, stats, dbscan, subspace, mclust, SOMbrero, ppclust, Repp, utils, SummarizedExperiment, GSEABase, org.Hs.eg.db, clusterProfiler, ReactomePA, DOSE, MatrixCorrelation, energy

Suggests BiocStyle, knitr, rmarkdown, testthat

VignetteBuilder knitr

BugReports <https://github.com/zhangjunpeng411/miRSM/issues>

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BRCA_genes

BRCA genes

Description

BRCA genes

Format

BRCA_genes: A SummarizedExperiment object with 4819 BRCA related genes (including lncRNAs and mRNAs).

Details

The BRCA related lncRNAs are from LncRNADisease v2.0, Lnc2Cancer v2.0 and MNDR v2.0. The BRCA related mRNAs are from DisGeNET v5.0 and COSMIC v86.

References

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- Pinero J, Bravo A, Queralt-Rosinach N, Gutierrez-Sacristan A, Deu-Pons J, Centeno E, Garcia-Garcia J, Sanz F, Furlong LI. (2017) "DisGeNET: a comprehensive platform integrating information on human disease-associated genes and variants". *Nucleic Acids Res.*, 45, D833-D839.

ceRExp

ceRNA expression data

Description

ceRNA expression data

Format

ceRExp: A SummarizedExperiment object with 72 BRCA and 72 normal samples (rows) and 305 lncRNAs (columns).

Details

The matched breast invasive carcinoma (BRCA) miRNA, lncRNA and mRNA expression data is obtained from TCGA (<http://cancergenome.nih.gov/>). lncRNA expression data is regarded as ceRNA expression data. The data focuses on 72 individuals for which the complete sets of tumor and matched normal (i.e., normal tissue taken from the same patient) profiles are available. A lncRNA which has missing values in more than 10 are imputed using the k-nearest neighbours (KNN) algorithm from the impute R package. We use the limma R package to infer differentially expressed lncRNAs between tumour and normal samples. After the analysis, we select top 305 lncRNAs which are differentially expressed at a significant level (adjusted p-value < 1E-02, adjusted by Benjamini & Hochberg method).

`cor_binary`*cor_binary*

Description

Generation of positively correlated binary matrix between ceRNAs, or ceRNAs and mRNAs

Usage

```
cor_binary(  
  ceExp,  
  mRExp = NULL,  
  cor.method = "pearson",  
  pos.p.value.cutoff = 0.01  
)
```

Arguments

<code>ceExp</code>	A SummarizedExperiment object. ceRNA expression data: rows are samples and columns are ceRNAs.
<code>mRExp</code>	NULL (default) or a SummarizedExperiment object. mRNA expression data: rows are samples and columns are mRNAs.
<code>cor.method</code>	The method of calculating correlation selected, including 'pearson' (default), 'kendall', 'spearman'.
<code>pos.p.value.cutoff</code>	The significant p-value cutoff of positive correlation.

Value

A binary matrix.

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

References

Langfelder P, Horvath S. WGCNA: an R package for weighted correlation network analysis. BMC Bioinformatics. 2008, 9:559.

Examples

```
data(BRCASampleData)  
cor_binary_matrix <- cor_binary(ceExp, mRExp)
```

diff_module	<i>diff_module</i>
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Description

Inferring differential modules between two list of module groups

Usage

```
diff_module(  
  Module.group1,  
  Module.group2,  
  sim.cutoff = 0.8,  
  sim.method = "Simpson"  
)
```

Arguments

Module.group1	List object, the first list of module group.
Module.group2	List object, the second list of module group.
sim.cutoff	Similarity cutoff between modules, the interval is [0 1].
sim.method	Methods for calculating similarity between two modules, select one of three methods (Simpson, Jaccard and Lin). Default method is Simpson.

Value

A list of differential modules

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

Examples

```
library(GSEABase)  
data(BRCASampleData)  
modulegenes_WGCNA_all <- module_WGCNA(ceRExp, mRExp)  
modulegenes_WGCNA_1 <- module_WGCNA(ceRExp[-1, ], mRExp[-1, ])  
Differential_module <- diff_module(geneIds(modulegenes_WGCNA_all), geneIds(modulegenes_WGCNA_1))
```

miRExp	<i>miRNA expression data</i>
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Description

miRNA expression data

Format

miRExp: A SummarizedExperiment object with 72 BRCA and 72 normal samples (rows) and 226 miRNAs (columns).

