# Package 'singscore'

April 16, 2019

```
Title Rank-based single-sample gene set scoring method
Description A simple single-sample gene signature scoring method that uses rank-based
     statistics to analyze the sample's gene expression profile. It scores the expression activities of
     gene sets at a single-sample level.
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```

Type Package

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generateNull

Permutation test for the derived scores of each sample

# **Description**

This function generates a number of random gene sets that have the same number of genes as the scored gene set. It scores each random gene set and returns a matrix of scores for all samples. The empirical scores are used to calculate the empirical p-values and plot the null distribution. The implementation uses BiocParallel::bplapply() for easy access to parallel backends. Note that one should pass the same values to the upSet, downSet, centerScore and bidirectional arguments as what they provide for the simpleScore() function to generate a proper null distribution.

# Usage

```
generateNull(upSet, downSet = NULL, rankData, centerScore = TRUE,
   knownDirection = TRUE, B = 1000, ncores = 1,
   seed = sample.int(1e+06, 1), useBPPARAM = NULL)

## S4 method for signature 'vector,missing'
generateNull(upSet, downSet = NULL, rankData,
   centerScore = TRUE, knownDirection = TRUE, B = 1000, ncores = 1,
   seed = sample.int(1e+06, 1), useBPPARAM = NULL)

## S4 method for signature 'GeneSet,missing'
generateNull(upSet, downSet = NULL, rankData,
   centerScore = TRUE, knownDirection = TRUE, B = 1000, ncores = 1,
   seed = sample.int(1e+06, 1), useBPPARAM = NULL)
```

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```
## S4 method for signature 'vector,vector'
generateNull(upSet, downSet = NULL, rankData,
   centerScore = TRUE, knownDirection = TRUE, B = 1000, ncores = 1,
   seed = sample.int(1e+06, 1), useBPPARAM = NULL)

## S4 method for signature 'GeneSet,GeneSet'
generateNull(upSet, downSet = NULL, rankData,
   centerScore = TRUE, knownDirection = TRUE, B = 1000, ncores = 1,
   seed = sample.int(1e+06, 1), useBPPARAM = NULL)
```

#### **Arguments**

upSet GeneSet object or a vector of gene Ids, up-regulated gene set downSet GeneSet object or a vector of gene Ids, down-regulated gene

rankData matrix, outcome of function rankGenes()

centerScore A Boolean, specifying whether scores should be centered around 0, default as

**TRUE** 

knownDirection A boolean flag, it deterimines whether the scoring method should derive the

scores in a directional mannar when the gene signature only contains one set of gene set (passing the gene set via upSet). It is default as TRUE but one can set the argument to be FALSE to derive the score for a single gene set in a undirectional way. This parameter becomes irrelevant when both upSet and

downSet are provided.

B integer, the number of permutation repeats or the number of random gene sets

to be generated, default as 1000

ncores, integer, the number of CPU cores the function can use

seed integer, set the seed for randomisation

useBPPARAM, the backend the function uses, if NULL is provided, the function uses the default

parallel backend which is the first on the list returned by BiocParallel::registered()

i.e BiocParallel::registered()[[1]] for your machine. It can be changed

explicitly by passing a BPPARAM

#### Value

A matrix of empirical scores for all samples

#### Author(s)

Ruqian Lyu

#### See Also

Post about BiocParallel browseVignettes("BiocParallel")

# Examples

```
ranked <- rankGenes(toy_expr_se)
scoredf <- simpleScore(ranked, upSet = toy_gs_up, downSet = toy_gs_dn)
# find out what backends can be registered on your machine</pre>
```

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```
BiocParallel::registered()
# the first one is the default backend
# ncores = ncores <- parallel::detectCores() - 2
permuteResult = generateNull(upSet = toy_gs_up, downSet = toy_gs_dn, ranked, centerScore = TRUE, B =10, seed = 1, ncores = 1)</pre>
```

getPvals

Estimate the empirical p-values

# **Description**

With null distributions estimated using the generateNull() function, p-values are estimated using a one-tailed test. A minimum p-value of 1/B can be achieved with B permutations.

#### Usage

