# Package 'vissE'

July 4, 2025

Title Visualising Set Enrichment Analysis Results

Version 1.17.0

**Description** This package enables the interpretation and analysis of results from a gene set enrichment analysis using network-based and text-mining approaches. Most enrichment analyses result in large lists of significant gene sets that are difficult to interpret. Tools in this package help build a similarity-based network of significant gene sets from a gene set enrichment analysis that can then be investigated for their biological function using text-mining approaches.

biocViews Software, GeneExpression, GeneSetEnrichment, NetworkEnrichment, Network

License GPL-3

**Encoding** UTF-8

LazyDataCompression bzip2

**Roxygen** list(markdown = TRUE)

RoxygenNote 7.2.3

**Depends** R (>= 4.1)

- **Imports** igraph, methods, plyr, ggplot2, scico, RColorBrewer, tm, ggwordcloud, GSEABase, reshape2, grDevices, ggforce, msigdb, ggrepel, textstem, tidygraph, stats, scales, ggraph
- **Suggests** testthat, org.Hs.eg.db, org.Mm.eg.db, patchwork, singscore, knitr, rmarkdown, prettydoc, BiocStyle, pkgdown, covr

URL https://davislaboratory.github.io/vissE

BugReports https://github.com/DavisLaboratory/vissE/issues

VignetteBuilder knitr

git\_url https://git.bioconductor.org/packages/vissE

git\_branch devel

git\_last\_commit f662525

git\_last\_commit\_date 2025-04-15

**Repository** Bioconductor 3.22

Date/Publication 2025-07-03

Author Dharmesh D. Bhuva [aut, cre] (ORCID: <https://orcid.org/0000-0002-6398-9157>), Ahmed Mohamed [ctb]

Maintainer Dharmesh D. Bhuva <bhuva.d@wehi.edu.au>

# Contents

vissE-package	2
bhuvad_theme	3
characteriseGeneset	3
computeMsigNetwork	4
computeMsigOverlap	5
computeMsigWordFreq	6
findMsigClusters	7
getMsigExclusionList	8
hgsc	8
plotGeneStats	9
plotMsigNetwork	10
plotMsigPPI	11
plotMsigWordcloud	
	15

# Index

vissE-package

vissE: Visualising Set Enrichment Analysis Results

#### Description

This package enables the interpretation and analysis of results from a gene set enrichment analysis using network-based and text-mining approaches. Most enrichment analyses result in large lists of significant gene sets that are difficult to interpret. Tools in this package help build a similarity-based network of significant gene sets from a gene set enrichment analysis that can then be investigated for their biological function using text-mining approaches.

# Details

This package supports four workflows to enhance gene set enrichment analysis:

- 1. Clustering results from a gene set enrichment analysis (e.g. using limma::fry, singscore or GSEA). The functions required for this analysis are computeMsigOverlap, computeMsigNetwork and plotMsigNetwork.
- 2. Interpreting gene set clusters (identified in the first analysis) by performing text-mining of gene set names and descriptions. The main function required to perform text-mining of gene sets is plotMsigWordcloud. Other functions can be used to access intermmediate results.
- 3. Visualise gene-level statistics for gene set clusters identified in the first analysis to link back gene set clusters to the genes of interest. This can be done using the plotGeneStats function.
- 4. Identifying gene sets similar to a list of genes identified from a DE analysis using set overlap measures. This can be done using the characteriseGeneset function.

#### Author(s)

Maintainer: Dharmesh D. Bhuva <bhuva.d@wehi.edu.au> (ORCID)

Other contributors:

• Ahmed Mohamed <mohamed.a@wehi.edu.au> [contributor]

#### bhuvad\_theme

#### See Also

Useful links:

- https://davislaboratory.github.io/vissE
- Report bugs at https://github.com/DavisLaboratory/vissE/issues

bhuvad_theme	Custom theme	

# Description

Custom theme

# Usage

 $bhuvad_theme(rl = 1.1)$ 

# Arguments

rl

a numeric, scaling factor to apply to text sizes

# Value

a ggplot2 theme

#### Examples

p1 = ggplot2::ggplot()
p1 + bhuvad\_theme()

characteriseGeneset Functionally characterise a list of genes

# Description

This function can be used to perform a network-based enrichment analysis of a list of genes. The list of genes are characterised based on their similarity with gene sets from the MSigDB. A network of similar gene sets is retrieved using this function.