Details

The matched breast invasive carcinoma (BRCA) miRNA, lncRNA and mRNA expression data is obtained from TCGA (<http://cancergenome.nih.gov/>). The data focuses on 72 individuals for which the complete sets of tumor and matched normal (i.e., normal tissue taken from the same patient) profiles are available. A miRNA which has missing values in more than 10 are imputed using the k-nearest neighbours (KNN) algorithm from the impute R package. We use the limma R package to infer differentially expressed miRNAs, ceRNAs and mRNAs between tumour and normal samples. After the analysis, we select top 226 miRNAs which are differentially expressed at a significant level (adjusted p-value < 1E-02, adjusted by Benjamini & Hochberg method).

miRSM	<i>miRSM</i>
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Description

Identify miRNA sponge modules using sensitivity canonical correlation (SCC), sensitivity distance correlation (SDC), sensitivity RV coefficient (SRVC), sensitivity similarity index (SSI), sensitivity generalized coefficient of determination (SGCD), sensitivity Coxhead's or Rozeboom's coefficient (SCRC), and sponge module (SM) methods.

Usage

```
miRSM(
  miRExp = NULL,
  ceRExp,
  mRExp = NULL,
  miRTarget,
  CandidateModulegenes,
  typex = "standard",
  typez = "standard",
  nperms = 100,
  method = c("SCC", "SDC", "SRVC", "SM", "SSI", "SGCD", "SCRC"),
```

```

num_shared_miRNAs = 3,
pvalue.cutoff = 0.05,
MC.cutoff = 0.8,
SMC.cutoff = 0.1,
RV_method = c("RV", "RV2", "RVadjMaye", "RVadjGhaziri"),
BCmethod = "BCPlaid",
CRC_method = c("Coxhead", "Rozeboom")
)

```

Arguments

miRExp	NULL (default) or a SummarizedExperiment object. miRNA expression data: rows are samples and columns are miRNAs.
ceRExp	A SummarizedExperiment object. ceRNA expression data: rows are samples and columns are ceRNAs.
mRExp	NULL (default) or a SummarizedExperiment object. mRNA expression data: rows are samples and columns are mRNAs.
miRTarget	A SummarizedExperiment object. Putative miRNA-target binding information.
CandidateModulegenes	List object: a list of candidate miRNA sponge modules. Only for the SCC, SDC, SRVC, SSI, SGCD and SCRC methods.
typex	The columns of x unordered (type='standard') or ordered (type='ordered'). Only for the SCC method.
typez	The columns of z unordered (type='standard') or ordered (type='ordered'). Only for the SCC method.
nperms	The number of permutations. Only for the SCC method.
method	The method selected to identify miRNA sponge modules, including 'SCC', 'SDC', 'SRVC', 'SM', 'SSI', 'SGCD' and 'SCRC'.
num_shared_miRNAs	The number of common miRNAs shared by a group of ceRNAs and mRNAs. Only for the SCC, SDC, SRVC, SSI, SGCD and SCRC methods.
pvalue.cutoff	The p-value cutoff of significant sharing of common miRNAs by a group of ceRNAs and mRNAs or significant correlation.
MC.cutoff	The cutoff of matrix correlation (canonical correlation, distance correlation and RV coefficient). Only for the SCC, SDC, SRVC, SSI, SGCD and SCRC methods.
SMC.cutoff	The cutoff of sensitivity matrix correlation (sensitivity canonical correlation, sensitivity distance correlation and sensitivity RV coefficient). Only for the SCC, SDC, SRVC, SSI, SGCD and SCRC methods when miRExp is not NULL.
RV_method	the method of calculating RV coefficients. Select one of 'RV', 'RV2', 'RVadj-Maye' and 'RVadjGhaziri' methods. Only for the SRVC method.
BCmethod	Specification of the biclustering method, including 'BCBimax', 'BCCC', 'BC-Plaid' (default), 'BCQuest', 'BCSpectral', 'BCXmotifs'. Only for the SM method.
CRC_method	the method of calculating matrix correlation. Select one of 'Coxhead' and 'Rozeboom' methods. Only for the SCRC method.

Value

List object: Group competition of miRNA sponge modules, and miRNA sponge modules.

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

References

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- Coxhead P. Measuring the relationship between two sets of variables. *British journal of mathematical and statistical psychology*, 1974, 27(2): 205-212.
- Rozeboom WW. Linear correlations between sets of variables. *Psychometrika*, 1965, 30(1): 57-71.