```
getPvals(permuteResult, scoredf)
```

# **Arguments**

permuteResult A matrix, null distributions for each sample generated using the generateNull()

function

scoredf A dataframe, the scored results of samples under test generated using the simpleScore()

function

#### Value

Estimated p-values for enrichment of the signature in each sample. A p-value of 1/B indicates that the estimated p-value is less than or equal to 1/B.

# **Examples**

```
ranked <- rankGenes(toy_expr_se)
scoredf <- simpleScore(ranked, upSet = toy_gs_up, downSet = toy_gs_dn)
# find out what backends can be registered on your machine
BiocParallel::registered()
# the first one is the default backend, and it can be changed explicitly.
# See vignette for more details
permuteResult = generateNull(upSet = toy_gs_up, downSet = toy_gs_dn, ranked,
B =10, seed = 1, useBPPARAM = NULL)
# call the permutation function to generate the empirical scores
# for B times.
pvals <- getPvals(permuteResult,scoredf)</pre>
```

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plotDispersion	Plot the score v.s. despersion for all samples	

# **Description**

This function takes the output from the simpleScore() function and generates scatter plots of score vs. dispersion for the total score, the up score and the down score of samples. If you wish to use the plotting function but with some customized inputs (instead of outputs from simpleScore function), you need to make sure the formats are the same. To be specific, you need to have columns names "TotalScore" "TotalDispersion" "UpScore" "UpDispersion" "DownScore" "DownDispersion" and rows names as samples.

# Usage

```
plotDispersion(scoredf, annot = NULL, alpha = 1, size = 1,
  textSize = 1.5, isInteractive = FALSE)
```

# **Arguments**

scoredf	data.frame, generated using the simpleScore() function
annot	annot any numeric or factor annotation provided by the user that needs to be plot. Annotations must be ordered in the same way as the scores
alpha	numeric, set the transparency of points
size	numeric, set the size of each point
textSize	numeric, relative text sizes for title, labels, and axis values
isInteractive	Boolean, determine whether the plot is interactive

#### Value

A ggplot object

# Examples

```
ranked <- rankGenes(toy_expr_se)
scoredf <- simpleScore(ranked, upSet = toy_gs_up, downSet = toy_gs_dn)
plotDispersion(scoredf)
plotDispersion(scoredf, isInteractive = TRUE)</pre>
```

plotNull Plot the empirically estimated null distribution and associated p-values

# Description

This function takes the results from function <code>generateNull()</code> and plots the density curves of permuted scores for the provided samples via <code>sampleNames</code> parameter. It can plot null distribution(s) for a single sample or multiple samples.

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### Usage

```
plotNull(permuteResult, scoredf, pvals, sampleNames = NULL,
  cutoff = 0.01, textSize = 2, labelSize = 5)
```

# **Arguments**

permuteResult A matrix, null distributions for each sample generated using the generateNull()

function

scoredf A dataframe, singscores generated using the simpleScore() function

pvals A vector, estimated p-values using the getPvals() function permuteResult,scoredf

and pvals are the results for the same samples.

sampleNames A character vector, sample IDs for which null distributions will be plotted

cutoff numeric, the cutoff value for determining significance textSize numeric, size of axes labels, axes values and title

labelSize numeric, size of label texts

#### Value

a ggplot object

#### Author(s)

Ruqian Lyu

#### **Examples**

```
ranked <- rankGenes(toy_expr_se)
scoredf <- simpleScore(ranked, upSet = toy_gs_up, downSet = toy_gs_dn)
# find out what backends can be registered on your machine
BiocParallel::registered()
# the first one is the default backend, and it can be changed explicitly.
permuteResult = generateNull(upSet = toy_gs_up, downSet = toy_gs_dn, ranked,
B =10, seed = 1,useBPPARAM = NULL)
# call the permutation function to generate the empirical scores
#for B times.
pvals <- getPvals(permuteResult,scoredf)
# plot for all samples
plotNull(permuteResult,scoredf,pvals,sampleNames = names(pvals))
#plot for the first sample
plotNull(permuteResult,scoredf,pvals,sampleNames = names(pvals)[1])</pre>
```

plotRankDensity

Plot the densities of ranks for one sample

#### **Description**

This function takes a single-column data frame, which is a single-column subset of the ranked matrix data generated using rankGenes() function, and the gene sets of interest as inputs. It plots the density of ranks for genes in the gene set and overlays a barcode plot of these ranks. Ranks are normalized by dividing them by the maximum rank. Densities are estimated using KDE.

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#### Usage

