Package 'NetPathMiner'

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Title NetPathMiner for Biological Network Construction, Path Mining and Visualization

Description NetPathMiner is a general framework for network path mining using genome-scale networks. It constructs networks from KGML, SBML and BioPAX files, providing three network representations, metabolic, reaction and gene representations. NetPathMiner finds active paths and applies machine learning methods to summarize found paths for easy interpretation. It also provides static and interactive visualizations of networks and paths to aid manual investigation.

Depends R (>= 3.0.2), igraph (>= 1.0)

Suggests rBiopaxParser (>= 2.1), RCurl, graph, knitr, rmarkdown, BiocStyle

VignetteBuilder knitr

License GPL (>= 2)

URL https://github.com/ahmohamed/NetPathMiner

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NetPathMiner-package General framework for network extraction, path mining.

Description

NetPathMiner implements a flexible module-based process flow for network path mining and visualization, which can be fully inte-grated with user-customized functions. NetPathMiner supports construction of various types of genome scale networks from KGML, SBML and BioPAX formats, enabling its utility to most common pathway databases. NetPathMiner also provides different visualization techniques to facilitate the analysis of even thousands of output paths.

Author(s)

Ahmed Mohamed <mohamed@kuicr.kyoto-u.ac.jp>

 $assign {\sf EdgeWeights}$

Assigning weights to network edges

Description

This function computes edge weights based on a gene expression profile.

Usage

```
assignEdgeWeights(
  microarray,
  graph,
  use.attr,
  y,
  weight.method = "cor",
  complex.method = "max",
  missing.method = "median",
  same.gene.penalty = "median",
  bootstrap = 100,
  verbose = TRUE
)
```

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Arguments

microarray Microarray should be a Dataframe or a matrix, with genes as rownames, and

samples as columns.

graph An annotated igraph object.

use.attr An attribute name to map microarray rows (genes) to graph vertices. The at-

tribute must be annotated in graph, and the values correspond to rownames of microarray. You can check the coverage and if there are complex vertices using

getAttrStatus. You can eliminate complexes using expandComplexes.

y Sample labels, given as a factor or a character vector. This must be the same

size as the columns of microarray

weight.method A function, or a string indicating the name of the function to be used to compute

the edge weights. The function is provided with 2 numerical verctors (2 rows from microarray), and it should return a single numerical value (or NA). The $\frac{1}{2}$

default computes Pearson's correlation.

complex.method A function, or a string indicating the name of the function to be used in weight-

ing edges connecting complexes. If a vertex has >1 attribute value, all possible pairwise weights are first computed, and given to complex.method. The default

function is max.

missing.method A function, or a string indicating the name of the function to be used in weight-

ing edges when one of the vertices lack expression data. The function is passed

all edge weights on the graph. Default is median.

same.gene.penalty

A numerical value to be assigned when 2 adjacent vertices have the same attribute value, since correlation and similarity measure will give perfect scores. Alternatively, same.gene.penalty can be a function, computing the penalty from all edge weights on the graph (excluding same-gene and missing values).

The default is to take the median

An integer n, where the weight. method is performed on n permutations of the

gene profiles, and taking the median value. Set it to NA to disable bootstrapping.

verbose Print the progress of the function.

Value

The input graph with edge.weight as an edge attribute. The attribute can be a list of weights if y labels were provided.

Author(s)

Ahmed Mohamed

```
## Convert a metabolic network to a reaction network.
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)</pre>
```

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biopax2igraph

Processes BioPAX objects into igraph objects

Description

This function takes BioPAX objects (level 2 or 3) as input, and returns either a metabolic or a signaling network as output.

Usage

```
biopax2igraph(
  biopax,
  parse.as = c("metabolic", "signaling"),
  expand.complexes = FALSE,
  inc.sm.molecules = FALSE,
  verbose = TRUE
)
```

Arguments

biopax BioPAX object generated by readBiopax.

parse.as Whether to process file into a metabolic or a signaling network.

expand.complexes

Split protein complexes into individual gene nodes. Ignored if parse.as="metabolic".

inc.sm.molecules

Include small molecules that are participating in signaling events. Ignored if

parse.as="metabolic".

verbose Whether to display the progress of the function.

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Details

This function requires rBiopaxParser installed.

Users can specify whether files are processes as metabolic or signaling networks.

Metabolic networks are given as bipartite graphs, where metabolites and reactions represent vertex types. Reactions are constructed from Conversion classes, connecting them to their corresponding Lefts and Rights. Each reaction vertex has genes attribute, listing all Catalysis relationships of this reaction. As a general rule, reactions inherit all annotation attributes of its catalyzig genes.

Signaling network have genes as vertices and edges represent interactions, such as activiation / inhibition. Genes participating in successive reactions are also connected. Signaling interactions are constructed from Control classes, where edges are drawn from controller to controlled.

All annotation attributes are exacted from XRefs associated with the vertices, and are stored according to MIRIAM guidelines (miraim.db, where db is the database name).

Value

An igraph object, representing a metbolic or a signaling network.

Author(s)

Ahmed Mohamed

See Also

Other Database extraction methods: KGML2igraph(), SBML2igraph()

Examples

```
if(requireNamespace("rBiopaxParser")){
   data(ex_biopax)
   # Process biopax as a metabolic network
   g <- biopax2igraph(ex_biopax)
   plotNetwork(g)

# Process SBML file as a signaling network
   g <- biopax2igraph(ex_biopax, parse.as="signaling", expand.complexes=TRUE)
}</pre>
```

colorVertexByAttr

Computes colors for vertices according to their attributes.

Description

This function returns a list of colors for vertices, assigned similar colors if they share a common attribute (ex: in the same pathway, etc).

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Usage

```
colorVertexByAttr(graph, attr.name, col.palette = palette())
```

Arguments

graph An annotated igraph object.

attr.name The attribute name (ex: "pathway") by which vertices will be colored. Complex attributes, where a vertex belongs to more than one group, are supported.

col.palette A color palette, or a palette generating function (ex:

col.palette=rainbow

).

Value

A list of colors (in HEX format) for vertices.

Author(s)

Ahmed Mohamed

See Also

```
Other Plotting methods: layoutVertexByAttr(), plotAllNetworks(), plotClassifierROC(), plotClusterMatrix(), plotCytoscapeGML(), plotNetwork(), plotPathClassifier(), plotPaths()
```

Examples

```
data("ex_kgml_sig")
v.colors <- colorVertexByAttr(ex_kgml_sig, "pathway")
plotNetwork(ex_kgml_sig, vertex.color=v.colors)</pre>
```

expandComplexes

Expand reactions / complexes into their gene constituents.