Examples

```
data(BRCASampleData)
modulegenes_igraph <- module_igraph(ceRExp[, seq_len(10)],
  mRExp[, seq_len(10)])
# Identify miRNA sponge modules using sensitivity RV coefficient (SRVC)
miRSM_igraph_SRVC <- miRSM(miRExp, ceRExp, mRExp, miRTarget,
  modulegenes_igraph, method = "SRVC",
  SMC.cutoff = 0.01, RV_method = "RV")
```

miRSM_SS

miRSM_SS

Description

Inferring sample-specific miRNA sponge modules

Usage

```
miRSM_SS(
  Modulelist.all,
  Modulelist.exceptk,
  sim.cutoff = 0.8,
  sim.method = "Simpson"
)
```

Arguments

`Modulelist.all` List object, modules using all of samples.

`Modulelist.exceptk` List object, modules using all of samples excepting sample k.

`sim.cutoff` Similarity cutoff between modules, the interval is [0 1].

`sim.method` Methods for calculating similarity between two modules, select one of three methods (Simpson, Jaccard and Lin). Default method is Simpson.

Value

A list of sample-specific miRNA sponge modules

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

Examples

```
data(BRCASampleData)
nsamples <- 3
modulegenes_all <- module_igraph(ceRExp[, 151:300], mRExp[, 151:300])
modulegenes_exceptk <- lapply(seq(nsamples), function(i)
  module_WGCNA(ceRExp[-i, seq(150)],
  mRExp[-i, seq(150)]))

miRSM_SRVC_all <- miRSM(miRExp, ceRExp[, 151:300], mRExp[, 151:300],
  miRTarget, modulegenes_all,
  method = "SRVC", SMC.cutoff = 0.01,
  RV_method = "RV")
miRSM_SRVC_exceptk <- lapply(seq(nsamples), function(i) miRSM(miRExp[-i, ],
  ceRExp[-i, seq(150)], mRExp[-i, seq(150)],
```

```

miRTarget, modulegenes_exceptk[[i]],
method = "SRVC",
SMC.cutoff = 0.01, RV_method = "RV"))

Modulegenes_all <- miRSM_SRVC_all[[2]]
Modulegenes_exceptk <- lapply(seq(nsamples), function(i)
  miRSM_SRVC_exceptk[[i]][[2]])

Modules_SS <- miRSM_SS(Modulegenes_all, Modulegenes_exceptk)

```

miRTarget

miRNA-target interactions

Description

miRNA-target interactions

Format

miRTarget: A SummarizedExperiment object with 29901 miRNA-target interactions.

Details

The miRNA-target binding information is from miRTarBase v8.0 (<http://mirtarbase.mbc.nctu.edu.tw/php/index.php>), and LncBase v2.0 (http://carolina.imis.athena-innovation.gr/diana_tools/web/index.php?r=lnccbasev2/index). Among 226 miRNAs, 305 lncRNAs and 500 mRNAs which are differentially expressed, we obtain 29901 miRNA-target interactions (including miRNA-lncRNA and miRNA-mRNA interactions).

References

Hastie T, Tibshirani R, Narasimhan B, Chu G. impute: Imputation for microarray data. R package version 1.54.0. doi: 10.18129/B9.bioc.impute.

Ritchie ME, Phipson B, Wu D, Hu Y, Law CW, Shi W, et al. limma powers differential expression analyses for RNA-sequencing and microarray studies. *Nucleic Acids Res.* 2015; 43(7):e47.

module_biclust

module_biclust

Description

Identification of gene modules from matched ceRNA and mRNA expression data or single gene expression data using a series of biclustering packages, including biclust, iBBiG, fabia, BicARE, isa2, s4vd, BiBitR and rqubic

Usage

```

module_biclust(
  ceRExp,
  mRExp = NULL,
  BCmethod = "fabia",
  num.modules = 10,
  num.ModuleceRs = 2,
  num.ModulemRs = 2
)

```

Arguments

ceRExp	A SummarizedExperiment object. ceRNA expression data: rows are samples and columns are ceRNAs.
mRExp	NULL (default) or a SummarizedExperiment object. mRNA expression data: rows are samples and columns are mRNAs.
BCmethod	Specification of the biclustering method, including 'BCBimax', 'BCCC', 'BC-Plaid' (default), 'BCQuest', 'BCSpectral', 'BCXmotifs', 'iBBiG', 'fabia', 'fabiap', 'fabias', 'mfsc', 'nmfdiv', 'nmfeu', 'nmfsc', 'FLOC', 'isa', 'BCs4vd', 'BC-ssvd', 'bibit' and 'quBicluster'.
num.modules	The number of modules to be identified. For the 'BCPlaid', 'BCSpectral', 'isa' and 'bibit' methods, no need to set the parameter. For the 'quBicluster' method, the parameter is used to set the number of biclusters that should be reported.
num.ModuleceRs	The minimum number of ceRNAs in each module.
num.ModulemRs	The minimum number of mRNAs in each module.