```
plotRankDensity(rankData, upSet, downSet = NULL, isInteractive = FALSE,
    textSize = 1.5)

## S4 method for signature 'ANY,vector,missing'
plotRankDensity(rankData, upSet,
    downSet = NULL, isInteractive = FALSE, textSize = 1.5)

## S4 method for signature 'ANY,GeneSet,missing'
plotRankDensity(rankData, upSet,
    downSet = NULL, isInteractive = FALSE, textSize = 1.5)

## S4 method for signature 'ANY,vector,vector'
plotRankDensity(rankData, upSet,
    downSet = NULL, isInteractive = FALSE, textSize = 1.5)

## S4 method for signature 'ANY,GeneSet,GeneSet'
plotRankDensity(rankData, upSet,
    downSet = NULL, isInteractive = FALSE, textSize = 1.5)
```

#### **Arguments**

rankData one column of the ranked gene expression matrix obtained from the rankGenes()

function, use drop = FALSE when subsetting the ranked gene expression matrix,

see examples.

upSet GeneSet object or a vector of gene Ids, up-regulated gene set
downSet GeneSet object or a vector of gene Ids, down-regulated gene set
isInteractive Boolean, determine whether the returned plot is interactive

textSize numberic, set the size of text on the plot

#### Value

A ggplot object (or a plotly object) with a rank density plot overlayed with a barcode plot

# **Examples**

```
ranked <- rankGenes(toy_expr_se)
plotRankDensity(ranked[,2,drop = FALSE], upSet = toy_gs_up)</pre>
```

plotScoreLandscape

Plot landscape of two gene signatures scores

#### **Description**

This function takes two data frames which are outputs from the simpleScore() function and plots the relationship between the two gene set scores for samples in the gene expression matrix.Scoredf1 and Scoredf2 are two scoring results of the same set of samples against two different gene signatures. If you wish to use the plotting function but with some customized inputs (instead of outputs from the simpleScore function), you need to make sure the formats are the same To be specific, you need to have column names "TotalScore" "TotalDispersion" "UpScore" "UpDispersion" "DownScore" "DownDispersion" and rows names as samples.

#### Usage

```
plotScoreLandscape(scoredf1, scoredf2, scorenames = c(),
  textSize = 1.5, isInteractive = FALSE, hexMin = 100)
```

#### **Arguments**

scoredf1 data.frame, result of the simpleScore() function which scores the gene expres-

sion matrix against a gene set of interest

scoredf2 data.frame, result of the simpleScore() function which scores the gene expres-

sion matrix against another gene set of interest

scorenames character vector of length 2, names for the two scored gene set/signatures stored

in scoredf1 and scoredf2

textSize numeric, set the text size for the plot, default as 1.5

isInteractive boolean, whether the plot is interactive default as FALSE

hexMin integer, the threshold which decides whether hex bin plot or scatter plot is dis-

played, default as 100

#### Value

A ggplot object, a scatter plot, demonstrating the relationship between scores from two signatures on the same set of samples.

# **Examples**

```
ranked <- rankGenes(toy_expr_se)
scoredf <- simpleScore(ranked, upSet = toy_gs_up, downSet = toy_gs_dn)
scoredf2 <- simpleScore(ranked, upSet = toy_gs_up)
plotScoreLandscape(scoredf, scoredf2)</pre>
```

#### **Description**

This function takes the output (ggplot object) of the function plotScoreLandscape() and a new dataset. It projects the new data points onto the landscape plot and returns a new ggplot object with projected data points.

#### Usage