Description

These are general functions to expand vertices by their attributes, i.e. create a separate vertex for each attribute value.

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Usage

```
expandComplexes(
  graph,
  v.attr,
  keep.parent.attr = "^pathway",
  expansion.method = c("normal", "duplicate"),
 missing.method = c("keep", "remove", "reconnect")
)
makeGeneNetwork(
  graph,
  v.attr = "genes",
  keep.parent.attr = "^pathway",
  expansion.method = "duplicate",
 missing.method = "remove"
)
```

Arguments

graph An annotated igraph object.

Name of the attribute which vertices are expanded to. v.attr

keep.parent.attr

A (List of) regex experssions representing attributes to be inherited by daughter vertices. If "all" is passed, all parent attributes are inherited.

expansion.method

If "duplicate", attribute values sharing more than one parent vertex are duplicated for each vertex they participate in. For exmaple, if one gene G1 catalyzes reactions R1, R2; then G1##R1, and G1##R2 vertices are created. If "normal" only one vertex (G1) is created, and inherit all R1 and R2 connections and attributes.

missing.method How to deal with vertices with no attribute values. "keep" retains the parent node, "remove" simply deletes the vertex, and "reconnect" removes the vertex and connect its neighbours to each other (to prevent graph cuts).

Details

These functions can be very useful when merging networks constructed from different databases. For example, to match a network created from Reactome to a KEGG network, you can expand metabolite vertices by "miriam.kegg.compound" attribute.

Value

A new graph with vertices expanded.

makeGeneNetwork returns a graph, where nodes are genes, and edges represent participation in succesive reactions.

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Author(s)

Ahmed Mohamed

See Also

Other Network processing methods: makeMetaboliteNetwork(), makeReactionNetwork(), reindexNetwork(), rmSmallCompounds(), simplifyReactionNetwork(), vertexDeleteReconnect()

Examples

```
## Make a gene network from a reaction network.
data(ex_sbml) # A bipartite metbaolic network.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)</pre>
ggraph <- makeGeneNetwork(rgraph)</pre>
## Expand vertices into their contituent genes.
data(ex_kgml_sig) # Ras and chemokine signaling pathways in human
ggraph <- expandComplexes(ex_kgml_sig, v.attr = "miriam.ncbigene",</pre>
     keep.parent.attr= c("^pathway", "^compartment"))
## Create a separate vertex for each compartment. This is useful in duplicating
## metabolite vertices in a network.
## Not run:
graph <- expandComplexes(graph, v.attr = "compartment",</pre>
       keep.parent.attr = "all",
        expansion.method = "duplicate",
        missing.method = "keep")
## End(Not run)
```

extractPathNetwork

Creates a subnetwork from a ranked path list

Description

Creates a subnetwork from a ranked path list generated by pathRanker.

Usage

```
extractPathNetwork(paths, graph)
```

Arguments

paths The paths extracted by pathRanker.

graph A annotated igraph object.

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Value

A subnetwork from all paths provided. If paths are computed for several labels (sample categories), a subnetwork is returned for each label.

Author(s)

Ahmed Mohamed

See Also

Other Path ranking methods: getPathsAsEIDs(), pathRanker()

Examples

```
## Prepare a weighted reaction network.
## Conver a metabolic network to a reaction network.
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)</pre>
## Assign edge weights based on Affymetrix attributes and microarray dataset.
# Calculate Pearson's correlation.
data(ex_microarray) # Part of ALL dataset.
rgraph <- assignEdgeWeights(microarray = ex_microarray, graph = rgraph,</pre>
weight.method = "cor", use.attr="miriam.uniprot",
y=factor(colnames(ex_microarray)), bootstrap = FALSE)
## Get ranked paths using probabilistic shortest paths.
ranked.p <- pathRanker(rgraph, method="prob.shortest.path",</pre>
   K=20, minPathSize=6)
## Get the subnetwork of paths in reaction graph.
reaction.sub <- getPathsAsEIDs(ranked.p, rgraph)</pre>
## Get the subnetwork of paths in the original metabolic graph.
metabolic.sub <- getPathsAsEIDs(ranked.p, ex_sbml)</pre>
```

ex_biopax

Biopax example data

Description

A dataset containing Porphyrin metabolism pathway in Biopax Level 3 and parsed with readBiopax.

```
data(ex_biopax)
ex_biopax
```

ex_kgml_sig

ex_kgml_sig

Singaling network from KGML example

Description

An example igraph object representing Ras and chemokine signaling pathways in human extracted from KGML files.

Examples

```
data(ex_kgml_sig)
plotNetwork(ex_kgml_sig, vertex.color="pathway")
```

ex_microarray

An microarray data example.

Description

An microarray data example. This is part of the ALL dataset, for demonstration purposes.

Examples

```
data(ex_microarray)
```

ex_sbml

Metabolic network from SBML example

Description

An example igraph object representing bipartite metabolic network of Carbohydrate metabolism extracted from SBML file from Reactome database.

```
data(ex_sbml)
plotNetwork(ex_sbml, vertex.color="compartment.name")
```

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getAttrStatus	Get / Set vertex attribute names and coverage

Description

These functions report the annotation status of the vertices of a given network, modify or remove certain annotations.

Usage

```
getAttrStatus(graph, pattern = "^miriam.")
getAttrNames(graph, pattern = "")
getAttribute(graph, attr.name)
setAttribute(graph, attr.name, attr.value)
rmAttribute(graph, attr.name)
```

Arguments

graph An annotated igraph object.

pattern A regex experssion representing attribute name pattern.

attr.name The attribute name

attr.value A list of attribute values. This must be the same size as the number of vertices.

Details

NetPathMiner stores all its vertex annotation attributes in a list, and stores them collectively as a single attr. This is not to interfer with graph_attr_names from igraph package. All functions here target NetPathMiner annotations only.

Value

For getAttrStatus, a dataframe summarizing the number of vertices with no (missing), one (single) or more than one (complex) attribute value. The coverage

For getAttrNames, a character vector of attribute names matching the pattern.

For getAttribute, a list of vertex annotation values for the query attribute.

For setAttribute, a graph with the new attribute set.

For rmAttrNames, a new igraph object with the attibute removed.

Author(s)

Ahmed Mohamed

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See Also

Other Attribute handling methods: stdAttrNames()

Examples

```
data(ex_kgml_sig) # Ras and chemokine signaling pathways in human

# Get status of attribute "pathway" only
getAttrStatus(ex_kgml_sig, "^pathway$")

# Get status of all attributes starting with "pathway" and "miriam" keywords
getAttrStatus(ex_kgml_sig, "(^miriam)|(^pathway)")

# Get all attribute names containing "miriam"
getAttrNames(ex_kgml_sig, "miriam")
# Get all attribute names containing "miriam"
getAttribute(ex_kgml_sig, "miriam.ncbigene")

# Remove an attribute from graph
graph <- rmAttribute(ex_kgml_sig, "miriam.ncbigene")</pre>
```

getGeneSetNetworks

Generate geneset networks from an annotated network.