# Usage

```
characteriseGeneset(
  gs,
  thresh = 0.2,
  measure = c("ovlapcoef", "jaccard"),
  gscolcs = c("h", "c2", "c5"),
  org = c("auto", "hs", "mm")
)
```

# Arguments

gs	a GeneSet object, representing the list of genes that need to be characterised.
thresh	a numeric, specifying the threshold to discard pairs of gene sets.
measure	a character, specifying the similarity measure to use: ari for the Adjusted Rand Index, jaccard for the Jaccard Index and ovlapcoef for the Overlap Coeffi- cient.
gscolcs	a character, listing the MSigDB collections to use as a background (defaults to h, c2, and c5). Collection types can be retrieved using msigdb::listCollections().
org	a character, specifying the organism to use. This can either be "auto" (default), "hs" or "mm".

# Value

an igraph object, containing gene sets that are similar to the query set. The network contains relationships between results of the query too.

# Examples

```
library(GSEABase)
data(hgsc)

#create a geneset using one of the Hallmark gene sets
mySet <- GeneSet(
  geneIds(hgsc[[2]]),
  setName = 'MySet',
  geneIdType = SymbolIdentifier()
)

#characterise the custom gene set
ig <- characteriseGeneset(mySet)
plotMsigNetwork(ig)</pre>
```

computeMsigNetwork Compute a network using computed gene set overlap

# Description

Computes an igraph object using information on gene sets and gene sets computed using the computeMsigOverlap() function.

# Usage

```
computeMsigNetwork(genesetOverlap, msigGsc)
```

# Arguments

genesetOverlap	a data.frame, containing results of an overlap analysis computed using the computeMsigOverlap()
	function.
msigGsc	a GeneSetCollection object, containing gene sets used to compute overlap.

#### computeMsigOverlap

#### Value

an igraph object

# Examples

```
data(hgsc)
ovlap <- computeMsigOverlap(hgsc)
ig <- computeMsigNetwork(ovlap, hgsc)</pre>
```

computeMsigOverlap Compute gene set overlap

# Description

Compute overlap between gene sets from a GeneSetCollection using the Jaccard index or the overlap coefficient. These values can then be used to compute a network of gene set overlaps.

# Usage

```
computeMsigOverlap(
  msigGsc1,
  msigGsc2 = NULL,
  thresh = 0.25,
  measure = c("ari", "jaccard", "ovlapcoef")
)
```

# Arguments

msigGsc1	a GeneSetCollection object.
msigGsc2	a GeneSetCollection object or NULL if pairwise overlaps are to be computed.
thresh	a numeric, specifying the threshold to discard pairs of gene sets.
measure	a character, specifying the similarity measure to use: ari for the Adjusted Rand Index, jaccard for the Jaccard Index and ovlapcoef for the Overlap Coeffi- cient.

# Value

a data.frame, containing the overlap structure of gene sets represented as a network in the simple interaction format (SIF).

#### Examples

```
data(hgsc)
ovlap <- computeMsigOverlap(hgsc)</pre>
```

computeMsigWordFreq Compute word frequencies for a single MSigDB collection

# Description

Compute word frequencies for a single MSigDB collection

# Usage

```
computeMsigWordFreq(
  msigGsc,
  weight = NULL,
  measure = c("tfidf", "tf"),
  version = msigdb::getMsigdbVersions(),
  org = c("auto", "hs", "mm"),
  rmwords = getMsigExclusionList(),
  idf = NULL
)
```

# Arguments

msigGsc	a GeneSetCollection object, containing gene sets from the MSigDB. The GSEABase::getBroadSets function can be used to parse XML files downloaded from MSigDB.
weight	a named numeric vector, containing weights to apply to each gene-set. This can be -log10(FDR), -log10(p-value) or an enrichment score (ideally unsigned).
measure	a character, specifying how frequencies should be computed. "tf" uses term frequencies and "tfidf" (default) applies inverse document frequency weights to term frequencies.
version	a character, specifying the version of msigdb to use (see msigdb::getMsigdbVersions()).
org	a character, specifying the organism to use. This can either be "auto" (default), "hs" or "mm".
rmwords	a character vector, containing an exclusion list of words to discard from the analysis.
idf	a list of named numeric vectors, specifying inverse document frequencies to use to penalise terms from gene-set names and short descriptions. This should be a vector of length 2 with names "Name" and "Short". Numeric vectors should contain weights and names should represent the term. Precomputed versions can be retrieved using the msigdb::getMsigdbIDF().

# Value

a list, containing two data.frames summarising the results of the frequency analysis on gene set names and short descriptions.