```
projectScoreLandscape(plot0bj = NULL, scoredf1, scoredf2,
   subSamples = NULL, sampleLabels = NULL, annot = NULL,
   isInteractive = FALSE)
```

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#### **Arguments**

plotObj	a dataframe, resulted from plotScoreLandscape()
scoredf1	data.frame, result of the simpleScore() function which scores the gene expression matrix against a gene set of interest
scoredf2	data.frame, result of the simpleScore() function which scores the gene expression matrix against another gene set of interest. Scores in scoredf1 and scoredf2 consist of the new data points that will be projected on the plotObj landscape plot.
subSamples	vector of character or indices for subsetting the scoredfs, default as NULL and all samples in scoredfs will be plotted. The subsetted samples are projected onto the landscape plot of plot0bj.
sampleLabels	vector of character, sample names to display, ordered in the same way as samples are ordered in the 'scoredfs' data matrix, default as NULL which means the projected points are not labelled.
annot	vector of characters, annotations used to colour the data and should have the same number of samples as in scoredfs
isInteractive	boolean, whether the plot is interactive default as FALSE

#### Value

New data points on the already plotted ggplot object from plotScoreLanscape()

#### See Also

```
plotScoreLandscape()
```

#### **Examples**

```
ranked <- rankGenes(toy_expr_se)
scoredf1 <- simpleScore(ranked, upSet = toy_gs_up, downSet = toy_gs_dn)
scoredf2 <- simpleScore(ranked, upSet = toy_gs_up)
psl <- plotScoreLandscape(scoredf1, scoredf2)
projectScoreLandscape(psl,scoredf1, scoredf2)</pre>
```

rankGenes

Rank genes by the gene expression intensities

# **Description**

The rankGenes function is a generic function that can deal with mutilple types of inputs. Given a matrix of gene expression that has samples in columns, genes in rows, and values being gene expression intensity, rankGenes ranks gene expression intensities in each sample.

It can also work with S4 objects that have gene expression matrix as a component (i.e ExpressionSet, DGEList,SummarizedExperiment). It calls the rank function in the base package which ranks the gene expression matrix by its absolute expression level. If the input is S4 object of DGEList, ExpressionSet, or SummarizedExperiment, it will extract the gene expression matrix from the object and rank the genes. The default 'tiesMethod' is set to 'min'.

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#### Usage

# Arguments

expreMatrix A gene expression matrix (matrix,data.frame) or S4 object (ExpressionSet,DGEList,

SummarizedExperiment)

tiesMethod A character indicating what method to use when dealing with ties

# Value

The ranked gene expression matrix that has samples in columns and genes in rows

### See Also

rank ExpressionSet SummarizedExperiment DGEList

#### **Examples**