Description

This function generates geneset networks based on a given netowrk, by grouping vertices sharing common attributes (in the same pathway or compartment).

Usage

```
getGeneSetNetworks(
  graph,
  use.attr = "pathway",
  format = c("list", "pathway-class")
```

Arguments

graph An annotated igraph object..

use.attr The attribute by which vertices are grouped (tepically pathway, or GO)

format The output format. If "list" is specified, a list of subgraphs are returned (de-

fault). If "pathway-class" is specified, a list of pathway-class objects are returned. Pathway-class is used by graphite package to run several methods of

topology-based enrichment analyses.

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Value

A list of geneset networks as igraph or Pathway-class objects.

Author(s)

Ahmed Mohamed

See Also

```
getGeneSets
```

Examples

```
data(ex_kgml_sig) # Ras and chemokine signaling pathways in human
genesetnets <- getGeneSetNetworks(ex_kgml_sig, use.attr="pathway")</pre>
# Integration with graphite package
## Not run:
if(requireNamespace("graphite") & requireNamespace("clipper") & requireNamespace("ALL")){
genesetnets <- getGeneSetNetworks(ex_kgml_sig,</pre>
     use.attr="pathway", format="pathway-class")
path <- convertIdentifiers(genesetnets$`Chemokine signaling pathway`,</pre>
      "entrez")
genes <- nodes(path)</pre>
data(ALL)
all <- as.matrix(exprs(ALL[1:length(genes),1:20]))</pre>
classes <- c(rep(1,10), rep(2,10))
rownames(all) <- genes</pre>
runClipper(path, all, classes, "mean", pathThr=0.1)
## End(Not run)
```

getGeneSets

Generate genesets from an annotated network.

Description

This function generates genesets based on a given netowrk, by grouping vertices sharing common attributes (in the same pathway or compartment). Genes associated with each vertex can be specified through gene.attr argument.

Usage

```
getGeneSets(graph, use.attr = "pathway", gene.attr = "genes", gmt.file)
```

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Arguments

graph	An annotated igraph object
use.attr	The attribute by which vertices are grouped (tepically pathway, or GO)
gene.attr	The attribute listing genes annotated with each vertex (ex: miriam.ncbigene, miriam.uniprot,)
gmt.file	Optinal. If provided, Results are exported to a GMT file. GMT files are readily used by most gene set analysis packages.

Value

A list of genesets or written to gmt file if provided.

Author(s)

Ahmed Mohamed

See Also

```
getGeneSetNetworks
```

Examples

```
data(ex_kgml_sig) # Ras and chemokine signaling pathways in human
genesets <- getGeneSets(ex_kgml_sig, use.attr="pathway", gene.attr="miriam.ncbigene")

# Write the genesets in a GMT file, and read it using GSEABase package.
getGeneSets(ex_kgml_sig, use.attr="pathway", gene.attr="miriam.ncbigene", gmt.file="kgml.gmt")

## Not run:
if(requireNamespace("GSEABase"))
toGmt("kgml.gmt")

## End(Not run)

# Create genesets using compartment information
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
genesets <- getGeneSets(ex_sbml, use.attr="compartment.name", gene.attr="miriam.uniprot")</pre>
```

getPathsAsEIDs

Convert a ranked path list to edge ids of a graph

Description

Convert a ranked path list to Edge ids of a graph, where paths can come from a different representation (for example matching path from a reaction network to edges on a metabolic network).

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Usage

```
getPathsAsEIDs(paths, graph)
```

Arguments

paths The paths extracted by pathRanker.

graph A annotated igraph object.

Value

A list of edge ids on the provided graph.

Author(s)

Ahmed Mohamed

See Also

Other Path ranking methods: extractPathNetwork(), pathRanker()

```
## Prepare a weighted reaction network.
## Conver a metabolic network to a reaction network.
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)</pre>
## Assign edge weights based on Affymetrix attributes and microarray dataset.
# Calculate Pearson's correlation.
data(ex_microarray) # Part of ALL dataset.
rgraph <- assignEdgeWeights(microarray = ex_microarray, graph = rgraph,</pre>
weight.method = "cor", use.attr="miriam.uniprot",
y=factor(colnames(ex_microarray)), bootstrap = FALSE)
## Get ranked paths using probabilistic shortest paths.
ranked.p <- pathRanker(rgraph, method="prob.shortest.path",</pre>
    K=20, minPathSize=6)
## Get the edge ids along paths in the reaction graph.
path.eids <- getPathsAsEIDs(ranked.p, rgraph)</pre>
## Get the edge ids along paths in the original metabolic graph.
path.eids <- getPathsAsEIDs(ranked.p, ex_sbml)</pre>
```

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KGML2igraph

Processes KGML files into igraph objects

Description

This function takes KGML files as input, and returns either a metabolic or a signaling network as output.

Usage

```
KGML2igraph(
  filename,
  parse.as = c("metabolic", "signaling"),
  expand.complexes = FALSE,
  verbose = TRUE
)
```

Arguments

filename A character vector containing the KGML files to be processed. If a directory

path is provided, all *.xml files in it and its subdirectories are included.

parse. as Whether to process file into a metabolic or a signaling network.

expand.complexes

Split protein complexes into individual gene nodes. This argument is ignored if

parse.as="metabolic"

verbose Whether to display the progress of the function.

Details

Users can specify whether files are processes as metabolic or signaling networks.

Metabolic networks are given as bipartite graphs, where metabolites and reactions represent vertex types. This is constructed from <reaction> xml node in KGML file, connecting them to their corresponding substrates and products. Each reaction vertex has genes attribute, listing all genes associated with the reaction. As a general rule, reactions inherit all annotation attributes of its catalyzig genes.

Signaling network have genes as vertices and edges represent interactions, such as activiation / inhibition. Genes participating in successive reactions are also connected. Signaling parsing method processes <ECrel>, <PPrel> and <PCrel> interactions from KGML files.

To generate a genome scale network, simply provide a list of files to be parsed, or put all file in a directory, as pass the directory path as filename

Value

An igraph object, representing a metbolic or a signaling network.