Value

GeneSetCollection object: a list of module genes.

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

References

- Prelìc A, Bleuler S, Zimmermann P, Wille A, Buhlmann P, Grissem W, Hennig L, Thiele L, Zitzler E. A systematic comparison and evaluation of biclustering methods for gene expression data. *Bioinformatics*. 2006, 22(9):1122-9.
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- Li G, Ma Q, Tang H, Paterson AH, Xu Y. QUBIC: a qualitative biclustering algorithm for analyses of gene expression data. *Nucleic Acids Res.* 2009, 37(15):e101.

Examples

```
data(BRCASampleData)
modulegenes_biclust <- module_biclust(ceRExp[, seq_len(30)],
  mRExp[, seq_len(30)])
```

module_CEA

module_CEA

Description

Cancer enrichment analysis of miRNA sponge modules using hypergeometric distribution test

Usage

```
module_CEA(ceRExp, mRExp = NULL, Cancergenes, Modulelist)
```

Arguments

- | | |
|-------------|--|
| ceRExp | A SummarizedExperiment object. ceRNA expression data: rows are samples and columns are ceRNAs. |
| mRExp | NULL (default) or a SummarizedExperiment object. mRNA expression data: rows are samples and columns are mRNAs. |
| Cancergenes | A SummarizedExperiment object: a list of cancer genes given. |
| Modulelist | List object: a list of the identified miRNA sponge modules. |

Value

Cancer enrichment significance p-values of the identified miRNA sponge modules

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

References

Johnson NL, Kotz S, Kemp AW (1992) "Univariate Discrete Distributions", Second Edition. New York: Wiley.

Examples

```
data(BRCASampleData)
modulegenes_WGCNA <- module_WGCNA(ceRExp, mRExp)
# Identify miRNA sponge modules using sensitivity RV coefficient (SRVC)
miRSM_WGCNA_SRVC <- miRSM(miRExp, ceRExp, mRExp, miRTarget,
                          modulegenes_WGCNA, method = "SRVC",
                          SMC.cutoff = 0.01, RV_method = "RV")
miRSM_WGCNA_SRVC_genes <- miRSM_WGCNA_SRVC[[2]]
miRSM.CEA.pvalue <- module_CEA(ceRExp, mRExp, BRCA_genes,
                              miRSM_WGCNA_SRVC_genes)
```

module_clust

module_clust

Description

Identification of gene modules from matched ceRNA and mRNA expression data or single gene expression data using a series of clustering packages, including stats, flashClust, dbscan, subspace, mclust, SOMbrero and ppclust packages.

Usage

```
module_clust(
  ceRExp,
  mRExp = NULL,
  cluster.method = "kmeans",
  num.modules = 10,
  num.ModuleceRs = 2,
  num.ModulemRs = 2
)
```

Arguments

ceRExp	A SummarizedExperiment object. ceRNA expression data: rows are samples and columns are ceRNAs.
mRExp	NULL (default) or a SummarizedExperiment object. mRNA expression data: rows are samples and columns are mRNAs.
cluster.method	Specification of the clustering method, including 'kmeans'(default), 'hclust', 'dbscan', 'clique', 'gmm', 'som' and 'fcm'.
num.modules	Parameter of the number of modules to be identified for the 'kmeans', 'hclust', 'gmm' and 'fcm' methods. Parameter of the number of intervals for the 'clique' method. For the 'dbscan' and 'som' methods, no need to set the parameter.
num.ModuleceRs	The minimum number of ceRNAs in each module.
num.ModulemRs	The minimum number of mRNAs in each module.

Value

GeneSetCollection object: a list of module genes.

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

References

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- Bezdek JC. Pattern recognition with fuzzy objective function algorithms. Plenum, NY, 1981.