```
rankGenes(toy_expr_se) # toy_expr_se is a gene expression dataset
tiesMethod = 'min'
# get counts from toy_expr_se
counts <- SummarizedExperiment::assay(toy_expr_se)
# or it can be a ExpressionSet object
e <- Biobase::ExpressionSet(assayData = as.matrix(counts))
rankGenes(e)</pre>
```

scoredf\_ccle\_epi

Pre-computed scores of the CCLE dataset against an epithelial gene signature

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### **Description**

This data.frame stores pre-computed scores of the CCLE dataset Barretina et al calculated using the simpleScore() function against the epithelial gene signature from Tan, Tuan Zea et al. The data.frame has scores for 55 samples. Please refer to the vignettes for instructions on how to obtain the full datasets.

#### Usage

scoredf\_ccle\_epi

#### **Format**

An object of class data. frame with 55 rows and 2 columns.

#### References

Barretina, Jordi, Giordano Caponigro, Nicolas Stransky, Kavitha Venkatesan, Adam A Margolin, Sungjoon Kim, Christopher J Wilson, et al. 2012. "The Cancer Cell Line Encyclopedia Enables Predictive Modelling of Anticancer Drug Sensitivity." Nature 483 (7391): 603–7.

Tan, Tuan Zea, Qing Hao Miow, Yoshio Miki, Tetsuo Noda, Seiichi Mori, Ruby Yun-Ju Huang, and Jean Paul Thiery. 2014–10AD. "Epithelial-Mesenchymal Transition Spectrum Quantification and Its Efficacy in Deciphering Survival and Drug Responses of Cancer Patients." EMBO Molecular Medicine 6 (10). Oxford, UK: BlackWell Publishing Ltd: 1279–93. doi:10.15252/emmm.201404208.

#### See Also

scoredf\_ccle\_mes

scoredf\_ccle\_mes

Pre-computed scores of the CCLE dataset against a mesenchymal gene signature

# Description

This data.frame stores pre-computed scores of the CCLE dataset Barretina et al calculated using the simpleScore() function against the mesenchymal gene signature from Tan, Tuan Zea et al. The data.frame has scores for 55 samples. Please refer to the vignettes for instructions on how to obtain the full datasets.

# Usage

scoredf\_ccle\_mes

#### **Format**

An object of class data. frame with 55 rows and 2 columns.

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#### References

Barretina, Jordi, Giordano Caponigro, Nicolas Stransky, Kavitha Venkatesan, Adam A Margolin, Sungjoon Kim, Christopher J Wilson, et al. 2012. "The Cancer Cell Line Encyclopedia Enables Predictive Modelling of Anticancer Drug Sensitivity." Nature 483 (7391): 603–7.

Tan, Tuan Zea, Qing Hao Miow, Yoshio Miki, Tetsuo Noda, Seiichi Mori, Ruby Yun-Ju Huang, and Jean Paul Thiery. 2014–10AD. "Epithelial-Mesenchymal Transition Spectrum Quantification and Its Efficacy in Deciphering Survival and Drug Responses of Cancer Patients." EMBO Molecular Medicine 6 (10). Oxford, UK: BlackWell Publishing Ltd: 1279–93 doi:10.15252/emmm.201404208.

#### See Also

scoredf\_ccle\_epi

scoredf\_tcga\_epi

Pre-computed scores of the TCGA breast cancer gene expression matrix against an epithelial signature

# **Description**

This data frame stores pre-computed scores of the TCGA dataset calculated using the simpleScore() function against the epithelial gene signature from Tan, Tuan Zea et al. Please refer to the vignettes for instructions on how to obtain the full datasets.

# Usage

```
scoredf_tcga_epi
```

#### **Format**

An object of class data. frame with 1119 rows and 2 columns.

# References

Tan, Tuan Zea, Qing Hao Miow, Yoshio Miki, Tetsuo Noda, Seiichi Mori, Ruby Yun-Ju Huang, and Jean Paul Thiery. 2014–10AD. "Epithelial-Mesenchymal Transition Spectrum Quantification and Its Efficacy in Deciphering Survival and Drug Responses of Cancer Patients." EMBO Molecular Medicine 6 (10). Oxford, UK: BlackWell Publishing Ltd: 1279–93 doi:10.15252/emmm.201404208.

#### See Also

scoredf\_tcga\_mes

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scoredf_tcga_mes	Pre-computed scores of the TCGA breast cancer gene expression matrix against a mesenchymal signature

#### **Description**

This data.frame stores pre-computed scores of the TCGA dataset calculated using the simpleScore() function against the mesenchymal gene signature from Tan, Tuan Zea et al. Please refer to the vignettes for instructions on how to obtain the full datasets.

# Usage

```
scoredf_tcga_mes
```

#### **Format**

An object of class data. frame with 1119 rows and 2 columns.

#### References

Tan, Tuan Zea, Qing Hao Miow, Yoshio Miki, Tetsuo Noda, Seiichi Mori, Ruby Yun-Ju Huang, and Jean Paul Thiery. 2014–10AD. "Epithelial-Mesenchymal Transition Spectrum Quantification and Its Efficacy in Deciphering Survival and Drug Responses of Cancer Patients." EMBO Molecular Medicine 6 (10). Oxford, UK: BlackWell Publishing Ltd: 1279–93 doi:10.15252/emmm.201404208.

#### See Also