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Author(s)

Ahmed Mohamed

See Also

Other Database extraction methods: SBML2igraph(), biopax2igraph()

Examples

```
if(is.loaded("readkgmlfile")){ # This is false if libxml2 wasn't available at installation.
    filename <- system.file("extdata", "hsa00860.xml", package="NetPathMiner")

# Process KGML file as a metabolic network
g <- KGML2igraph(filename)
plotNetwork(g)

# Process KGML file as a signaling network
g <- KGML2igraph(filename, parse.as="signaling", expand.complexes=TRUE)
plotNetwork(g)
}</pre>
```

layoutVertexByAttr

A graph layout function, which groups vertices by attribute.

Description

This function generates a layout for igraph objects, keeping vertices with the same attribute (ex: in the same pathway, etc) close to each other.

Usage

```
layoutVertexByAttr(
  graph,
  attr.name,
  cluster.strength = 1,
  layout = layout.auto
)
```

Arguments

graph An annotated igraph object.

attr.name The attribute name by which vertices are laid out.

cluster.strength

A number indicating tie strengths between vertices with the same attribute. The

larger it is, the closer the vertices will be.

layout A layout function, ideally a force-directed layout fuction, such as layout_with_fr

and layout_with_kk.

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Value

A two-column matrix indicating the x and y postions of vertices.

Author(s)

Ahmed Mohamed

See Also

```
Other Plotting methods: colorVertexByAttr(), plotAllNetworks(), plotClassifierROC(), plotClusterMatrix(), plotCytoscapeGML(), plotNetwork(), plotPathClassifier(), plotPaths()
```

Examples

```
data("ex_kgml_sig")
v.layout <- layoutVertexByAttr(ex_kgml_sig, "pathway")
plotNetwork(ex_kgml_sig, vertex.color="pathway", layout=v.layout)
v.layout <- layoutVertexByAttr(ex_kgml_sig, "pathway", cluster.strength=5)
plotNetwork(ex_kgml_sig, vertex.color="pathway", layout=v.layout)</pre>
```

makeMetaboliteNetwork Convert metabolic network to metabolite network.

Description

This function removes reaction nodes keeping them as edge attributes. The resulting network contains metabolite nodes only, where edges indicate that reaction conversions.

Usage

```
makeMetaboliteNetwork(graph)
```

Arguments

graph

A metabolic network.

Value

A reaction network.

Author(s)

Ahmed Mohamed

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See Also

Other Network processing methods: expandComplexes(), makeReactionNetwork(), reindexNetwork(), rmSmallCompounds(), simplifyReactionNetwork(), vertexDeleteReconnect()

Examples

```
## Conver a metabolic network to a metbolite network.
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
mgraph <- makeMetaboliteNetwork(ex_sbml)</pre>
```

makeReactionNetwork

Convert metabolic network to reaction network.

Description

This function removes metabolite nodes keeping them as edge attributes. The resulting network contains reaction nodes only, where edges indicate that a metabolite produced by one reaction is consumed by the other.

Usage

```
makeReactionNetwork(graph, simplify = FALSE)
```

Arguments

graph A metabolic network.

simplify An option to remove translocation and spontaneous reactions that require no cat-

alyzing genes. Translocation reactions are detected from reaction name (SBML,

BioPAX), or by having identical substrates and products.

Value

A reaction network.

Author(s)

Ahmed Mohamed

See Also

Other Network processing methods: expandComplexes(), makeMetaboliteNetwork(), reindexNetwork(), rmSmallCompounds(), simplifyReactionNetwork(), vertexDeleteReconnect()

```
## Conver a metabolic network to a reaction network.
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)</pre>
```

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NPMdefaults

Default values for NetPathMiner

Description

This function gets a NetPathMiner default value for a variable.

Usage

```
NPMdefaults(value)
```

Arguments

value

a character string indicating the variable name.

Details

NetPathMiner defines the following defaults:

- small.comp.ls Dataframe of ubiquitous metabolites. Used by rmSmallCompounds.
- bridge Dataframe of attributes supported by Brigde Database. Used by fetchAttribute.
- bridge.organisms A list of bridge supported organisms. Used by fetchAttribute.
- bridge.web The base URL for Brigde Database webservices. Used by fetchAttribute.

Value

The defuult value for the given variable.

Author(s)

Ahmed Mohamed

```
\# Get the default list of small compounds (uniquitous metabolites). NPMdefaults("small.comp.ls")
```

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pathClassifier

HME3M Markov pathway classifier.

Description

HME3M Markov pathway classifier.

Usage

```
pathClassifier(
  paths,
  target.class,
  M,
  alpha = 1,
  lambda = 2,
  hme3miter = 100,
  plriter = 1,
  init = "random"
)
```

Arguments

paths The training paths computed by pathsToBinary target.class he label of the targe class to be classified. This label must be present as a label within the paths\\$y object М Number of components within the paths to be extracted. alpha The PLR learning rate. (between 0 and 1). lambda The PLR regularization parameter. (between 0 and 2) hme3miter Maximum number of HME3M iterations. It will stop when likelihood change is < 0.001. Maximum number of PLR iteractions. It will stop when likelihood change is < plriter 0.001. Specify whether to initialize the HME3M responsibilities with the 3M model init

Details

Take care with selection of lambda and alpha - make sure you check that the likelihood is always increasing.

Value

A list with the following elements. A list with the following values

random is recommended.

h A dataframe with the EM responsibilities.

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theta A dataframe with the Markov parameters for each component.

A dataframe with the PLR coefficients for each component.

proportions The probability of each HME3M component.

posterior.probs

The HME3M posterior probability.

likelihood The likelihood convergence history.

plrplr The posterior predictions from each components PLR model.

path.probabilities

The 3M probabilities for each path belonging to each component.

params The parameters used to build the model.

y The binary response variable used by HME3M. A 1 indicates the location of the

target.class labels in paths\\$y

perf The training set ROC curve AUC.

label The HME3M predicted label for each path.

component The HME3M component assignment for each path.