Examples

```
data(BRCASampleData)
modulegenes_clust <- module_clust(ceRExp[, seq_len(30)],
  mRExp[, seq_len(30)])
```

module_Coexpress	<i>module_Coexpress</i>
------------------	-------------------------

Description

Co-expression analysis of each miRNA sponge module and its corresponding random miRNA sponge modules

Usage

```
module_Coexpress(
  ceRExp,
  mRExp = NULL,
  Modulelist,
  resample = 1000,
  method = c("mean", "median"),
  test.method = c("t.test", "wilcox.test")
)
```

Arguments

- | | |
|------------|---|
| ceRExp | A SummarizedExperiment object. ceRNA expression data: rows are samples and columns are ceRNAs. |
| mRExp | NULL (default) or a SummarizedExperiment object. mRNA expression data: rows are samples and columns are mRNAs. |
| Modulelist | List object: a list of the identified miRNA sponge modules. |
| resample | The number of random miRNA sponge modules generated, and 1000 times in default. |
| method | The method used to evaluate the co-expression level of each miRNA sponge module. Users can select "mean" or "median" to calculate co-expression value of each miRNA sponge module and its corresponding random miRNA sponge module. |

`test.method` The method used to evaluate statistical significance p-value of co-expression level higher than random miRNA sponge modules. Users can select "t.test" or "wilcox.test" to calculate statistical significance p-value of co-expression level higher than random miRNA sponge modules.

Value

List object: co-expression values of miRNA sponge modules and their corresponding random miRNA sponge modules, and statistical significance p-value of co-expression level higher than random miRNA sponge modules.

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

Examples

```
data(BRCASampleData)
modulegenes_WGCNA <- module_WGCNA(ceRExp, mRExp)
# Identify miRNA sponge modules using sensitivity RV coefficient (SRVC)
miRSM_WGCNA_SRVC <- miRSM(miRExp, ceRExp, mRExp, miRTarget,
                          modulegenes_WGCNA, method = "SRVC",
                          SMC.cutoff = 0.01, RV_method = "RV")
miRSM_WGCNA_SRVC_genes <- miRSM_WGCNA_SRVC[[2]]
miRSM_WGCNA_Coexpress <- module_Coexpress(ceRExp, mRExp,
                                          miRSM_WGCNA_SRVC_genes,
                                          resample = 10, method = "mean",
                                          test.method = "t.test")
```

module_FA

module_FA

Description

Functional analysis of miRNA sponge modules, including functional enrichment and disease enrichment analysis

Usage

```
module_FA(
  Modulelist,
  GOont = "BP",
  Diseaseont = "DO",
  KEGGorganism = "hsa",
  Reactomeorganism = "human",
  OrgDb = "org.Hs.eg.db",
  padjustvaluecutoff = 0.05,
  padjustedmethod = "BH",
```



```

miRSM_WGCNA_SRVC_genes <- miRSM_WGCNA_SRVC[[2]]
miRSM_WGCNA_SRVC_FEA <- module_FA(miRSM_WGCNA_SRVC_genes, Analysis.type = 'FEA')
miRSM_WGCNA_SRVC_DEA <- module_FA(miRSM_WGCNA_SRVC_genes, Analysis.type = 'DEA')

## End(Not run)

```

 module_GFA

module_GFA

Description

Identification of gene modules from matched ceRNA and mRNA expression data or single gene expression data using GFA package

Usage

```

module_GFA(
  ceRExp,
  mRExp = NULL,
  StrengthCut = 0.9,
  iter.max = 5000,
  num.ModuleceRs = 2,
  num.ModulemRs = 2
)

```

Arguments

ceRExp	A SummarizedExperiment object. ceRNA expression data: rows are samples and columns are ceRNAs.
mRExp	NULL (default) or a SummarizedExperiment object. mRNA expression data: rows are samples and columns are mRNAs.
StrengthCut	Desired minimum strength (absolute value of association with interval [0 1]) for each bicluster.
iter.max	The total number of Gibbs sampling steps (default 1000).
num.ModuleceRs	The minimum number of ceRNAs in each module.
num.ModulemRs	The minimum number of mRNAs in each module.

Value

GeneSetCollection object: a list of module genes.

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

References

Bunte K, Leppäaho E, Saarinen I, Kaski S. Sparse group factor analysis for biclustering of multiple data sources. *Bioinformatics*. 2016, 32(16):2457-63.

Leppäaho E, Ammad-ud-din M, Kaski S. GFA: exploratory analysis of multiple data sources with group factor analysis. *J Mach Learn Res*. 2017, 18(39):1-5.