```
scoredf_tcga_epi
```

simpleScore sing	le-sample gene-se	t scoring method
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# **Description**

This function computes 'singscores' using a ranked gene expression matrix obtained from the rankGenes() function and a gene set or a pair of up-regulated and down-regulated gene sets. It returns a data.frame of scores and dispersions for each sample. The gene sets can be in vector format or as GeneSet objects (from GSEABase packages). If samples need to be scored against a single gene set, the upSet argument should be used to pass the gene set while the downSet argument is set to NULL. This setting is ideal for gene sets representing gene ontologies where the nature of the genes is unknown (up- or down-regulated).

#### Usage

```
simpleScore(rankData, upSet, downSet = NULL, subSamples = NULL,
  centerScore = TRUE, dispersionFun = mad, knownDirection = TRUE)
## S4 method for signature 'ANY, vector, missing'
simpleScore(rankData, upSet,
  downSet = NULL, subSamples = NULL, centerScore = TRUE,
```

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```
dispersionFun = mad, knownDirection = TRUE)

## S4 method for signature 'ANY,GeneSet,missing'
simpleScore(rankData, upSet,
   downSet = NULL, subSamples = NULL, centerScore = TRUE,
   dispersionFun = mad, knownDirection = TRUE)

## S4 method for signature 'ANY,vector,vector'
simpleScore(rankData, upSet,
   downSet = NULL, subSamples = NULL, centerScore = TRUE,
   dispersionFun = mad, knownDirection = TRUE)

## S4 method for signature 'ANY,GeneSet,GeneSet'
simpleScore(rankData, upSet,
   downSet = NULL, subSamples = NULL, centerScore = TRUE,
   dispersionFun = mad, knownDirection = TRUE)
```

#### **Arguments**

rankData A matrix object, ranked gene expression matrix data generated using the rankGenes()

function

upSet A GeneSet object or character vector of gene IDs of up-regulated gene set or a

gene set where the nature of genes is not known

downSet A GeneSet object or character vector of gene IDs of down-regulated gene set or

NULL where only a single gene set is provided

subSamples A vector of sample labels/indices that will be used to subset the rankData matrix.

All samples will be scored if not provided

centerScore A Boolean, specifying whether scores should be centered around 0, default as

TRUE

dispersionFun A function, dispersion function with default being mad

knownDirection A boolean flag, it deterimines whether the scoring method should derive the

scores in a directional mannar when the gene signature only contains one set of gene set (passing the gene set via upSet). It is default as TRUE but one can set the argument to be FALSE to derive the score for a single gene set in a undirectional way. This parameter becomes irrelevant when both upSet and

downSet are provided.

#### Value

A data frame consists of singscores and dispersions for all samples

#### See Also

```
rank "GeneSet"
```

## **Examples**

```
ranked <- rankGenes(toy_expr_se)
scoredf <- simpleScore(ranked, upSet = toy_gs_up, downSet = toy_gs_dn)
# toy_gs_up is a GeneSet object, alternatively a vector of gene ids may also
# be supplied.</pre>
```

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singscore	singscore: A package for deriving gene-set scores at a single sample level

# **Description**

The package provides functions for calculating gene-set enrichment scores at a single-sample level using gene expression data. It includes functions to perform hypothesis testing and provides visualisations to enable diagnosis of scores and gene sets along with visualisations to enable exploration of results.

tgfb\_expr\_10\_se

An example gene expression dataset

# Description

A microarray gene expression dataset that was originally obtained from the integrated TGFb-EMT data published by (Foroutan et al, 2017). (ComBat corrected values). tgfb\_expr\_10 is a subset of the integrated TGFb-EMT data consisting of 10 samples (4 TGFb treated and 6 controls) each with expression values for 11900 genes.