Author(s)

Timothy Hancock and Ichigaku Takigawa

References

Hancock, Timothy, and Mamitsuka, Hiroshi: A Markov Classification Model for Metabolic Pathways, Workshop on Algorithms in Bioinformatics (WABI), 2009

Hancock, Timothy, and Mamitsuka, Hiroshi: A Markov Classification Model for Metabolic Pathways, Algorithms for Molecular Biology 2010

See Also

Other Path clustering & classification methods: pathCluster(), pathsToBinary(), plotClassifierROC(), plotClusterMatrix(), plotPathClassifier(), plotPathCluster(), predictPathClassifier(), predictPathCluster()

```
## Prepare a weighted reaction network.
## Conver a metabolic network to a reaction network.
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)

## Assign edge weights based on Affymetrix attributes and microarray dataset.
# Calculate Pearson's correlation.
data(ex_microarray) # Part of ALL dataset.
rgraph <- assignEdgeWeights(microarray = ex_microarray, graph = rgraph, weight.method = "cor", use.attr="miriam.uniprot",
    y=factor(colnames(ex_microarray)), bootstrap = FALSE)</pre>
```

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```
## Get ranked paths using probabilistic shortest paths.
ranked.p <- pathRanker(rgraph, method="prob.shortest.path",
        K=20, minPathSize=6)

## Convert paths to binary matrix.
ybinpaths <- pathsToBinary(ranked.p)
p.class <- pathClassifier(ybinpaths, target.class = "BCR/ABL", M = 3)

## Contingency table of classification performance
table(ybinpaths$y,p.class$label)

## Plotting the classifier results.
plotClassifierROC(p.class)
plotClusters(ybinpaths, p.class)</pre>
```

pathCluster

3M Markov mixture model for clustering pathways

Description

3M Markov mixture model for clustering pathways

Usage

```
pathCluster(ybinpaths, M, iter = 1000)
```

Arguments

ybinpaths The training paths computed by pathsToBinary.

M The number of clusters.

iter The maximum number of EM iterations.

Value

A list with the following items:

h The posterior probabilities that each path belongs to each cluster.

labels The cluster membership labels.

theta The probabilities of each gene for each cluster.

proportions The mixing proportions of each path.

likelihood The likelihood convergence history.

params The specific parameters used.

Author(s)

Ichigaku Takigawa

Timothy Hancock

pathRanker 25

References

Mamitsuka, H., Okuno, Y., and Yamaguchi, A. 2003. Mining biologically active patterns in metabolic pathways using microarray expression profiles. SIGKDD Explor. News 1. 5, 2 (Dec. 2003), 113-121.

See Also

Other Path clustering & classification methods: pathClassifier(), pathsToBinary(), plotClassifierROC(), plotClusterMatrix(), plotPathClassifier(), plotPathCluster(), predictPathClassifier(), predictPathCluster()

Examples

```
## Prepare a weighted reaction network.
## Conver a metabolic network to a reaction network.
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)</pre>
## Assign edge weights based on Affymetrix attributes and microarray dataset.
# Calculate Pearson's correlation.
data(ex_microarray) # Part of ALL dataset.
rgraph <- assignEdgeWeights(microarray = ex_microarray, graph = rgraph,</pre>
weight.method = "cor", use.attr="miriam.uniprot", bootstrap = FALSE)
## Get ranked paths using probabilistic shortest paths.
ranked.p <- pathRanker(rgraph, method="prob.shortest.path",</pre>
    K=20, minPathSize=8)
## Convert paths to binary matrix.
ybinpaths <- pathsToBinary(ranked.p)</pre>
p.cluster <- pathCluster(ybinpaths, M=2)</pre>
plotClusters(ybinpaths, p.cluster)
```

pathRanker

Extracting and ranking paths from a network

Description

Given a weighted igraph object, path ranking finds a set of node/edge sequences (paths) to maximize the sum of edge weights. pathRanker(method="prob.shortest.path") extracts the K most probable paths within a weighted network. pathRanker(method="pvalue") extracts a list of paths whose sum of edge weights are significantly higher than random paths of the same length.

Usage

```
pathRanker(
  graph,
```

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```
method = "prob.shortest.path",
  start,
  end,
  verbose = TRUE,
  ...
)
```

Arguments

graph A weighted igraph object. Weights must be in edge.weights or weight edge

attributes.

method Which path ranking method to use.

start A list of start vertices, given by their vertex id.

A list of terminal vertices, given by their vertex id.

Whether to display the progress of the function.

Method-specific parameters. See Details section.

Details

The input here is graph. A weight must be assigned to each edge. Bootstrapped Pearson correlation edge weights can be assigned to each edge by assignEdgeWeights. However the specification of the edge weight is flexible with the condition that increasing values indicate stronger relationships between vertices.

Probabilistic Shortest Paths: pathRanker (method="prob.shortest.path") finds the K most probable loopless paths given a weighted network. Before the paths are ranked the edge weights are converted into probabilistic edge weights using the Empirical Cumulative Distribution (ECDF) over all edge weights. This is called ECDF edge weight. The ECDF edge weight serves as a probabilistic rank of the most important gene-gene interactions. The probabilistic nature of the ECDF edge weights allow for a significance test to determine if a path contains any functional structure or is simply a random walk. The probability of a path is simily the product of all ECDF weights along the path. This is computed as a sum of the logs of the ECDF edge weights.

The follwing arguments can be passed to pathRanker(method="prob.shortest.path"):

K Maximum number of paths to extract. Defaults to 10.

minPathSize The minimum number of edges for each extracted path. Defualts to 1.

normalize Specify if you want to normalize the probabilistic edge weights (across different labels) before extracting the paths. Defaults to TRUE.

P-value method: pathRanker(method="pvalue") is deprecated. Please use prob. shortest.path instead.

Value

A list of paths where each path has the following items:

gene The ordered sequence of genes visited along the path.

compounds The ordered sequence of compounds visited along the path.

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weights The ordered sequence of the log(ECDF edge weights) along the path.

distance The sum of the log(ECDF edge weights) along each path. (a sum of logs is a

product)

Author(s)

Timothy Hancock, Ichigaku Takigawa, Nicolas Wicker and Ahmed Mohamed

See Also

```
getPathsAsEIDs, extractPathNetwork
Other Path ranking methods: extractPathNetwork(), getPathsAsEIDs()
```

Examples

```
## Prepare a weighted reaction network.
## Conver a metabolic network to a reaction network.
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)

## Assign edge weights based on Affymetrix attributes and microarray dataset.
# Calculate Pearson's correlation.
data(ex_microarray) # Part of ALL dataset.
rgraph <- assignEdgeWeights(microarray = ex_microarray, graph = rgraph, weight.method = "cor", use.attr="miriam.uniprot",
    y=factor(colnames(ex_microarray)), bootstrap = FALSE)

## Get ranked paths using probabilistic shortest paths.
ranked.p <- pathRanker(rgraph, method="prob.shortest.path",
    K=20, minPathSize=6)</pre>
```

pathsToBinary

Converts the result from pathRanker into something suitable for path-Classifier or pathCluster.

Description

Converts the result from pathRanker into something suitable for pathClassifier or pathCluster.

Usage

```
pathsToBinary(ypaths)
```

Arguments

ypaths The result of pathRanker.

28 paths ToBinary

Details

Converts a set of pathways from pathRanker into a list of binary pathway matrices. If the pathways are grouped by a response label then the *pathsToBinary* returns a list labeled by response class where each element is the binary pathway matrix for each class. If the pathways are from pathRanker then a list wiht a single element containing the binary pathway matrix is returned. To look up the structure of a specific binary path in the corresponding ypaths object simply use matrix index by calling ypaths[[ybinpaths\\$pidx[i,]]], where i is the row in the binary paths object you wish to reference.

