# Package 'BioTIP'

April 24, 2025

```
Tipping-Point
Version 1.23.0
Author Zhezhen Wang, Andrew Goldstein, Yuxi Sun, Biniam Feleke, Qier An, Antonio Feliciano, Xi-
     nan Yang
Maintainer Yuxi (Jen-
     nifer) Sun <ysun11@uchicago.edu>, Zhezhen Wang <zhezhen@uchicago.edu>, and X Holly Yang <xyang2@uchicago.edu>
Description Adopting tipping-point theory to transcriptome profiles
      to unravel disease regulatory trajectory.
Depends R (>= 3.6)
License GPL-2
Encoding UTF-8
LazyData true
Roxygen list(markdown = TRUE)
RoxygenNote 7.1.0
Suggests knitr, markdown, base, rmarkdown, ggplot2
VignetteBuilder knitr
biocViews Sequencing, RNASeq, GeneExpression, Transcription, Software
URL https://github.com/xyang2uchicago/BioTIP
Imports igraph, cluster, psych, stringr, GenomicRanges, MASS, scran
git_url https://git.bioconductor.org/packages/BioTIP
git_branch devel
git_last_commit 6d52a64
git_last_commit_date 2025-04-15
Repository Bioconductor 3.22
Date/Publication 2025-04-24
```

Title BioTIP: An R package for characterization of Biological

Type Package

2 BioTIP-package

## **Contents**

	BioTIP-package	2
	avg.cor.shrink	3
	cod	2
	cor.shrink	
	gencode	(
	getBiotypes	1
	getCluster_methods	8
	getCTS	9
	getIc	10
	getIc.new	12
	getMaxMCImember	13
	getMaxStats	1.
	getMCI	10
	getMCI_inner	1
	getNetwork	18
	getReadthrough	19
	GSE6136_cli	20
	GSE6136_matrix	2
	ILEF	2
	intron	22
	optimize.sd_selection	22
	plotBar_MCI	24
	plotIc	25
	plotMaxMCI	20
	plot_Ic_Simulation	2
	plot_MCI_Simulation	28
	plot_SS_Simulation	29
	sd_selection	3
	simulationMCI	32
	simulation_Ic	33
	simulation_Ic_sample	3.
T., J.,		21
Index		3'
		,
BioT.	IP-package BioTIP: BioTIP: An R package for characterization of Biologica	l
	Tipping-Point	

## Description

Adopting tipping-point theory to transcriptome profiles to unravel disease regulatory trajectory.

## See Also

Useful links:

• https://github.com/xyang2uchicago/BioTIP

avg.cor.shrink 3

avg.cor.shrink	Estimation of average values of correlation	

## **Description**

This function takes in ine (or two) matrix X (rows are genes, columns are samples) (or Y). It then calculates the average pairwise correlation between genes or samples. This method uses the method outlined by Schafer and Strimmer in "A Shrinkage Approach to Large-Scale Covariance Matrix Estimation and Implications for Functional Genomics" (2005)

## Usage

```
avg.cor.shrink(
   X,
   Y = NULL,
   MARGIN = c(1, 2),
   shrink = TRUE,
   abs = FALSE,
   target = c("zero", "average", "half")
)
```

## **Arguments**

X	A G1 x S matrix of counts. Rows correspond to genes, columns correspond to samples.
Υ	A G2 x S matrix of counts. Rows correspond to genes, columns correspond to samples. By default is NULL.
MARGIN	An integer indicate whether the rows (1, genes) or the columns (2, samples) to be calculated for pairwise correlation.
shrink	A flag specifying whether to shrink the correlation or not.
abs	A flag specifying whether to take the absolute value before taking the average (used for gene-gene correlations, not sample-sample correlations)
target	A character choose among ('zero', 'average', 'half'), indicating whether to shrink towards zero (for gene-gene correlations), shrink towards their empirical common average, one or 0.5 (for sample-sample correlations).

## Value

The average pairwise correlation between genes or samples.

## Author(s)

Andrew Goldstein <andrewgoldstein@uchicago.edu>; Xinan H Yang <xyang2@uchicago.edu>

## References

Schafer and Strimmer (2005) "A Shrinkage Approach to Large-Scale Covariance Matrix Estimation and Implications for Functional Genomics"

4 cod

### **Examples**

```
## Generating a data X as coming from a multivariate normal distribution
## with 10 highly correlated variables, roughly simulating correlated genes.
M = matrix(.9, nrow = 10, ncol = 10)
diag(M) = 1
mu = rnorm(10)
X = MASS::mvrnorm(1000, mu, M)
dim(X) #1000 10
## Mean value of standard pairwise correlation between 1000 genes
\# cor_tX = cor(t(X))
# mean(abs(cor_tX[upper.tri(cor_tX, diag = FALSE)])) # 0.9150228
## Calculating estimated pairwise correlation between 1000 genes
avg.cor.shrink(X, MARGIN=1,shrink = TRUE, targe='zero') # 0.8287838
M = matrix(.9, nrow = 20, ncol = 10)
diag(M) = 1
Y = rbind(M, X)
dim(Y) #1020 10
avg.cor.shrink(X, Y, MARGIN=1, shrink = TRUE, targe='zero') \ \#0.8197959
```

cod

cod dataset

## Description

A subset of gencode\_gr extracted as: cod <- subset(gencode\_gr, biotype == 'protein\_coding')

## Usage

cod

### Format

A dataframe with 3 data columns and 1 metadata column.

```
seqnames chromosome names (chr21)
ranges rangeschromosome ranges on the genome (10906201-11029719)
strand specific strand of the genomic location (+,-,*)
```

cor.shrink 5

Estimation of correlation matrix

## Description

This function takes in one (or two) matrix X (rows are genes, columns are samples) (or Y). It then calculates the average pairwise correlation between genes or samples. This method uses the method outlined by Schafer and Strimmer (2005).

### Usage

```
cor.shrink(
   X,
   Y = NULL,
   MARGIN = c(1, 2),
   shrink = TRUE,
   target = c("zero", "average", "half")
)
```

## **Arguments**

X	A G1 x S matrix of counts. Rows correspond to genes, columns correspond to samples.
Υ	A G2 x S matrix of counts. Rows correspond to genes, columns correspond to samples. By default is NULL.
MARGIN	An integer indicate whether the rows (1, genes) or the columns (2, samples) to be calculated for pairwise correlation.
shrink	A flag specifying whether to shrink the correlation or not.
target	A character choose among ('zero', 'average', 'half'), indicating whether to shrink towards zero (for gene-gene correlations), shrink towards their empirical common average, one or 0.5 (for sample-sample correlations).

## Value

The pairwise correlation between genes or samples. If Y==NULL, a G1 x G1 matrix is returned; otherwise, a G1 x G2 matrix is returned.