# Examples

```
data(hgsc)
freq <- computeMsigWordFreq(hgsc, measure = 'tfidf')</pre>
```

findMsigClusters Identify gene-set clusters from a gene-set overlap network

# Description

This function identifies gene-set clusters from a gene-set overlap network produced using vissE. Various graph clustering algorithms from the igraph package can be used for clustering. Gene-set clusters identified are then sorted based on their size and a given statistic of interest (absolute of the statistic is maximised per cluster).

#### Usage

```
findMsigClusters(
    ig,
    genesetStat = NULL,
    minSize = 2,
    alg = igraph::cluster_walktrap,
    algparams = list()
)
```

#### Arguments

ig	an igraph object, containing a network of gene set overlaps computed using computeMsigNetwork().
genesetStat	a named numeric, containing statistics for each gene-set that are to be used in cluster prioritisation. If NULL, clusters are prioritised based on their size (number of gene-sets in them).
minSize	a numeric, stating the minimum size a cluster can be (default is 2).
alg	a function, from the igraph package that should be used to perform graph- clustering (default is igraph::cluster_walktrap). The function should pro- duce a communities object.
algparams	a list, specifying additional parameters that are to be passed to the graph cluster- ing algorithm.

#### Details

Gene-sets clusters are identified using graph clustering and are prioritised based on a combination of cluster size and optionally, a statistic of interest (e.g., enrichment scores). A product-of-ranks approach is used to prioritise clusters when gene-set statistics are available. In this approach, clusters are ranked based on their cluster size (largest to smallest) and on the median absolute statistic of gene-sets within it (largest to smallest). The product of these ranks is computed and clusters are ranked based on these product-of-rank statistic (smallest to largest).

When prioritising using cluster size and gene-set statistics, if statistics for some gene-sets in the network are missing, only the size is used in cluster prioritisation.

#### Value

a list, containing gene-sets that belong to each cluster. Items in the list are organised based on prioritisation.

# Examples

```
data(hgsc)
ovlap <- computeMsigOverlap(hgsc, thresh = 0.25)
ig <- computeMsigNetwork(ovlap, hgsc)
findMsigClusters(ig)</pre>
```

getMsigExclusionList Exclusion list of words for MSigDB gene set text mining

#### Description

List of words to discard when performing text mining MSigDB gene set names and short descriptions.

# Usage

```
getMsigExclusionList(custom = c())
```

#### Arguments

custom a character vector, containing list of words to add onto existing exclusion list.

#### Value

a character vector, containing words to be excluded from the text mining analysis.

#### Examples

getMsigExclusionList('remove')

hgsc

The Hallmark collection from the MSigDB

#### Description

The molecular signatures database (MSigDB) is a collection of over 25000 gene expression signatures. Signatures in v7.2 are divided into 9 categories. The Hallmarks collection contains gene expression signatures representing molecular processes that are hallmarks in cancer development and progression.

#### Usage

hgsc

# Format

A GeneSetCollection object with 50 GeneSet objects representing the 50 Hallmark gene expression signatures.

#### plotGeneStats

#### References

Subramanian, A., Tamayo, P., Mootha, V. K., Mukherjee, S., Ebert, B. L., Gillette, M. A., ... & Mesirov, J. P. (2005). Gene set enrichment analysis: a knowledge-based approach for interpreting genome-wide expression profiles. Proceedings of the National Academy of Sciences, 102(43), 15545-15550.

Liberzon, A., Subramanian, A., Pinchback, R., Thorvaldsdóttir, H., Tamayo, P., & Mesirov, J. P. (2011). Molecular signatures database (MSigDB) 3.0. Bioinformatics, 27(12), 1739-1740.

Liberzon, A., Birger, C., Thorvaldsdóttir, H., Ghandi, M., Mesirov, J. P., & Tamayo, P. (2015). The molecular signatures database hallmark gene set collection. Cell systems, 1(6), 417-425.

plotGeneStats

Plot gene statistics for clusters of gene sets

# Description

This function plots gene statistics against gene frequencies for any given cluster of gene sets. The plot can be used to identify genes that are over-represented in a cluster of gene-sets (identified based on gene-set overlaps) and have a strong statistic (e.g. log fold-chage or p-value).