Examples

```
data(BRCASampleData)
modulegenes_GFA <- module_GFA(ceRExp[seq_len(20), seq_len(15)],
  mRExp[seq_len(20), seq_len(15)], iter.max = 3000)
```

module_group_sim *module_group_sim*

Description

Calculating similarity between two list of module groups

Usage

```
module_group_sim(Module.group1, Module.group2, sim.method = "Simpson")
```

Arguments

Module.group1 List object, the first list of module group.
Module.group2 List object, the second list of module group.
sim.method Methods for calculating similarity between two modules, select one of three methods (Simpson, Jaccard and Lin). Default method is Simpson.

Value

Similarity between two list of module groups

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

References

Simpson E H. Measurement of diversity. *Nature*, 1949, 163(4148): 688-688.

Jaccard P. The distribution of the flora in the alpine zone. 1. *New phytologist*, 1912, 11(2): 37-50.

Lin D. An information-theoretic definition of similarity. in: *Icml*. 1998, 98(1998): 296-304.

Examples

```
library(GSEABase)
data(BRCASampleData)
modulegenes_WGCNA <- module_WGCNA(ceRExp, mRExp)
modulegenes_igraph <- module_igraph (ceRExp, mRExp)
Sim <- module_group_sim(geneIds(modulegenes_WGCNA), geneIds(modulegenes_igraph))
```

module_igraph	<i>module_igraph</i>
---------------	----------------------

Description

Identification of gene modules from matched ceRNA and mRNA expression data or single gene expression data using igraph package

Usage

```
module_igraph(
  ceRExp,
  mRExp = NULL,
  cor.method = "pearson",
  pos.p.value.cutoff = 0.01,
  cluster.method = "greedy",
  num.ModuleceRs = 2,
  num.ModulemRs = 2
)
```

Arguments

ceRExp	A SummarizedExperiment object. ceRNA expression data: rows are samples and columns are ceRNAs.
mRExp	NULL (default) or a SummarizedExperiment object. mRNA expression data: rows are samples and columns are mRNAs.
cor.method	The method of calculating correlation selected, including 'pearson' (default), 'kendall', 'spearman'.
pos.p.value.cutoff	The significant p-value cutoff of positive correlation.
cluster.method	The clustering method selected in igraph package, including 'betweenness', 'greedy' (default), 'infomap', 'prop', 'eigen', 'louvain', 'walktrap'.
num.ModuleceRs	The minimum number of ceRNAs in each module.
num.ModulemRs	The minimum number of mRNAs in each module.

Value

GeneSetCollection object: a list of module genes.

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

References

Csardi G, Nepusz T. The igraph software package for complex network research, InterJournal, Complex Systems. 2006:1695.

Examples

```
data(BRCASampleData)
modulegenes_igraph <- module_igraph(ceRExp[, seq_len(10)],
  mRExp[, seq_len(10)])
```

module_miRdistribute *module_miRdistribute*

Description

miRNA distribution analysis of sharing miRNAs by the identified miRNA sponge modules

Usage

```
module_miRdistribute(share_miRs)
```

Arguments

share_miRs List object: a list of common miRNAs of each miRNA sponge module generated by share_miRs function.

Value

Matrix object: miRNA distribution in each miRNA sponge module.

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

Examples

```
data(BRCASampleData)
modulegenes_WGCNA <- module_WGCNA(ceRExp, mRExp)
# Identify miRNA sponge modules using sensitivity RV coefficient (SRVC)
miRSM_WGCNA_SRVC <- miRSM(miRExp, ceRExp, mRExp, miRTarget,
  modulegenes_WGCNA, method = "SRVC",
  SMC.cutoff = 0.01, RV_method = "RV")
miRSM_WGCNA_SRVC_genes <- miRSM_WGCNA_SRVC[[2]]
miRSM_WGCNA_share_miRs <- share_miRs(miRExp, miRTarget, miRSM_WGCNA_SRVC_genes)
```

```
miRSM_WGCNA_miRdistribute <- module_miRdistribute(miRSM_WGCNA_share_miRs)
```

module_miRsp sponge	<i>module_miRsp sponge</i>
---------------------	----------------------------

Description

Extract miRNA sponge interactions of each miRNA sponge module

Usage

```
module_miRsp sponge(Modulelist)
```

Arguments

Modulelist List object: a list of the identified miRNA sponge modules.

Value

List object: miRNA sponge interactions of each miRNA sponge module.