## Author(s)

Andrew Goldstein <andrewgoldstein@uchicago.edu>

## References

Schafer and Strimmer (2005) "A Shrinkage Approach to Large-Scale Covariance Matrix Estimation and Implications for Functional Genomics"

6 gencode

#### **Examples**

```
require(MASS)
## Generating a data X as coming from a multivariate normal distribution
## with 10 highly correlated variables, roughly simulating correlated genes.
M = matrix(.9, nrow = 10, ncol = 10)
diag(M) = 1
mu = rnorm(10)
X = MASS::mvrnorm(500, mu, M)
dim(X) #500 10
## Calculating pairwise correlation among 500 correlated genes
cor_tX = cor(t(X))
mean(abs(cor_tX[upper.tri(cor_tX, diag = FALSE)])) # 0.9468098
## Calculating estimated pairwise correlation among 500 correlated genes
cor.matrix <- cor.shrink(X, MARGIN=1, shrink = TRUE, targe='zero') # 0.8287838</pre>
dim(cor.matrix) #[1] 500 500
mean(upper.tri(cor.matrix, diag=FALSE)) # 0.499
## Calculating stimated pairwise correlation among 500 correlated genes
## and additional 100 random genes
Y = rbind(X, matrix(rnorm(300*10), nrow = 300))
dim(Y) #800 10
cor.matrix <- cor.shrink(X, Y, MARGIN=1, shrink = TRUE, targe='zero')</pre>
dim(cor.matrix)
                #[1] 500 800
mean(upper.tri(cor.matrix, diag=FALSE)) # 0.6868
```

gencode

A chr21 data from GENCODE GRCh37

## **Description**

A chr21 data from GENCODE GRCh37

## Usage

gencode

## Format

A data frame of chr21 with 2619444 rows and 9 variables:

Rows Rows, include chromosome numbers

**Columns** Columns include seqname, source, feature, start, end, score, strand, frame and attribute **cod\_gr** cod\_gr is a subset of sample data 'gencode\_gr'

getBiotypes 7

getBiotypes	Assigning Transcript Biotypes

## **Description**

The purpose of the getBiotypes() function is to class both coding and noncoding transcripts into biotypes using the most recent GENCODE annotations. This tool can also be used to define potential lncRNAs, given an available genome transcriptome assembly (a gtf file) or any genomic loci of interest.

### Usage

```
getBiotypes(full_gr, gencode_gr, intron_gr = NULL, minoverlap = 1L)
```

### **Arguments**

Bernenes	
full_gr	A GRanges object which contains either coding or noncoding transcripts. Each GRanges objects' columns requires a unique identifications. For further details refer to the GRanges package.
gencode_gr	A GRanges object contatining a human Chr21 GENCODE reference annotation. A metadata column, "biotype", describes the transcript type.
intron_gr	A GRanges object containing the coordinates of non-coding transcripts.
minoverlap	An IRanges argument which detects minimum overlap between two IRanges objects. For more information about minoverlap argument refer to the IRanges package.

## Details

For details of findOverlaps, type.partialOverlap, type.50Overlap type.toPlot, queryhits, and subjecthits see GenomicRanges https://www.bioconductor.org/packages/release/bioc/html/GenomicRanges.html, IRanges https://www.bioconductor.org/packages/release/bioc/html/IRanges.html, and BiocManager http://bioconductor.org/install/index.html.

### Value

A GRanges object that returns classified transcriptome biotypes.

#### Note

Replace the PATH\_FILE when loading your data locally.

## Author(s)

Zhezhen Wang and Biniam Feleke

## Source

Reference GRCh37 genome https://www.gencodegenes.org/human/release\_25lift37.html for details on gtf format visit ensemble https://useast.ensembl.org/info/website/upload/gff.html

8 getCluster\_methods

#### References

Wang, Z.Z., J. M. Cunningham and X. H. Yang (2018). 'CisPi: a transcriptomic score for disclosing cis-acting disease-associated lincRNAs.' Bioinformatics 34(17): 664-670', PMID: 30423099'

### **Examples**

```
# Input datasets from our package's data folder
library(GenomicRanges)
data("gencode")
data("intron")
data("ILEF")

# Converting datasets to GRanges object
gencode_gr = GRanges(gencode)
ILEF_gr = GRanges(ILEF)
cod_gr = GRanges(cod)
intron_gr= GRanges(intron)

# Filtering non-coding transcripts
getBiotypes(ILEF_gr, gencode_gr, intron_gr)

## Not run: getBiotypes(intron_gr)
```

getCluster\_methods

Clustering Network Nodes

### **Description**

This function runs over all states which are grouped samples. For each state, this function splits the correlation network generated from the function <code>getNetwork</code> into several sub-networks (which we called 'module'). The network nodes will be defined by the end-user. For transcriptome analysis, network nodes can be the expressed transcripts. The outputs of this function include the module IDs and node IDs per module.

#### Usage

```
getCluster_methods(
   igraphL,
   method = c("rw", "hcm", "km", "pam", "natural"),
   cutoff = NULL
)
```

## **Arguments**

igraphL

A list of numerical matrices or a list of igraph objects. The list of igraph objects can be the output from the getNetwork function.

method

A mathematical clustering model for analyzing network nodes. Default is a random walk ('rw'). A method could be 'rw', 'hcm', 'km', 'pam', or 'natural', where:

• rw: random walk using cluster\_walktrap function in igraph package. 'igraphL' has to be a list of igraph.

getCTS 9

• hcm: hierarchical clustering using function hclust) and dist, using method 'complete'.

- km and pam: k-medoids or PAM algorithm using KMedoids.
- natrual: if nodes are disconnected, they may naturally cluster and form subnetworks.

cutoff

A numeric value, default is NULL. For each method it means:

- rw: the number of steps needed, see cluster\_walktrap for more detail. If "cutoff" is not assigned, default of 4 will be used.
- hcm, km and pam: number of clusters wanted. No default assigned.
- natural: does not use this parameter.

#### Value

When method=rw: A list of communities objects of R package igraph, whose length is the length of the input object igraphL. These communities objects can be used for visualization when being assigned to the 'mark.groups' parameter of the plot.igraph function of the igraph package. Otherwise this function returns a list of vectors, whose length is the length of the input object igraphL. The names of each vector are the pre-selected transcript IDs by th function sd\_selection. Each vector, whose length is the number of pre-selected transcript in a state, contains the module IDs.

### Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>

### **Examples**

```
test = list('state1' = matrix(sample(1:10, 6), 3, 3), 'state2' =
matrix(sample(1:10, 6), 3, 3), 'state3' = matrix(sample(1:10, 6), 3, 3))
#assign colnames and rownames to the matrix

for(i in names(test)){
   colnames(test[[i]]) = 1:3
   row.names(test[[i]]) = 1:3}

#using 'rw' or 'natural' method
   igraphL <- getNetwork(test, fdr=1)
#[1] "state1:3 nodes"
#[1] "state2:3 nodes"
#[1] "state3:3 nodes"

cl <- getCluster_methods(igraphL)

#using 'km', 'pam' or 'hcm'
   cl <- getCluster_methods(test, method = 'pam', cutoff=2)</pre>
```

getCTS

Obtain the identified BioTiP and its length

### **Description**

getCTS obtains the identified BioTiP and its length based off of MCI scores.

10 getIc

### Usage

```
getCTS(maxMCI, maxMCIms)
```

#### **Arguments**

maxMCI

A numeric vector, whose length is the number of states. This parameter is the maximum MCI score of each state, and it can be obtained from the output of getMaxStats. Names need to be included in names of maxMCIms.

maxMCIms

A list of character vectors per state. The vectors are network nodes (e.g. transcript ids). This parameter is the second element of the output of the function getMaxMCImember.