Value

A list with the following elements.

paths All paths within ypaths converted to a binary string and concatenated into the

one matrix.

y The response variable.

pidx An matrix where each row specifies the location of that path within the ypaths

object.

Author(s)

Timothy Hancock and Ichigaku Takigawa

See Also

```
Other Path clustering & classification methods: pathClassifier(), pathCluster(), plotClassifierROC(), plotClusterMatrix(), plotPathClassifier(), plotPathCluster(), predictPathClassifier(), predictPathCluster()
```

```
## Prepare a weighted reaction network.
## Conver a metabolic network to a reaction network.
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)</pre>
## Assign edge weights based on Affymetrix attributes and microarray dataset.
# Calculate Pearson's correlation.
data(ex_microarray) # Part of ALL dataset.
rgraph <- assignEdgeWeights(microarray = ex_microarray, graph = rgraph,</pre>
weight.method = "cor", use.attr="miriam.uniprot",
y=factor(colnames(ex_microarray)), bootstrap = FALSE)
## Get ranked paths using probabilistic shortest paths.
ranked.p <- pathRanker(rgraph, method="prob.shortest.path",</pre>
    K=20, minPathSize=6)
## Convert paths to binary matrix.
ybinpaths <- pathsToBinary(ranked.p)</pre>
p.cluster <- pathCluster(ybinpaths, M=3)</pre>
```

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```
plotClusters(ybinpaths, p.cluster, col=c("red", "green", "blue") )
```

plotAllNetworks

Higlighting ranked paths over multiple network representations.

Description

This function highlighting ranked paths over different network representations, metabolic, reaction and gene networks. The functions finds equivalent paths across different networks and marks them.

Usage

```
plotAllNetworks(
  paths,
  metabolic.net = NULL,
  reaction.net = NULL,
  gene.net = NULL,
  path.clusters = NULL,
  plot.clusters = TRUE,
  col.palette = palette(),
  layout = layout.auto,
  ...
)
```

Arguments

```
The result of pathRanker.
paths
metabolic.net
                 A bipartite metabolic network.
                 A reaction network, resulting from makeReactionNetwork.
reaction.net
                 A gene network, resulting from makeGeneNetwork.
gene.net
path.clusters
                 The result from pathCluster or pathClassifier.
plot.clusters
                  Whether to plot clustering information, as generated by plotClusters
col.palette
                 A color palette, or a palette generating function (ex:
                 col.palette=rainbow
layout
                 Either a graph layout function, or a two-column matrix specifiying vertex coor-
                  dinates.
                 Additional arguments passed to plotNetwork.
```

Value

Highlights the path list over all provided networks.

30 plotClassifierROC

Author(s)

Ahmed Mohamed

See Also

Other Plotting methods: colorVertexByAttr(), layoutVertexByAttr(), plotClassifierROC(), plotClusterMatrix(), plotCytoscapeGML(), plotNetwork(), plotPathClassifier(), plotPaths()

Examples

plotClassifierROC

Diagnostic plots for pathClassifier.

Description

Diagnostic plots for pathClassifier.

Usage

```
plotClassifierROC(mix)
```

Arguments

mix

The result from pathClassifier.

plotClusterMatrix 31

Value

Diagnostic plots of the result from pathClassifier. itemTopROC curves for the posterior probabilities (mix\\$posterior.probs) and for each HME3M component (mix\\$h). This gives information about what response label each relates to. A ROC curve with an AUC < 0.5 relates to y = 0. Conversely ROC curves with AUC > 0.5 relate to y = 1. itemBottomThe likelihood convergence history for the HME3M model. If the parameters alpha or lambda are set too large then the likelihood may decrease.

Author(s)

Timothy Hancock and Ichigaku Takigawa

See Also

```
Other Path clustering & classification methods: pathClassifier(), pathCluster(), pathSToBinary(), plotClusterMatrix(), plotPathClassifier(), plotPathCluster(), predictPathClassifier(), predictPathCluster()
```

Other Plotting methods: colorVertexByAttr(), layoutVertexByAttr(), plotAllNetworks(), plotClusterMatrix(), plotCytoscapeGML(), plotNetwork(), plotPathClassifier(), plotPaths()

plotClusterMatrix

Plots the structure of all path clusters

Description

Plots the structure of all path clusters

Usage

```
plotClusterMatrix(
  ybinpaths,
  clusters,
  col = rainbow(clusters$params$M),
  grid = TRUE
)

plotClusterProbs(clusters, col = rainbow(clusters$params$M))

plotClusters(ybinpaths, clusters, col, ...)
```

Arguments

ybinpaths The training paths computed by pathsToBinary.

clusters The pathway cluster model trained by pathCluster or pathClassifier.

col Colors for each path cluster.

grid A logical, whether to add a grid to the plot
... Extra paramaters passed to plotClusterMatrix

32 plotClusterMatrix

Value

plotClusterMatrix plots an image of all paths the training dataset. Rows are the paths and columns are the genes (features) included within each path. Paths are colored according to cluster membership.

plotClusterProbs The training set posterior probabilities for each path belonging to a 3M component.

plotClusters: combines the two plots produced by plotClusterProbs and plotClusterMatrix.

Author(s)

Ahmed Mohamed

See Also

Other Path clustering & classification methods: pathClassifier(), pathCluster(), pathSToBinary(), plotClassifierROC(), plotPathClassifier(), plotPathCluster(), predictPathClassifier(), predictPathCluster()

Other Plotting methods: colorVertexByAttr(), layoutVertexByAttr(), plotAllNetworks(), plotClassifierROC(), plotCytoscapeGML(), plotNetwork(), plotPathClassifier(), plotPaths()

```
## Prepare a weighted reaction network.
## Conver a metabolic network to a reaction network.
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)</pre>
## Assign edge weights based on Affymetrix attributes and microarray dataset.
# Calculate Pearson's correlation.
data(ex_microarray) # Part of ALL dataset.
rgraph <- assignEdgeWeights(microarray = ex_microarray, graph = rgraph,</pre>
weight.method = "cor", use.attr="miriam.uniprot",
y=factor(colnames(ex_microarray)), bootstrap = FALSE)
## Get ranked paths using probabilistic shortest paths.
ranked.p <- pathRanker(rgraph, method="prob.shortest.path",</pre>
    K=20, minPathSize=8)
## Convert paths to binary matrix.
ybinpaths <- pathsToBinary(ranked.p)</pre>
p.cluster <- pathCluster(ybinpaths, M=2)</pre>
plotClusters(ybinpaths, p.cluster, col=c("red", "blue") )
```

plotCytoscapeGML 33

plotCytoscapeGML

Plots an annotated igraph object in Cytoscape.

Description

plotCytoscape function has been removed because RCytoscape is no longer prensent in Bioconductor. Future plans will use RCy3 for Cytoscape plotting, once RCy3 is supported on MacOS and Windows. plotCytoscapeGML exports the network plot in GML format, that can be later imported into Cytoscape (using "import network from file" option). This fuction is compatible with all Cytoscape versions.

Usage