#### Usage

```
plotGeneStats(
  geneStat,
  msigGsc,
  groups,
  statName = "Gene-level statistic",
  topN = 5
)
```

#### Arguments

geneStat	a named numeric, containing the statistic to be displayed. The vector must be named with either gene Symbols or Entrez IDs depending on annotations in msigGsc.
msigGsc	a GeneSetCollection object, containing gene sets from the MSigDB. The GSEABase::getBroadSets( function can be used to parse XML files downloaded from MSigDB.
groups	a named list, of character vectors or numeric indices specifying node groupings. Each element of the list represent a group and contains a character vector with node names.
statName	a character, specifying the name of the statistic.
topN	a numeric, specifying the number of genes to label. The top genes are those with the largest count and statistic.

# Value

a ggplot object, plotting the gene-level statistic against gene frequencies in the cluster of gene sets.

# Examples

```
library(GSEABase)
```

```
data(hgsc)
groups <- list('g1' = names(hgsc)[1:25], 'g2' = names(hgsc)[26:50])
#create statistics
allgenes = unique(unlist(geneIds(hgsc)))
gstats = rnorm(length(allgenes))
names(gstats) = allgenes
#plot
plotGeneStats(gstats, hgsc, groups)</pre>
```

plotMsigNetwork *Plot a gene set overlap network* 

# Description

Plots a network of gene set overlap with overlap computed using the computeMsigOverlap() and a graph created using computeMsigNetwork().

# Usage

```
plotMsigNetwork(
    ig,
    markGroups = NULL,
    genesetStat = NULL,
    nodeSF = 1,
    edgeSF = 1,
    lytFunc = "graphopt",
    lytParams = list(),
    rmUnmarkedGroups = FALSE,
    maxGrp = 12
)
```

# Arguments

ig	an igraph object, containing a network of gene set overlaps computed using computeMsigNetwork().
markGroups	a named list, of character vectors. Each element of the list represent a group and contains a character vector with node names. Up to 12 groups can be visualised in the plot.
genesetStat	a named numeric, statistic to project onto the nodes. These could be p-values, log fold-changes or gene set score from a singscore-based analysis.
nodeSF	a numeric, indicating the scaling factor to apply to node sizes.
edgeSF	a numeric, indicating the scaling factor to apply to edge widths.
lytFunc	a character, specifying the layout to use (see ggraph::create_layout()).
lytParams	a named list, containing additional parameters needed for the layout (see ggraph::create_layout()

10

#### plotMsigPPI

rmUnmarkedGrou	ps
	a logical, indicating whether unmarked groups should be removed from the net- work (TRUE) or retained (FALSE - default).
maxGrp	a numeric, specifying the maximum number of groups to plot.

# Value

a ggplot2 object

#### Examples

```
data(hgsc)
ovlap <- computeMsigOverlap(hgsc, thresh = 0.15)
ig <- computeMsigNetwork(ovlap, hgsc)
groups <- list(
   'g1' = c("HALLMARK_HYPOXIA", "HALLMARK_GLYCOLYSIS"),
   'g2' = c("HALLMARK_INTERFERON_GAMMA_RESPONSE")
)</pre>
```

```
plotMsigNetwork(ig, markGroups = groups)
```

plotMsigPPI

Plot PPI network for gene-set clusters identified using vissE

#### Description

This function plots the protein-protein interaction (PPI) network for a gene-set cluster identified using vissE. The international molecular exchange (IMEx) PPI is used to obtain PPIs for genes present in a gene-set cluster.

# Usage

```
plotMsigPPI(
  ppidf,
  msigGsc,
  groups,
  geneStat = NULL,
  statName = "Gene-level statistic",
  threshConfidence = 0,
  threshFrequency = 0.25,
  threshStatistic = 0,
  threshUseAbsolute = TRUE,
  topN = 5,
  nodeSF = 1,
  edgeSF = 1,
  lytFunc = "graphopt",
  lytParams = list()
)
```