### Value

A character vector, in which the elements are the unique IDs of the network nodes of the BioTiP.

#### Author(s)

Antonio Feliciano y Pleyto and Zhezhen Wang <zhezhen@uchicago.edu>

### **Examples**

getIc

Get Index for Critical transition (Ic score)

## **Description**

Retrieve Ic scores (Pearson correlation of genes / Pearson correlation of samples) for the identified critical transition state

## Usage

```
getIc(
  counts,
  sampleL,
  genes,
  output = c("Ic", "PCCg", "PCCs"),
  fun = c("cor", "BioTIP"),
  shrink = TRUE,
```

getIc 11

#### **Arguments**

counts A numeric matrix or data frame. The rows and columns represent unique tran-

script IDs (geneID) and sample names, respectively.

sampleL A list of vectors, whose length is the number of states. Each vector gives the

sample names in a state. Note that the vector s (sample names) has to be among

the column names of the R object 'df'.

genes A character vector consisting of unique CTS gene ids. This can be obtained

from getMaxMCImember

output A string. Please select from 'Ic', 'PCCg', or 'PCCs'. Uses 'Ic' by default.

'PCCg' is the PCC between genes (numerator) and 'PCCs' is PCC between

samples (denominator)

fun An optional character string indicating the R function to calculate correlations for

all possible pairs of columns of a matrix. When using "BioTIP", The method is

modified to ignore missing values, analogous to how cor(X, use = "pairwise.complete.obs")

works. Note that the "BioTIP" option only function together with shrink =

TRUE.

shrink A flag specifying whether to shrink the correlation or not. This appraoch uses the

method outlined by Schafer and Strimmer in "A Shrinkage Approach to Large-Scale Covariance Matrix Estimation and Implications for Functional Genomics" (2005) Comparing to fun='cor', the 'BioTIP' method without shinkage is modi-

 $fied \ to \ ignore \ missing \ values, \ analogous \ to \ how \ cor(X, \ use = "pairwise.complete.obs")$ 

works.

use An optional character string, when fun=="cor", it gives a method for computing

covariances in the presence of missing values. This must be (an abbreviation of) one of the strings "everything", "all.obs", "complete.obs", "na.or.complete", or

"pairwise.complete.obs".

#### Value

A list of numeric values, whose length and names are inherited from sampleL

#### Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>; Xinan H Yang <xyang2@uchicago.edu>

## References

Schafer and Strimmer (2005) "A Shrinkage Approach to Large-Scale Covariance Matrix Estimation and Implications for Functional Genomics"

M. Mojtahedi et al., Cell Fate Decision as High-Dimensional Critical State Transition. PLoS Biol 14, e2000640 (2016).

### **Examples**

```
counts = matrix(sample(1:100, 27), 3, 9)
colnames(counts) = 1:9
row.names(counts) = c('loci1', 'loci2', 'loci3')
```

12 getIc.new

```
cli = cbind(1:9, rep(c('state1', 'state2', 'state3'), each = 3))
colnames(cli) = c('samples', 'group')
samplesL <- split(cli[, 1],  f = cli[, 'group'])
CTS = c('loci1', 'loci2')

## Comparing the results with an estiamted correlation matrix with that without estimation.
Ic = getIc(counts, samplesL, CTS, fun='cor')
Ic.2 = getIc(counts, samplesL, CTS, fun='BioTIP', shrink=FALSE)
BioTIP = getIc(counts, samplesL, CTS, fun='BioTIP')</pre>
```

getIc.new

Index of criticality Scoring System with estimated correlation, an updated Ic-score

## Description

This function calculates the BioTIP score on a given data matrix X (or two matrixes X and Y). It can also calculate the  $I_c$  score, if desired.

This appraoch uses the method outlined by Schafer and Strimmer in "A Shrinkage Approach to Large-Scale Covariance Matrix Estimation and Implications for Functional Genomics" (2005)

This approach is modified to ignore missing values, analogous to how cor(X, use = "pairwise.complete.obs") works.

The gene-gene correlations are shrunk towards 0, whereas the sample-sample correlations are shrunk towards their empirical average.

#### Usage

```
getIc.new(
   X,
   method = c("BioTIP", "Ic"),
   PCC_sample.target = c("average", "zero", "half"),
   output = c("IndexScore", "PCCg", "PCCs")
)
```

## **Arguments**

Χ

A G x S matrix of counts. Rows correspond to genes, columns correspond to samples.

method

A flag specifying whether to calculate the BioTIP score or the  $I_c$  score

PCC\_sample.target

A character choose among ('average', 'zero', 'half'), indicating whether to shrink PCC towards towards their empirical common average, zero or 0.5 (for sample-sample correlations).

output

A string. Please select from 'IndexScore', 'PCCg', or 'PCCs'. Uses 'IndexScore' by default. 'PCCg' is the PCC between genes (numerator) and 'PCCs' is PCC between samples (denominator).

## Value

A value containing the shrunk BioTIP or non-shrunk  $I_c$  score

getMaxMCImember 13

#### Author(s)

Andrew Goldstein <andrewgoldstein@uchicago.edu>

### **Examples**

```
## Generating a data X as coming from a multivariate normal distribution
## with 10 highly correlated variables, roughly simulating correlated genes.
M = matrix(.9, nrow = 10, ncol = 10)
diag(M) = 1
mu = rnorm(10)
X = MASS::mvrnorm(1000, mu, M)
dim(X) #1000 10
## Calculating pairwise correlation between 1000 genes; then the mean value
## in two ways, respectively
cor_tX = cor(t(X))
mean(abs(cor_tX[upper.tri(cor_tX, diag = FALSE)])) # 0.9150228
getIc.new(X, method = "Ic", output ='PCCg') # 0.9150228
getIc.new(X, method = "BioTIP", output = 'PCCg') # 0.8287838
## Uisng the Index of critical scoreing system, in two ways, respectively
(newscore = getIc.new(X, method = "BioTIP"))
(oldscore = getIc.new(X, method = "Ic"))
```

 ${\tt getMaxMCImember}$ 

Identifying the 'Biomodule'

## **Description**

This function reports the 'biomodule', which is the module with the maximum Module Critical Index (MCI) scores for each state. Each state can have multiple modules (groups of subnetworks derived from the function getCluster\_methods). This function runs over all states.

## Usage

```
getMaxMCImember(membersL, MCI1, minsize = 1)
```

membersL	A list of integer vectors with unique ids as names. Each vector represents the cluster number assign to that unique id. The length of this list is equal to the number of states in the study. This can be the first element of the output from function getMCI or the output from getCluster_methods, see Examples for more detail.
MCI1	A list of numeric vectors with unique cluster numbers as names. Each vector represents the MCI scores of that module. This can be the second element of the output from function getMCI.
minsize	A numerical value of the minimum module size (the number of transcripts in a cluster) to output for downstream analysis.

14 getMaxMCImember

#### Value

A nested list whose length is the length of the input object membersL. Each internal list contains two objects: one object is the vector of biomodule IDs across states, and the other object is a list of transcript IDs (each defines the biomodule per state) across states.

### Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>

## **Examples**

```
#1st option: get the input directly from getMCI function
test = list('state1' = matrix(sample(1:10, 6), 4, 3),
     'state2' = matrix(sample(1:10, 6), 4, 3),
     'state3' = matrix(sample(1:10, 6), 4, 3))
# assign colnames and rownames to the matrix
for(i in names(test)){
  colnames(test[[i]]) = 1:3
    row.names(test[[i]]) = c('g1', 'g2', 'g3', 'g4')}
cluster = list(c(1, 2, 2, 1), c(1, 2, 3, 1), c(2, 2, 1, 1))
names(cluster) = names(test)
for(i in names(cluster)){
  names(cluster[[i]]) = c('g1', 'g2', 'g3', 'g4')}
membersL_noweight <- getMCI(cluster, test)</pre>
maxMCIms <- getMaxMCImember(membersL_noweight[[1]], membersL_noweight[[2]], min =3)</pre>
#The same as
maxMCIms <- getMaxMCImember(cluster, membersL_noweight[[2]], min =2)</pre>
## case1: using 'rw' method by default
igraphL <- getNetwork(test, fdr=1)</pre>
cl <- getCluster_methods(igraphL)</pre>
## make sure every element in list cl is a \code{communities} object
sapply(cl, class)
        state1
                       state2
                                      state3
##"communities" "communities" "communities"
## If there is(are) state(s) that is(are) empty which will not be a communities object(s),
## please manually remove that state(s).
cl = cl[which(sapply(cl, class) == 'communities')]
## and then run
library(igraph)
cluster = lapply(cl, membership)
maxCIms <- getMaxMCImember(cluster, membersL_noweight[[2]], min =2)</pre>
## or run function 'getMCI' and use the 1st option
membersL_noweight <- getMCI(cl, test)</pre>
## case2: using methods other than the default
cl <- getCluster_methods(test, method = "pam", cutoff = 2)</pre>
## check to make sure membersL_noweight[[2]] has values and run
maxCIms <- getMaxMCImember(cl, membersL_noweight[[2]], min =2)</pre>
```

getMaxStats 15

getMaxStats

Get the cluster index and network nodes of biomodule

### **Description**

This function retrieves the cluster index and network-node ids for the identified biomodule (that shows the maximum MCI score) at each state in the study.

### Usage

```
getMaxStats(membersL, idx)
```

## **Arguments**

membersL A two-layer nested list of character or numeric values, any one out of the five

elements output by the function getMCI.

idx A vector of integers that are cluster ids of the biomodule (the module with

the highest MCI score) per state. This is the first element of the result from

getMaxMCImember.

#### Value

A list describing the biomodule of each state, corresponding to one of the five elements (members, MCI, Sd, PCC, and PCCo) outputted by the function getMCI. The calss of the vector depends on the class of the input parameter membersL.

## Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>

### See Also

getMCI

## **Examples**

```
test = list('state1' = matrix(sample(1:10, 6), 4, 3),
    'state2' = matrix(sample(1:10, 6), 4, 3),
    'state3' = matrix(sample(1:10, 6), 4, 3))

# assign colnames and rownames to the matrix
for(i in names(test)){
    colnames(test[[i]]) = 1:3
    row.names(test[[i]]) = c('g1', 'g2', 'g3', 'g4')
}

cluster = list(c(1, 2, 2, 1), c(1, 2, 3, 1), c(2, 2, 1, 1))
names(cluster) = names(test)
for(i in names(cluster)){
    names(cluster[[i]]) = c('g1', 'g2', 'g3', 'g4')
}
```

16 getMCI

```
membersL_noweight <- getMCI(cluster, test)
idx = c(1, 2, 1)
names(idx) = names(membersL_noweight[['sd']])
selectedSD = getMaxStats(membersL_noweight[['sd']], idx)</pre>
```

getMCI

Calculating MCI Scores

### **Description**

This function calculates a module critical index (MCI) score for each module per state within a dataset. Each module is a cluster of transcripts generated from the function getCluster\_methods. Note that a dataset should contains three or more states (samples in groups).

### Usage

```
getMCI(
  groups,
  countsL,
  adjust.size = FALSE,
  fun = c("cor", "BioTIP"),
  df = NULL
)
```

## **Arguments**

groups	A list of elements whose	length is the member of	of states. The elements cou

either be vectors or communities object of the R package igraph. If a vector, it is the output of the function getCluster\_methods. The names of each vector are the pre-selected transcript IDs generated by the function sd\_selection. Each vector, whose length is the number of pre-selected transcripts in a state,

contains the module IDs. If a communities object, it can be obtained by getCluster\_methods using the "rw" method. It is also an output of the function sd\_selection.

countsL A list of x numeric count matrices or x data frame, where x is the number of

states.

adjust.size A Boolean value indicating if MCI score should be adjusted by module size (the

number of transcripts in the module) or not. Default FALSE. This parameter is

not recommended for fun=BioTIP.

fun A character chosen between ("cor", "BioTIP"), indicating where an adjusted

correlation matrix will be used to calculate the MCI score.

df NULL or a numeric matrix or data frame, where rows and columns represent

unique transcript IDs (geneID) and sample names, respectively. Used only when fun='BioTIP'. By default is NULL, estinating the correlation among selected genes. Otherwise, estinating the correlation among all genes in the df, ensuring

cross state comparision.

getMCI\_inner 17

#### Value

A list of five elements (members, MCI, Sd, PCC, and PCCo). Each of element is a two-layer nested list whose length is the length of the input object groups. Each internal nested list is structured according to the number of modules identified in that state.

- members: vectors of unique ids
- MCI: the MCI score
- · sd: standard deviation
- PCC: Mean of pairwise Pearson Correlation Coefficient calculated among the loci in a module.
- PCCo: Mean of pairwise Pearson Correlation Coefficient calculated between the loci in a module and the loci outside that module but inside the same state.

### Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>; Xinan H Yang <xyang2@uchicago.edu>

### **Examples**

```
test = list('state1' = matrix(sample(1:10, 6), 4, 3), 'state2' =
matrix(sample(1:10, 6), 4, 3), 'state3' = matrix(sample(1:10, 6), 4, 3))
## Assign colnames and rownames to the matrix
for(i in names(test)){
   colnames(test[[i]]) = 1:3
   row.names(test[[i]]) = c('g1', 'g2', 'g3', 'g4')}
cluster = list(c(1, 2, 2, 1), c(1, 2, 3, 1), c(2, 2, 1, 1))
names(cluster) = names(test)
for(i in names(cluster)){
   names(cluster[[i]]) = c('g1', 'g2', 'g3', 'g4')}
membersL_noweight <- getMCI(cluster, test, fun='cor')</pre>
names(membersL_noweight)
## [1] "members" "MCI"
                            "sd"
                                      "PCC"
                                                "PCCo"
```

getMCI\_inner

Calculating MCI Score for randomly selected

### **Description**

This function calculates random MCI score, allowing an estimation of correlation matrix using the Schafer-Strimmer Method for the PCC\_in component in the MCI score.