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

Examples

```
data(BRCASampleData)
modulegenes_WGCNA <- module_WGCNA(ceRExp, mRExp)
# Identify miRNA sponge modules using sensitivity RV coefficient (SRVC)
miRSM_WGCNA_SRVC <- miRSM(miRExp, ceRExp, mRExp, miRTarget,
                        modulegenes_WGCNA, method = "SRVC",
                        SMC.cutoff = 0.01, RV_method = "RV")
miRSM_WGCNA_SRVC_genes <- miRSM_WGCNA_SRVC[[2]]
miRSM_WGCNA_miRsp sponge <- module_miRsp sponge(miRSM_WGCNA_SRVC_genes)
```

 module_NMF

module_NMF

Description

Identification of gene modules from matched ceRNA and mRNA expression data or single gene expression data using NMF package

Usage

```
module_NMF(
  ceRExp,
  mRExp = NULL,
  NMF.algorithm = "brunet",
  num.modules = 10,
  num.ModuleceRs = 2,
  num.ModulemRs = 2
)
```

Arguments

ceRExp	A SummarizedExperiment object. ceRNA expression data: rows are samples and columns are ceRNAs.
mRExp	NULL (default) or a SummarizedExperiment object. mRNA expression data: rows are samples and columns are mRNAs.
NMF.algorithm	Specification of the NMF algorithm, including 'brunet' (default), 'Frobenius', 'KL', 'lee', 'nsNMF', 'offset', 'siNMF', 'snmf/l', 'snmf/r'.
num.modules	The number of modules to be identified.
num.ModuleceRs	The minimum number of ceRNAs in each module.
num.ModulemRs	The minimum number of mRNAs in each module.

Value

GeneSetCollection object: a list of module genes.

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

References

Gaujoux R, Seoighe C. A flexible R package for nonnegative matrix factorization. BMC Bioinformatics. 2010, 11:367.

Examples

```

data(BRCASampleData)
# Reimport NMF package to avoid conflicts with DelayedArray package
library(NMF)
modulegenes_NMF <- module_NMF(ceRExp[, seq_len(10)],
  mRExp[, seq_len(10)])

```

module_ProNet

module_ProNet

Description

Identification of gene modules from matched ceRNA and mRNA expression data or single gene expression data using ProNet package

Usage

```

module_ProNet(
  ceRExp,
  mRExp = NULL,
  cor.method = "pearson",
  pos.p.value.cutoff = 0.01,
  cluster.method = "MCL",
  num.ModuleceRs = 2,
  num.ModulemRs = 2
)

```

Arguments

ceRExp	A SummarizedExperiment object. ceRNA expression data: rows are samples and columns are ceRNAs.
mRExp	NULL (default) or a SummarizedExperiment object. mRNA expression data: rows are samples and columns are mRNAs.
cor.method	The method of calculating correlation selected, including 'pearson' (default), 'kendall', 'spearman'.
pos.p.value.cutoff	The significant p-value cutoff of positive correlation
cluster.method	The clustering method selected in ProNet package, including 'FN', 'MCL' (default), 'LINKCOMM', 'MCODE'.
num.ModuleceRs	The minimum number of ceRNAs in each module.
num.ModulemRs	The minimum number of mRNAs in each module.

Value

GeneSetCollection object: a list of module genes.

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

References

Clauset A, Newman ME, Moore C. Finding community structure in very large networks. *Phys Rev E Stat Nonlin Soft Matter Phys.*, 2004, 70(6 Pt 2):066111.

Enright AJ, Van Dongen S, Ouzounis CA. An efficient algorithm for large-scale detection of protein families. *Nucleic Acids Res.*, 2002, 30(7):1575-84.

Kalinka AT, Tomancak P. linkcomm: an R package for the generation, visualization, and analysis of link communities in networks of arbitrary size and type. *Bioinformatics*, 2011, 27(14):2011-2.

Bader GD, Hogue CW. An automated method for finding molecular complexes in large protein interaction networks. *BMC Bioinformatics*, 2003, 4:2.

Examples

```
data(BRCASampleData)
modulegenes_ProNet <- module_ProNet(ceRExp[, seq_len(10)],
  mRExp[, seq_len(10)])
```

module_Validate	<i>module_Validate</i>
-----------------	------------------------

Description

Validation of miRNA sponge interactions in each miRNA sponge module

Usage

```
module_Validate(Modulelist, Groundtruth)
```

Arguments

Modulelist	List object: a list of the identified miRNA sponge modules.
Groundtruth	Matrix object: a list of experimentally validated miRNA sponge interactions.