# Arguments

ppidf	a data.frame, containing a protein-protein interaction from the IMEx database. This can be retrieved from the msigdb::getIMEX() function.
msigGsc	a GeneSetCollection object, containing gene sets from the MSigDB. The GSEABase::getBroadSets function can be used to parse XML files downloaded from MSigDB.
groups	a named list, of character vectors or numeric indices specifying node groupings. Each element of the list represent a group and contains a character vector with node names.
geneStat	a named numeric, containing the statistic to be displayed. The vector must be named with either gene Symbols or Entrez IDs depending on annotations in msigGsc.
statName	a character, specifying the name of the statistic.
threshConfidend	ce
	a numeric, specifying the confidence threshold to apply to determine high con- fidence interactions. This should be a value between 0 and 1 (default is 0).
threshFrequency	/
	a numeric, specifying the frequency threshold to apply to determine more fre- quent genes in the gene-set cluster. The frequecy of a gene is computed as the proportion of gene-sets to which the gene belongs. This should be a value be- tween 0 and 1 (default is 0.25).
threshStatistic	
	a numeric, specifying the threshold to apply to gene-level statistics (e.g. a log fold-change). This should be a value between 0 and 1 (default is 0).
threshUseAbsolu	ıte
	a logical, indicating whether the threshStatistic threshold should be applied to absolute values (default TRUE). This can be used to threshold on statistics such as the log fold-chage from a differential expression analysis.
topN	a numeric, specifying the number of genes to label. The top genes are those with the largest count and statistic.
nodeSF	a numeric, indicating the scaling factor to apply to node sizes.
edgeSF	a numeric, indicating the scaling factor to apply to edge widths.
lytFunc	a character, specifying the layout to use (see ggraph::create_layout()).
lytParams	a named list, containing additional parameters needed for the layout (see ggraph::create_layout()

# Value

a ggplot object with the protein-protein interaction networks plot for each gene-set cluster.

# Examples

```
data(hgsc)
grps = list('early' = 'HALLMARK_ESTROGEN_RESPONSE_EARLY', 'late' = 'HALLMARK_ESTROGEN_RESPONSE_LATE')
ppi = msigdb::getIMEX(org = 'hs', inferred = TRUE)
plotMsigPPI(ppi, hgsc, grps)
```

plotMsigWordcloud Compute and plot word frequencies for multiple MSigDB collections

# Description

Given a gene set collection, this function computes the word frequency of gene set names from the Molecular Signatures Database (MSigDB) collection (split by \_). Word frequencies are also computed using short descriptions attached with each gene set object.

# Usage

```
plotMsigWordcloud(
   msigGsc,
   groups,
   weight = NULL,
   measure = c("tfidf", "tf"),
   version = msigdb::getMsigdbVersions(),
   org = c("auto", "hs", "mm"),
   rmwords = getMsigExclusionList(),
   type = c("Name", "Short"),
   idf = NULL
)
```

# Arguments

msigGsc	a GeneSetCollection object, containing gene sets from the MSigDB. The GSEABase::getBroadSets function can be used to parse XML files downloaded from MSigDB.
groups	a named list, of character vectors or numeric indices specifying node groupings. Each element of the list represent a group and contains a character vector with node names.
weight	a named numeric vector, containing weights to apply to each gene-set. This can be -log10(FDR), -log10(p-value) or an enrichment score (ideally unsigned).
measure	a character, specifying how frequencies should be computed. "tf" uses term frequencies and "tfidf" (default) applies inverse document frequency weights to term frequencies.
version	a character, specifying the version of msigdb to use (see msigdb::getMsigdbVersions()).
org	a character, specifying the organism to use. This can either be "auto" (default), "hs" or "mm".
rmwords	a character vector, containing an exclusion list of words to discard from the analysis.
type	a character, specifying the source of text mining. Either gene set names (Name) or descriptions (Short) can be used.
idf	a list of named numeric vectors, specifying inverse document frequencies to use to penalise terms from gene-set names and short descriptions. This should be a vector of length 2 with names "Name" and "Short". Numeric vectors should contain weights and names should represent the term. Precomputed versions can be retrieved using the msigdb::getMsigdbIDF().

# Value

a ggplot object.

# Examples

```
data("hgsc")
groups <- list('g1' = names(hgsc)[1:25], 'g2' = names(hgsc)[26:50])
plotMsigWordcloud(hgsc, groups, rmwords = getMsigExclusionList())</pre>
```

14

# Index

\* datasets hgsc, 8 \* internal vissE-package, 2 bhuvad\_theme, 3 characteriseGeneset, 2, 3 computeMsigNetwork, 2, 4 computeMsigNetwork(), 7, 10 computeMsigOverlap, 2, 5 computeMsigOverlap(), 4, 10 computeMsigWordFreq, 6 findMsigClusters,7 getMsigExclusionList, 8 GSEABase::getBroadSets(), 6, 9, 12, 13 hgsc, 8 msigdb::getIMEX(), 12 msigdb::getMsigdbIDF(), 6, 13 msigdb::listCollections(),4 plotGeneStats, 2, 9 plotMsigNetwork, 2, 10 plotMsigPPI, 11 plotMsigWordcloud, 2, 13 vissE (vissE-package), 2 vissE-package, 2