## Usage

```
getMCI_inner(
  members,
  countsL,
  adjust.size,
  fun = c("cor", "BioTIP"),
```

18 getNetwork

```
PCC_gene.target = "zero",
M = NULL
)
```

### **Arguments**

members An integer that is the length of genes in the CTS (critical transition signal).

countsL A list of subset of dat matrix. The list length is the number of states. Each subset

of matrix gives the genes in rows and samples in column.

adjust.size A boolean value indicating if MCI score should be adjust by module size (the

number of transcripts in the module) or not. Default FALSE.

fun A character chosen between ("cor", "BioTIP"), indicating where an adjusted

correlation matrix will be used to calculated the MCI score.

PCC\_gene.target

A character 'zero' indicating that gene-gene correlation matrix will be shrunk.

towards zero, used only for fun='BioTIP'.

M is the overall shrunk correlation matrix, used only for fun='BioTIP'.

#### Value

A vector recording one MCI score per state.

#### Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>; Xinan H Yang <xyang2@uchicago.edu>

getNetwork	Building Networks of Nodes	

### Description

This function builds one correlation network for each state (sample group) and runs across all states. The network nodes are defined by the context of the input dataset. For transcriptomic network analysis, network nodes can be the expressed transcript IDs and network links can be the correlation coefficients. Using the Pearson Correlation Coefficient (PCC) analysis, this function assembles a correlation network of nodes (e.g., co-expressed transcripts) for each state using the R package igraph.

## Usage

```
getNetwork(optimal, fdr = 0.05)
```

## **Arguments**

optimal A list of x numeric data frames, where x is the number of states studied. Each

data frame consists of loci with high standard deviations. This object can be

obtained through sd\_selection function.

fdr A numeric cutoff value for a Pearson Correlation Coefficient (PCC) analysis.

Default is 0.05. Transcripts are linked into a network if their correlations meet

this PCC-significance criterion.

getReadthrough 19

#### Value

A list of igraph objects whose length is the length of the input object optimal. Each object is a network of correlated nodes whose PCCs meet the significant criteria based on the false discovery rate (FDR) control. The length of the list is the number of states with PCC networks. If no PCC meets the significant criteria in a state, the state will be deleted from the output.

### Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>; Xinan H Yang <xyang2@uchicago.edu>

### **Examples**

```
test = list('state1' = matrix(sample(1:10, 6), 2, 3),
  'state2'=matrix(sample(1:10, 6), 2, 3),
  'state3' = matrix(sample(1:10, 6), 2, 3))

for(i in names(test)){
    colnames(test[[i]]) = 1:3
    row.names(test[[i]]) = 1:2}

igraphL <- getNetwork(test, fdr=1)
#[1] "state1:2 nodes"
#[1] "state2:2 nodes"
#[1] "state3:2 nodes</pre>
```

getReadthrough

Overlapping Coding Regions

### **Description**

The getReadthrough() function is used to find long transcripts that cover more than two coding regions for gene regions of interst.

## Usage

```
getReadthrough(gr, cod_gr)
```

## **Arguments**

gr A GRanges object that shows the start and end loci on genome.

cod\_gr A GRanges object containing coding regions.

#### **Details**

For details of findOverlaps, type.partialOverlap, type.50Overlap type.toPlot, queryhits, readthrough and subjecthits see, GenomicRanges https://www.bioconductor.org/packages/release/bioc/html/GenomicRanges.html, IRanges https://www.bioconductor.org/packages/release/bioc/html/IRanges.html, and BiocManager http://bioconductor.org/install/index.html.

### Value

A GRanges object that returns overlapping regions of the classified transcript biotypes.

20 GSE6136\_cli

#### Note

Replace the path\_file when loading data locally to the data directory.

### Author(s)

Zhezhen Wang and Biniam Feleke

### Source

Reference GRCh37 genome https://www.gencodegenes.org/human/release\_251ift37.html. For details on gtf format visit ensemble https://useast.ensembl.org/info/website/upload/gff.html.

#### References

Wang, Z. Z., J. M. Cunningham and X. H. Yang (2018). 'CisPi: a transcriptomic score for disclosing cis-acting disease-associated lincRNAs.' Bioinformatics 34(17): 664-670'

## **Examples**

```
#First Load datasets and libraries
library(GenomicRanges)
data("gencode")
data("ILEF")
data("cod")

# Assigning datasets a GRanges object
gencode_gr = GRanges(gencode)
ILEF_gr = GRanges(ILEF)
cod_gr = GRanges(cod)
getReadthrough(ILEF_gr, cod_gr)

## Not run: getReadthrough(cod_gr)
```

GSE6136\_cli

GSE6136 cli dataset

## Description

A gene annotation samples with their corresponding geneId extracted from GENCODE database. A python script was then run to extcat GSE6136\_cli dataset. The script is included in the R/data\_raw folder.

## Usage

GSE6136\_cli

GSE6136\_matrix 21

### **Format**

A dataframe with 22690 columns and 27 rows column.

GSM142398-GSM142423 Names of gene interest

Rows Summary of GSM genes

#### Source

https://www.gencodegenes.org/human/

GSE6136\_matrix

GSE6136 matrix dataset

## **Description**

A gene annotation samples with their corresponding geneId extracted from GENCODE database. A python script was then run to extcat GSE6136\_cli dataset. The script is included in the R/data\_raw folder.

## Usage

GSE6136\_matrix

### **Format**

A dataframe with 22690 columns and 27 rows column.

GSM142398-GSM142423 Names of gene interest

ID\_REF Reference ID of the target genes

### Source

https://www.gencodegenes.org/human/

**ILEF** 

Chromosome ranges of chr21 dataset

## **Description**

A dataset containing chromosomes in the genome regions of interest for 137 chromosome. The variables are as follows:

## Usage

ILEF

22 optimize.sd\_selection

#### **Format**

```
A data frame of chr21 with 137 rows and 4 variables:
```

```
seqnames chromosome names (chr1,chr6,chr21)
start gene read start position (167684657,167729508)
end end of gene read position (15710335,43717938)
width width of gene
strand specific strand of the genomic location (+,-,*
Row.names name of the data rows(A1BG,vawser)
```

intron

Coding transcriptome in chr21 dataset

### **Description**

A dataset containing chromosomes in the genome regions of interest. The variables are as follows:

## Usage

intron

### **Format**

A data frame with 659327 rows and 5 variables:

```
seqnames chromosome names (chr1,chrM,chr21)
ranges chromosome ranges on the genome(167684657–167729508)
strand specific strand of the genomic location (+,-,*)
name internal assigned names(uc001aaa.3_intron_0_0_chr1_12228_f)
score score not used in this data set(0)
```

optimize.sd\_selection Optimization of sd selection

## **Description**

The optimize.sd\_selection filters a multi-state dataset based on a cutoff value for standard deviation per state and optimizes. By default, a cutoff value of 0.01 is used. Suggested if each state contains more than 10 samples.

optimize.sd\_selection 23

#### Usage

```
optimize.sd_selection(
   df,
   samplesL,
   B = 100,
   percent = 0.8,
   times = 0.8,
   cutoff = 0.01,
   method = c("other", "reference", "previous", "itself", "longitudinal reference"),
   control_df = NULL,
   control_samplesL = NULL
)
```

### **Arguments**

df A dataframe of numerics. The rows and columns represent unique transcript IDs

(geneID) and sample names, respectively.

samplesL A list of n vectors, where n equals to the number of states. Each vector gives the

sample names in a state. Note that the vectors (sample names) has to be among the column names of the R object 'df'.