```
plotCytoscapeGML(
   graph,
   file,
   layout = layout.auto,
   vertex.size,
   vertex.label,
   vertex.shape,
   vertex.color,
   edge.color
)
```

Arguments

An annotated igraph object. graph file Output GML file name to which the network plot is exported. layout Either a graph layout function, or a two-column matrix specifiying vertex coordinates. vertex.size Vertex size. If missing, the vertex attribute "size" (V(g)\$size)) will be used. vertex.label Vertex labels. If missing, the vertex attribute "label" (V(g)\$label)) will be used. If missing, vertices are labeled by their name. vertex.shape Vertex shape in one of igraph shapes. If missing, the vertex attribute "shape" (V(g)\$shape)) will be used. Shapes are converted from igraph convention to Cytoscape convention. "square", "rectangle" and "vrectangle" are converted to "RECT", "csquare" and "crectangle" are converted to "ROUND_RECT", all other shapes are considered "ELLIPSE"

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```
A color or a list of colors for vertices. Vetices with multiple colors are not supported. If missing, the vertex attribute "color" (

V(g)$color)
) will be used.

A color or a list of colors for edges. If missing, the edge attribute "color" (

E(g)$color)
) will be used.
```

Value

For plotCytoscapeGML, results are written to file.

Author(s)

Ahmed Mohamed

See Also

```
Other Plotting methods: colorVertexByAttr(), layoutVertexByAttr(), plotAllNetworks(), plotClassifierROC(), plotClusterMatrix(), plotNetwork(), plotPathClassifier(), plotPaths()
```

Examples

```
data("ex_sbml")
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)
v.layout <- layoutVertexByAttr(rgraph, "compartment")
v.color <- colorVertexByAttr(rgraph, "compartment")

# Export network plot to GML file
plotCytoscapeGML(rgraph, file="example.gml", layout=v.layout,
    vertex.color=v.color, vertex.size=10)</pre>
```

plotNetwork

Plots an annotated igraph object.

Description

This function is a wrapper function for plot.igraph, with 2 main additions. 1. Add the ability to color vertices by their attributes (see examples), accompanied by an inofrmative legend. 2. Resize vertex.size, edge.arrow.size, label.cex according to the plot size and the size of the network.

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Usage

```
plotNetwork(
   graph,
   vertex.color,
   col.palette = palette(),
   layout = layout.auto,
   legend = TRUE,
   ...
)
```

Arguments

graph An annotated igraph object.

vertex.color A list of colors for vertices, or an attribute names (ex: "pathway") by which vertices will be colored. Complex attributes, where a vertex belongs to more than one group, are supported. This can also be the output of colorVertexByAttr.

col.palette A color palette, or a palette generating function (ex: col.palette=rainbow).

layout Either a graph layout function, or a two-column matrix specifiying vertex coordinates.

legend Wheter to plot a legend. The legend is only plotted if vertices are colored by attribute values.

... Additional arguments passed to plot.igraph.

Value

Produces a plot of the network.

Author(s)

Ahmed Mohamed

See Also

```
Other Plotting methods: colorVertexByAttr(), layoutVertexByAttr(), plotAllNetworks(), plotClassifierROC(), plotClusterMatrix(), plotCytoscapeGML(), plotPathClassifier(), plotPaths()
```

36 plotPathClassifier

plotPathClassifier	Plots the structure of specified path found by pathClassifier.

Description

Plots the structure of specified path found by pathClassifier.

Usage

```
plotPathClassifier(ybinpaths, obj, m, tol = NULL)
```

Arguments

The training paths computed by pathsToBinary ybinpaths

The pathClassifier pathClassifier. obj

The path component to view.

tol A tolerance for 3M parameter theta which is the probability for each edge

within each cluster. If the tolerance is set all edges with a theta below that

tolerance will be removed from the plot.

Value

Produces a plot of the paths with the path probabilities and prediction probabilities and ROC curve overlayed.

Center Plot An image of all paths the training dataset. Rows are the paths and columns

> are the genes (vertices) included within each pathway. A colour within image indicates if a particular gene (vertex) is included within a specific path. Colours

flag whether a path belongs to the current HME3M component (P > 0.5).

Center Right The training set posterior probabilities for each path belonging to the current 3M

component.

Center Top The ROC curve for this HME3M component.

Top Bar Plots Theta: The 3M component probabilities - indicates the importance of each edge

is to a path. Beta: The PLR coefficient - the magnitude indicates the importance

of the edge to the classify the response.

Author(s)

Timothy Hancock and Ichigaku Takigawa

See Also

Other Path clustering & classification methods: pathClassifier(), pathCluster(), pathsToBinary(), plotClassifierROC(), plotClusterMatrix(), plotPathCluster(), predictPathClassifier(), predictPathCluster()

Other Plotting methods: colorVertexByAttr(), layoutVertexByAttr(), plotAllNetworks(), plotClassifierROC(), plotClusterMatrix(), plotCytoscapeGML(), plotNetwork(), plotPaths() plotPathCluster 37

Examples

```
## Prepare a weighted reaction network.
## Conver a metabolic network to a reaction network.
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)</pre>
## Assign edge weights based on Affymetrix attributes and microarray dataset.
# Calculate Pearson's correlation.
data(ex_microarray) # Part of ALL dataset.
rgraph <- assignEdgeWeights(microarray = ex_microarray, graph = rgraph,</pre>
weight.method = "cor", use.attr="miriam.uniprot",
y=factor(colnames(ex_microarray)), bootstrap = FALSE)
## Get ranked paths using probabilistic shortest paths.
ranked.p <- pathRanker(rgraph, method="prob.shortest.path",</pre>
    K=20, minPathSize=6)
## Convert paths to binary matrix.
ybinpaths <- pathsToBinary(ranked.p)</pre>
p.class <- pathClassifier(ybinpaths, target.class = "BCR/ABL", M = 3)</pre>
## Plotting the classifier results.
plotClassifierROC(p.class)
plotClusters(ybinpaths, p.class)
```

plotPathCluster

Plots the structure of specified path cluster

Description

Plots the structure of specified path found by pathCluster.