Value

List object: a list of validated miRNA sponge interactions in each miRNA sponge module

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

Examples

```

data(BRCASampleData)
modulegenes_WGCNA <- module_WGCNA(ceRExp, mRExp)
# Identify miRNA sponge modules using sensitivity RV coefficient (SRVC)
miRSM_WGCNA_SRVC <- miRSM(miRExp, ceRExp, mRExp, miRTarget,
                          modulegenes_WGCNA, method = "SRVC",
                          SMC.cutoff = 0.01, RV_method = "RV")
miRSM_WGCNA_SRVC_genes <- miRSM_WGCNA_SRVC[[2]]
Groundtruthcsv <- system.file("extdata", "Groundtruth_high.csv", package="miRSM")
Groundtruth <- read.csv(Groundtruthcsv, header=TRUE, sep=",")
miRSM.Validate <- module_Validate(miRSM_WGCNA_SRVC_genes, Groundtruth)

```

module_WGCNA

module_WGCNA

Description

Identification of co-expressed gene modules from matched ceRNA and mRNA expression data or single gene expression data using WGCNA package

Usage

```

module_WGCNA(
  ceRExp,
  mRExp = NULL,
  RsquaredCut = 0.9,
  num.ModuleceRs = 2,
  num.ModulemRs = 2
)

```

Arguments

ceRExp	A SummarizedExperiment object. ceRNA expression data: rows are samples and columns are ceRNAs.
mRExp	NULL (default) or a SummarizedExperiment object. mRNA expression data: rows are samples and columns are mRNAs.
RsquaredCut	Desired minimum scale free topology fitting index R^2 with interval [0 1].
num.ModuleceRs	The minimum number of ceRNAs in each module.
num.ModulemRs	The minimum number of mRNAs in each module.

Value

GeneSetCollection object: a list of module genes.

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

References

Langfelder P, Horvath S. WGCNA: an R package for weighted correlation network analysis. *BMC Bioinformatics*. 2008, 9:559.#'

Examples

```
data(BRCASampleData)
modulegenes_WGCNA <- module_WGCNA(ceRExp[, seq_len(80)],
  mRExp[, seq_len(80)])
```

mRExp	<i>mRNA expression data</i>
-------	-----------------------------

Description

mRNA expression data

Format

mRExp: A SummarizedExperiment object with 72 BRCA and 72 normal samples (rows) and 226 miRNAs (columns).

Details

The matched breast invasive carcinoma (BRCA) miRNA, lncRNA and mRNA expression data is obtained from TCGA (<http://cancergenome.nih.gov/>). The data focuses on 72 individuals for which the complete sets of tumor and matched normal (i.e., normal tissue taken from the same patient) profiles are available. A mRNA which has missing values in more than 10 are imputed using the k-nearest neighbours (KNN) algorithm from the impute R package. We use the limma R package to infer differentially expressed mRNAs between tumour and normal samples. After the analysis, we select top 500 mRNAs which are differentially expressed at a significant level (adjusted p-value < 1E-02, adjusted by Benjamini & Hochberg method).

share_miRs	<i>share_miRs</i>
------------	-------------------

Description

Extract common miRNAs of each miRNA sponge module

Usage

```
share_miRs(miRExp = NULL, miRTarget, Modulelist)
```

Arguments

miRExp	NULL (default) or a SummarizedExperiment object. miRNA expression data: rows are samples and columns are miRNAs.
miRTarget	A SummarizedExperiment object. Putative miRNA-target binding information.
Modulelist	List object: a list of the identified miRNA sponge modules.

Value

List object: a list of common miRNAs of each miRNA sponge module.

Author(s)

Junpeng Zhang (<https://www.researchgate.net/profile/Junpeng-Zhang-2>)

Examples

```
data(BRCASampleData)
modulegenes_WGCNA <- module_WGCNA(ceRExp, mRExp)
# Identify miRNA sponge modules using sensitivity RV coefficient (SRVC)
miRSM_WGCNA_SRVC <- miRSM(miRExp, ceRExp, mRExp, miRTarget,
                          modulegenes_WGCNA, method = "SRVC",
                          SMC.cutoff = 0.01, RV_method = "RV")
miRSM_WGCNA_SRVC_genes <- miRSM_WGCNA_SRVC[[2]]
miRSM_WGCNA_share_miRs <- share_miRs(miRExp, miRTarget, miRSM_WGCNA_SRVC_genes)
```

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