B An integer indicating number of times to run this optimization, default 1000.

percent A numeric value indicating the percentage of samples will be selected in each

round of simulation.

times A numeric value indicating the percentage of B times a transcript need to be

selected in order to be considered a stable signature.

cutoff A positive numeric value. Default is 0.01. If < 1, automatically goes to se-

lect top x percentage transcripts using the a selecting method (which is either the reference, other or previous stage), e.g. by default it will select top 1

percentage of the transcripts.

method Selection of methods from reference, other, previous, default uses other.

Partial match enabled.

• itself, or longitudinal reference. Some specific requirements for each option:

- reference, the reference has to be the first.
- previous, make sure sampleL is in the right order from benign to malign.
- itself, make sure the cutoff is smaller than 1.
- longitudinal reference make sure control\_df and control\_samplesL are not NULL. The row numbers of control\_df is the same as df and all transcript in df are also in control\_df.

control\_df

A count matrix with unique loci as row names and samples names of control samples as column names, only used for method longitudinal reference.

control\_samplesL

A list of characters with stages as names of control samples, required for method 'longitudinal reference'.

## Value

A list of dataframe of filtered transcripts with the highest standard deviation are selected from df based on a cutoff value assigned. The resulting dataframe represents a subset of the raw input df.

24 plotBar\_MCI

#### Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>

#### See Also

```
sd_selection
```

plotBar\_MCI

plot MCI barplots

## **Description**

A barplot of MCI for all clusters in all states.

## Usage

```
plotBar_MCI(
   MCIl,
   ylim = NULL,
   nr = 1,
   nc = NULL,
   order = NULL,
   minsize = 3,
   states = NULL)
```

## Arguments

MCI1	list can be obtained throu	igh getMCI.

ylim A vector needed if the output barplots need to be on the same y scale.

nr The number of rows to plot.

nc The number of columns to plot, default length(groups).

order A character vector of the order of the barplot. Default is NULL which uses the

input list order.

minsize A non-negative numeric value of minimum size allowed for a cluster.

states A character of the state names to be shown on the plot. Default is NULL, assign

this if you want to show all states including states without nodes.

### Value

Return a barplot of MCI scores across states.

## Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>

### References

Chen L, Liu R, Liu Z, Li M & Aihara K (2012) Detecting early-warning signals for sudden deterioration of complex diseases by dynamical network biomarkers Scientific Reports 2:342

plotIc 25

### **Examples**

```
test = list('state1' = matrix(sample(1:10, 6), 4, 3),
    'state2' = matrix(sample(1:10, 6), 4, 3),
    'state3' = matrix(sample(1:10, 6), 4, 3))
# assign colnames and rownames to the matrix
for(i in names(test)){
    colnames(test[[i]]) = 1:3
    row.names(test[[i]]) = c('g1', 'g2', 'g3', 'g4')
}

cluster = list(c(1, 2, 2, 1), c(1, 2, 3, 1), c(2, 2, 1, 1))
names(cluster) = names(test)
for(i in names(cluster)){
    names(cluster[[i]]) = c('g1', 'g2', 'g3', 'g4')
}
membersL_noweight <- getMCI(cluster, test)
plotBar_MCI(membersL_noweight)</pre>
```

plotIc

plot a line plot of Ic scores for each state.

### **Description**

plot a line plot with Ic score for each state

## Usage

```
plotIc(
    Ic,
    las = 0,
    order = NULL,
    ylab = "Ic",
    col = "black",
    main = NULL,
    add = FALSE,
    ylim = NULL,
    lty = 1:5,
    lwd = 1
)
```

Ic	A vector with names of states. If order is not assigned, then plot by the order of this vector.
las	Numeric in 0, 1, 2, 3; the style of axis labels. Default is 0, meaning labels are parallel. (link to http://127.0.0.1:21580/library/graphics/html/par.html).
order	A vector of state names in the customized order to be plotted, setting to NULL by default.
ylab	titles y axes, as in plot.
col	vector of colors. Colors are used cyclically.

26 plotMaxMCI

main	A character vector. The title of the plot. Default is NULL.
add	logical. If TRUE, plots are added to current one. This is inherited from matplot.
ylim	An integer vector of length 2. Default is NULL.
lty	An vector of line types. This is also inherited from matplot.
lwd	Anineger of line widths. This is also inherited from matplot.

## Value

Return a line plot of Ic score across states.

## Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>

## **Examples**

```
Ic = c('state3' = 3.4, 'state1' = 5.6, 'state2' = 2)
plotIc(Ic, order = c('state1', 'state2', 'state3'))
```

plotMaxMCI

Plot the Maximized MCI per State

## Description

This function generates a line plot over multiple states with the maximum MCI score per state. The module size (i.e., number of network nodes) is specified at each state in parentheses.

## Usage

```
plotMaxMCI(maxMCIms, MCIl, las = 0, order = NULL, states = NULL)
```

maxMCIms	A list of 2 elements. The 1st element is an integer vector of module ids whose names are the state names. The 2nd element is a list of character vectors per state. The vectors are network nodes (e.g. transcript ids). This parameter can be obtained by running function getMaxMCImember.
MCI1	A list of numeric vectors whose names are unique cluster ids. Each vector represents the MCI scores of modules in a state. This can be the second element of the output from the function getMCI.
las	Numeric in 0, 1, 2, 3; the style of axis labels. Default is 0, meaning labels are parallel. See getMCI for more detail.
order	A vector of state names in the customized order to be plotted, setting to NULL by default.
states	A character vector of state names that will be shown on the plot, setting to NULL by default. Assign this if you want to show all states, including states with no resulted modules. This parameter will overwrite the parameter 'order'.

plot\_Ic\_Simulation 27

### Value

Returns a line plot of maximum MCI scores across the states

### Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>

### **Examples**

```
maxMCIms = list(c(state1 = 1, state2 = 2, state3 = 1),
    c(list(state1 = c('g1', 'g2', 'g3'),
    state2 = c('g3', 'g5'),
    state3 = c('g2', 'g6'))))

MCIl = list(state1=c('1' = 8.84, '2' = 6.4),
    state2 = c('1' = NA, '2' = 9.5, '3' = NA),
    state3 = c('1' = 2.3, '2' = 1.4))

plotMaxMCI(maxMCIms, MCIl)
```

plot\_Ic\_Simulation

Line or boxplot of an observed and its simulated scores

## **Description**

Generate a line (or box) plot of Ic score and simulated Ic scores, with three horizontal lines: the min, max and 2\*(max-min) value of the state of interests, or all values.

### Usage

```
plot_Ic_Simulation(
   Ic,
   simulation,
   las = 0,
   ylim = NULL,
   order = NULL,
   main = NULL,
   ylab = "Ic",
   fun = c("matplot", "boxplot"),
   which2point = NULL
)
```

Ic	A vector with names of states. If order is not assigned, then plot by the order of this vector.
simulation	A numeric matrix of Ic scores in which rows are states and columns are numbers of simulated times. It can be obtained from simulation_Ic
las	Numeric in 0, 1, 2, 3; the style of axis labels. Default is 0, meaning labels are parallel. (link to http://127.0.0.1:21580/library/graphics/html/par.html).