# Usage

```
tgfb_expr_10_se
```

# **Format**

A SummarizedExperiment object

# Source

Foroutan et al,2017

# References

Foroutan, Momeneh, Joseph Cursons, Soroor Hediyeh-Zadeh, Erik W Thompson, and Melissa J Davis. 2017. "A Transcriptional Program for Detecting Tgfbeta-Induced Emt in Cancer." Molecular Cancer Research. American Association for Cancer Research. doi:10.1158/1541-7786.MCR-16-0313.

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tgfb_gs_dn	Gene set of down-regulated genes for the TGFb-induced EMT gene signature
------------	--

# **Description**

A GeneSet object that contains the down-regulated genes of the TGFb-induced EMT gene signature that was derived by (Foroutan et al,2017), using two meta-analysis techniques. The gene signature contains an up-regulated gene set (up-set) and a down-regulated gene set (down-set). Please refer to the vignettes for the steps to acquire the exact data object.

# Usage

tgfb\_gs\_dn

#### **Format**

A GeneSet object

#### **Source**

Foroutan et al,2017

# References

Foroutan, Momeneh, Joseph Cursons, Soroor Hediyeh-Zadeh, Erik W Thompson, and Melissa J Davis. 2017. "A Transcriptional Program for Detecting Tgfbeta-Induced Emt in Cancer." Molecular Cancer Research. American Association for Cancer Research. doi:10.1158/1541-7786.MCR-16-0313.

#### See Also

```
"GeneSet",tgfb_gs_up
```

tgfb_gs_up	Gene set of up-regulated genes for the TGFb-induced EMT gene sig-
	nature

# Description

A GeneSet object that contains the up-regulated genes of the TGFb-induced EMT gene signature that was derived by (Foroutan et al.,2017), using two meta-analysis techniques. The gene signature contains an up-regulated gene set (up-set) and a down-regulated gene set (down-set). Please refer to the vignettes for the steps to acquire the exact data object.

# Usage

tgfb\_gs\_up

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#### **Format**

A GeneSet object

#### **Source**

Foroutan et al,2017

#### References

Foroutan, Momeneh, Joseph Cursons, Soroor Hediyeh-Zadeh, Erik W Thompson, and Melissa J Davis. 2017. "A Transcriptional Program for Detecting Tgfbeta-Induced Emt in Cancer." Molecular Cancer Research. American Association for Cancer Research. doi:10.1158/1541-7786.MCR-16-0313.

#### See Also

```
"GeneSet",tgfb_gs_dn
```

toy\_expr\_se

A toy gene expression dataset of two samples

# **Description**

A toy dataset consisting of 2 samples with the expression values of 20 genes. The data was created by sampling 2 samples and 20 genes from the dataset by Foroutan et al, 2017.

# Usage

toy\_expr\_se

## **Format**

A SummarizedExperiment of 2 samples each with 20 genes

D\_Ctrl\_R1 a control sample

D\_TGFb\_R1 a TGFb-treated sample

#### **Source**

Foroutan et al.,2017

# References

Foroutan, Momeneh, Joseph Cursons, Soroor Hediyeh-Zadeh, Erik W Thompson, and Melissa J Davis. 2017. "A Transcriptional Program for Detecting Tgfbeta-Induced Emt in Cancer." Molecular Cancer Research. American Association for Cancer Research. doi:10.1158/1541-7786.MCR-16-0313.

18 toy\_gs\_up

toy\_gs\_dn

A gene set object of down-regulated genes for the toy dataset

# Description

A GeneSet object with 5 genes randomly selected from the toy dataset. These genes are independent of those in toy\_gs\_up

# Usage

```
toy_gs_dn
```

#### **Format**

A GSEABase::GeneSet object with 5 genes

#### See Also

```
"GeneSet",toy_expr_se,toy_gs_up
```

toy\_gs\_up

A gene set object of up-regulated genes for the toy dataset

# Description

A GeneSet object with 5 genes randomly selected from the toy dataset. These genes are independent of those in  $toy_gs_dn$ 

# Usage

```
toy_gs_up
```

# **Format**

A GeneSet object with 5 genes

## See Also

```
\hbox{\tt "GeneSet",} toy\_expr\_se, toy\_gs\_dn
```

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