28 plot\_MCI\_Simulation

ylim	An integer vector of length 2. Default is NULL.
order	Characters of names of Ic to be plotted in a desired order. Default is NULL.
main	A character vector. The title of the plot. Default is NULL.
ylab	titles y axes, as in plot.
fun	A character choose between ('matplot', 'boxplot'), indicating plot type.
which2point	A character (or integer) which state's values were used to set up the three horizontal lines, by default is NULL, indicating the values of all states will be used.

### Value

Return a plot of the observed Ic (red) and simulated Ic (grey) scores per states.

## Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>; Xinan H Yang <xyang2@uchicago.edu>

## **Examples**

```
sim = matrix(sample(1:10, 9), 3, 3)
row.names(sim) = paste0('state', 1:3)
Ic = c('state1' = 3.4, 'state2' = 5.6, 'state3' = 2)
plot_Ic_Simulation(Ic, sim)
```

## Description

Box plots of observed (red) and simulated MCI scores by boostraping genes B times, with three horizontal lines: the min, max and 2\*(max-min) value of the state of interests, or all all values.

## Usage

```
plot_MCI_Simulation(
    MCI,
    simulation,
    las = 0,
    order = NULL,
    ylim = NULL,
    main = NULL,
    which2point = NULL,
    ...
)
```

plot\_SS\_Simulation 29

## **Arguments**

MCI	A named vector of max CI scores per state, can be obtained from function getMaxStats.
simulation	A matrix state $\ast$ number of simulated times, can be obtained from function simulationMCI.
las	Numeric in 0, 1, 2, 3; the style of axis labels. Default is 0, meaning labels are parallel. (link to http://127.0.0.1:21580/library/graphics/html/par.html)
order	A vector of state names in the customized order to be plotted, set to NULL by default.
ylim	An integer vector of length 2. Default is NULL.
main	A character vector. The title of the plot. Default is NULL.
which2point	A character (or integer) which state's values were used to set up the three horizontal lines. by default is NULL, indicating the values of all states will be used.
	Other parameters passed to this function

## Value

Return a box plot of MCI(red) and simulated MCI(grey) scores per state.

## Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>

## **Examples**

```
MCI = c(1:3); names(MCI) = c('a', 'b', 'c')
simMCI = matrix(sample(1:100, 9), 3, 3)
row.names(simMCI) = names(MCI)
plot_MCI_Simulation(MCI, simMCI)
```

 ${\tt plot\_SS\_Simulation}$ 

Density plot the leading two distance between any two states from random scores of all states in a system.

## Description

Generate a density plot of Ic score (orBioTIP score) from a simulation, which is the distance between the first-larget and the second-largest random scores. This is an alternative method to estimate the significance of an observed BioTIP (or Ic) score in a system. This measurement makes more sense to evaluate random scores of sample-label shuffling, in which the nature sample-sample correlation within a phenotypic state (or cell subpopulation) was removed.

30 plot\_SS\_Simulation

## Usage

```
plot_SS_Simulation(
   Ic,
   simulation,
   las = 0,
   xlim = NULL,
   ylim = NULL,
   order = NULL,
   main = "1st max - 2nd max",
   ylab = "1st max - 2nd max")
```

## Arguments

Ic	A vector with names of states. If order is not assigned, then plot by the order of this vector.
simulation	A numeric matrix of Ic scores in which rows are states and columns are numbers of simulated times. It can be obtained from simulation_Ic
las	Numeric in 0, 1, 2, 3; the style of axis labels. Default is 0, meaning labels are parallel. (link to http://127.0.0.1:21580/library/graphics/html/par.html).
xlim	An integer vector of length 2. Default is NULL.
ylim	An integer vector of length 2. Default is NULL.
order	Characters of names of Ic to be plotted in a desired order. Default is NULL.
main	A character vector. The title of the plot. Default is NULL.
ylab	titles y axes, as in plot.

## Value

Return a plot of the observed Ic (red) and simulated Ic (grey) scores per state.

## Author(s)

Xinan H Yang <xyang2@uchicago.edu>

## **Examples**

```
sim = matrix(sample(1:10, 9), 3, 3)
row.names(sim) = paste0('state', 1:3)
Ic = c('state1' = 3.4, 'state2' = 15.6, 'state3' = 2)
plot_SS_Simulation(Ic, sim)
```

sd\_selection 31

sd\_selection

Selecting Highly Oscillating Transcripts

### **Description**

sd\_selection pre-selects highly oscillating transcripts from the input dataset df. The dataset must contain multiple sample groups (or 'states'). For each state, the function filters the dataset using a cutoff value for standard deviation. The default cutoff value is 0.01 (i.e., higher than the top 1 percentage standard deviation).

## Usage

```
sd_selection(
    df,
    samplesL,
    cutoff = 0.01,
    method = c("other", "reference", "previous", "itself", "longitudinal reference"),
    control_df = NULL,
    control_samplesL = NULL
)
```

### **Arguments**

df

A numeric matrix or data frame. The rows and columns represent unique transcript IDs (geneID) and sample names, respectively.

samplesL

A list of vectors, whose length is the number of states. Each vector gives the sample names in a state. Note that the vectors (sample names) has to be among the column names of the R object 'df'.

cutoff

A positive numeric value. Default is 0.01. If < 1, automatically selects top x transcripts using the a selecting method (which is either the reference, other stages or previous stage), e.g. by default it will select top 1 percentage of the transcripts.

method

Selection of methods from reference, other, previous, default uses other. Partial match enabled.

- itself, or longitudinal reference. Some specific requirements for each option:
- reference, the reference has to be the first.
- previous, make sure sampleL is in the right order from benign to malign.
- itself, make sure the cutoff is smaller than 1.
- longitudinal reference make sure control\_df and control\_samplesL are not NULL. The row numbers of control\_df is the same as df and all transcripts in df are also in control\_df.

control\_df

A count matrix with unique loci as row names and samples names of control samples as column names, only used for method longitudinal reference

control\_samplesL

A list of characters with stages as names of control samples, required for method 'longitudinal reference'

32 simulationMCI

#### Value

sd\_selection() A list of data frames, whose length is the number of states. The rows in each data frame are the filtered transcripts with highest standard deviation selected from df and based on an assigned cutoff value. Each resulting data frame represents a subset of the raw input df, with the sample ID of the same state in the column.

### Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>

### See Also

```
optimize.sd_selection
```

### **Examples**

```
counts = matrix(sample(1:100, 18), 2, 9)
colnames(counts) = 1:9
row.names(counts) = c('loci1', 'loci2')
cli = cbind(1:9, rep(c('state1', 'state2', 'state3'), each = 3))
colnames(cli) = c('samples', 'group')
samplesL <- split(cli[, 1], f = cli[, 'group'])
test_sd_selection <- sd_selection(counts, samplesL, 0.01)</pre>
```

simulationMCI

Get MCI Scores for randomly selected genes

## **Description**

This function gets the MCI scores for randomly selected features (e.g. transcript ids),

### Usage

```
simulationMCI(
  len,
  samplesL,
  df,
  adjust.size = FALSE,
  B = 1000,
  fun = c("cor", "BioTIP")
)
```

## **Arguments**

len An integer that is the length of genes in the CTS (critical transition signal).

samplesL A list of vectors, whose length is the number of states. Each vector gives the sample names in a state. Note that the vector s (sample names) has to be among

the column names of the R object 'df'.

df A numeric matrix or dataframe of numerics, factor or character. The rows and

columns represent unique transcript IDs (geneID) and sample names, respec-

tively

simulation\_Ic 33

adjust.size	A boolean value indicating if MCI score should be adjust by module size (the
	number of transcripts in the module) or not. Default FALSE.
В	An integer, setting the permutation with B runs. Default is 1000.
fun	A character chosen between ("cor", "BioTIP"), indicating where an adjusted
	correlation matrix will be used to calculated the MCI score.

#### Value

A numeric matrix indicating the MCI scores of permutation. The dimension (row X column) of this matrix is the length of samples L \* B.

### Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>; Xinan H Yang <xyang2@uchicago.edu>

### **Examples**

simulation\_Ic

Calculating random Index of Critical transition (Ic scores) for randomly-selectd genes

## Description

Simulating Ic scores for x randomly selected samples, where x should be the same as the length of identified critical-transition signal (CTS) (e.g., number of genes) and B is self-defined running times.

## Usage

34 simulation\_Ic

#### **Arguments**

obs.x An integer, length of identified CTS. sampleL A list of vectors, whose length is the number of states. Each vector gives the sample names in a state. Note that the vector s (sample names) has to be among the column names of the R object 'df'. counts A numeric matrix or dataframe in which columns are samples and rows are transcripts. Each row needs to have a unique row name (i.e. transcript ID). В An integer, setting the permutation with B runs. Default is 1000. fun An optional character string indicating the R functon to calculate correlations for all possible pairs of columns of a matrix. When using "BioTIP", The method is modified to ignore missing values, analogous to how cor(X, use = "pairwise.complete.obs") works. Note that the "BioTIP" option only function together with shrink = TRUE. shrink A flag specifying whether to shrink the correlation or not. This appraoch uses the method outlined by Schafer and Strimmer in "A Shrinkage Approach to Large-Scale Covariance Matrix Estimation and Implications for Functional Genomics" (2005) Comparing to fun='cor', the 'BioTIP' method without shinkage is modified to ignore missing values, analogous to how cor(X, use = "pairwise.complete.obs") works. An optional character string, when fun=="cor", it gives a method for computing use covariances in the presence of missing values. This must be (an abbreviation of) one of the strings "everything", "all.obs", "complete.obs", "na.or.complete", or "pairwise.complete.obs".

output A string. Please select from 'Ic', 'PCCg', or 'PCCs'. Uses 'Ic' by default.

'PCCg' is the PCC between genes (numerator) and 'PCCs' is PCC between

samples (denominator)

#### Value

A matrix of y rows and B columns where y is the length of sampleL and B is self-defined. Each column is a set of Ic scores calculated for each state

### Author(s)

Zhezhen Wang <zhezhen@uchicago.edu>

#### **Examples**

```
counts = matrix(sample(1:100, 27), 3, 9)
colnames(counts) = 1:9
row.names(counts) = c('loci1', 'loci2', 'loci3')
cli = cbind(1:9, rep(c('state1', 'state2', 'state3'), each = 3))
colnames(cli) = c('samples', 'group')
samplesL <- split(cli[, 1], f = cli[, 'group'])
simulation_Ic(2, samplesL, counts, B = 3, fun="BioTIP", shrink=TRUE)</pre>
```

simulation\_Ic\_sample 35

```
simulation_Ic_sample Calculating (and plot) random Ic scores (Mojtahedi et al. 2016) based on shuffling sample labelling.
```

## **Description**

Run B times of sample-label shuffling to calculate the Ic score, where x should be the same as the length of identified BioTiP and B is self-defined.

## Usage

## Arguments

works.

counts	A numeric matrix or data frame. The rows and columns represent unique transcript IDs (geneID) and sample names, respectively.
sampleNo	An integer of sample size at the tipping-point state.
Ic	A numeric value. Ic score of identified CTS (gene-set), useful when plot is TRUE. Default is NULL.
genes	A character vector of identified CTS gene unique ids.
В	An integer indicating number of times to run this simulation, default 1000.
ylim	An integer vector of length 2. Default is NULL.
main	A character vector. The title of the plot. Default is NULL.
fun	An optional character string indicating the R functon to calculate correlations for all possible pairs of columns of a matrix. When using "BioTIP", The method is modified to ignore missing values, analogous to how cor(X, use = "pairwise.complete.obs") works. Note that the "BioTIP" option only function together with shrink = TRUE.
shrink	A flag specifying whether to shrink the correlation or not. This appraoch uses the method outlined by Schafer and Strimmer in "A Shrinkage Approach to Large-Scale Covariance Matrix Estimation and Implications for Functional Genomics" (2005) Comparing to fun='cor', the 'BioTIP' method without shinkage is modi-

fied to ignore missing values, analogous to how cor(X, use = "pairwise.complete.obs")

use	An optional character string, when fun=="cor", it gives a method for computing covariances in the presence of missing values. This must be (an abbreviation of) one of the strings "everything", "all.obs", "complete.obs", "na.or.complete", or "pairwise.complete.obs".
output	A string. Please select from 'Ic', 'PCCg', or 'PCCs'. Uses 'Ic' by default. 'PCCg' is the PCC between genes (numerator) and 'PCCs' is PCC between samples (denominator)

plot A Bollen value indicating whether a density plot will be plotted.

## Value

A vector of B values of BioTIP (or Ic) scores calculated for the state of interest.

### Author(s)

 $Zhezhen \ Wang \ \verb|\c zhezhen@uchicago.edu>; \ Xinan \ H \ Yang \ \verb|\c xyang2@uchicago.edu>| \\$ 

## **Examples**

# **Index**

```
* datasets
                                                  optimize.sd_selection, 22, 32
    cod, 4
                                                  plot.igraph, 9
    gencode, 6
                                                  plot_Ic_Simulation, 27
    GSE6136_cli, 20
                                                  plot_MCI_Simulation, 28
    GSE6136_matrix, 21
                                                  plot_SS_Simulation, 29
    ILEF, 21
                                                  plotBar_MCI, 24
    intron, 22
                                                  plotIc, 25
* internal
                                                  plotMaxMCI, 26
    BioTIP-package, 2
                                                  sd_selection, 9, 16, 24, 31
avg.cor.shrink, 3
                                                  simulation_Ic, 27, 30, 33
BioTIP (BioTIP-package), 2
                                                  simulation_Ic_sample, 35
BioTIP-package, 2
                                                  simulationMCI, 29, 32
cluster_walktrap, 9
cod, 4
communities, 9
cor.shrink, 5
dist, 9
gencode, 6
getBiotypes, 7
getCluster_methods, 8, 13, 16
getCTS, 9
getIc, 10
getIc.new, 12
getMaxMCImember, 10, 11, 13, 15, 26
getMaxStats, 10, 15, 29
getMCI, 15, 16, 26
getMCI_inner, 17
getNetwork, 8, 18
getReadthrough, 19
GSE6136_cli, 20
GSE6136_matrix, 21
hclust, 9
igraph, 16
ILEF, 21
intron, 22
KMedoids, 9
matplot, 26
```