Usage

```
plotPathCluster(ybinpaths, clusters, m, tol = NULL)
```

Arguments

ybinpaths The training paths computed by pathsToBinary.

clusters The pathway cluster model trained by pathCluster or pathClassifier.

m The path cluster to view.

tol A tolerance for 3M parameter theta which is the probability for each edge

within each cluster. If the tolerance is set all edges with a theta below that

tolerance will be removed from the plot.

38 plotPathCluster

Value

Produces a plot of the paths with the path probabilities and cluster membership probabilities.

Center Plot An image of all paths the training dataset. Rows are the paths and columns are

the genes (features) included within each path.

Right The training set posterior probabilities for each path belonging to the current 3M

component.

Top Bar Plots Theta, The 3M component probabilities - indicates the importance of each edge

to a pathway.

Author(s)

Timothy Hancock and Ichigaku Takigawa

See Also

Other Path clustering & classification methods: pathClassifier(), pathCluster(), pathsToBinary(), plotClassifierROC(), plotClusterMatrix(), plotPathClassifier(), predictPathClassifier(), predictPathCluster()

```
## Prepare a weighted reaction network.
## Conver a metabolic network to a reaction network.
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)</pre>
## Assign edge weights based on Affymetrix attributes and microarray dataset.
# Calculate Pearson's correlation.
data(ex_microarray) # Part of ALL dataset.
rgraph <- assignEdgeWeights(microarray = ex_microarray, graph = rgraph,</pre>
weight.method = "cor", use.attr="miriam.uniprot", bootstrap = FALSE)
## Get ranked paths using probabilistic shortest paths.
ranked.p <- pathRanker(rgraph, method="prob.shortest.path",</pre>
   K=20, minPathSize=8)
## Convert paths to binary matrix.
ybinpaths <- pathsToBinary(ranked.p)</pre>
p.cluster <- pathCluster(ybinpaths, M=2)</pre>
plotPathCluster(ybinpaths, p.cluster, m=2, tol=0.05)
```

plotPaths 39

plotPaths	Plots an annotated igraph object higlighting ranked paths.

Description

This function plots a network highlighting ranked paths. If path.clusters are provided, paths in the same cluster are assigned similar colors.

Usage

```
plotPaths(
  paths,
  graph,
  path.clusters = NULL,
  col.palette = palette(),
  layout = layout.auto,
  ...
)
```

Arguments

Value

Produces a plot of the network with paths highlighted. If paths are computed for several labels (sample categories), a plot is created for each label.

Author(s)

Ahmed Mohamed

See Also

```
Other Plotting methods: colorVertexByAttr(), layoutVertexByAttr(), plotAllNetworks(), plotClassifierROC(), plotClusterMatrix(), plotCytoscapeGML(), plotNetwork(), plotPathClassifier()
```

40 predictPathClassifier

Examples

```
## Prepare a weighted reaction network.
## Conver a metabolic network to a reaction network.
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)</pre>
## Assign edge weights based on Affymetrix attributes and microarray dataset.
# Calculate Pearson's correlation.
data(ex_microarray) # Part of ALL dataset.
rgraph <- assignEdgeWeights(microarray = ex_microarray, graph = rgraph,</pre>
weight.method = "cor", use.attr="miriam.uniprot",
y=factor(colnames(ex_microarray)), bootstrap = FALSE)
## Get ranked paths using probabilistic shortest paths.
ranked.p <- pathRanker(rgraph, method="prob.shortest.path",</pre>
   K=20, minPathSize=6)
## Plot paths.
plotPaths(ranked.p, rgraph)
## Convert paths to binary matrix, build a classifier.
ybinpaths <- pathsToBinary(ranked.p)</pre>
p.class <- pathClassifier(ybinpaths, target.class = "BCR/ABL", M = 3)</pre>
## Plotting with clusters, on a metabolic graph.
plotPaths(ranked.p, ex_sbml, path.clusters=p.class)
```

predictPathClassifier Predicts new paths given a pathClassifier model.

Description

Predicts new paths given a pathClassifier model.

Usage

```
predictPathClassifier(mix, newdata)
```

Arguments

mix The result from pathClassifier.

newdata A data.frame containing the new paths to be classified.

Value

A list with the following elements.

h The posterior probabilities for each HME3M component.

predictPathClassifier 41

```
posterior.probs
```

The posterior probabilities for HME3M model to classify the response.

label A vector indicating the HME3M cluster membership.

component The HME3M component membership for each pathway.

path.probabilities

The 3M path probabilities.

plr.probabilities

The PLR predictions for each component.

Author(s)

Timothy Hancock and Ichigaku Takigawa

See Also

```
Other Path clustering & classification methods: pathClassifier(), pathCluster(), pathSToBinary(), plotClassifierROC(), plotClusterMatrix(), plotPathClassifier(), plotPathCluster(), predictPathCluster()
```

```
## Prepare a weighted reaction network.
## Conver a metabolic network to a reaction network.
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)</pre>
## Assign edge weights based on Affymetrix attributes and microarray dataset.
# Calculate Pearson's correlation.
data(ex_microarray) # Part of ALL dataset.
rgraph <- assignEdgeWeights(microarray = ex_microarray, graph = rgraph,</pre>
weight.method = "cor", use.attr="miriam.uniprot",
y=factor(colnames(ex_microarray)), bootstrap = FALSE)
## Get ranked paths using probabilistic shortest paths.
ranked.p <- pathRanker(rgraph, method="prob.shortest.path",</pre>
    K=20, minPathSize=6)
## Convert paths to binary matrix.
ybinpaths <- pathsToBinary(ranked.p)</pre>
p.class <- pathClassifier(ybinpaths, target.class = "BCR/ABL", M = 3)</pre>
## Just an example of how to predict cluster membership
pclass.pred <- predictPathCluster(p.class, ybinpaths$paths)</pre>
```

42 predictPathCluster

predictPathCluster Predicts new paths given a pathCluster model

Description

Predicts new paths given a pathCluster model.

Usage

```
predictPathCluster(pfit, newdata)
```

Arguments

pfit The pathway cluster model trained by pathCluster or pathClassifier.

newdata The binary pathway dataset to be assigned a cluster label.

Value

A list with the following elements:

```
labels a vector indicating the 3M cluster membership.

posterior.probs a matrix of posterior probabilities for each path belonging to each cluster.
```

Author(s)

```
Ichigaku Takigawa
Timothy Hancock
```

See Also

Other Path clustering & classification methods: pathClassifier(), pathCluster(), pathsToBinary(), plotClassifierROC(), plotClusterMatrix(), plotPathClassifier(), plotPathCluster(), predictPathClassifier()

```
## Prepare a weighted reaction network.
## Conver a metabolic network to a reaction network.
data(ex_sbml) # bipartite metabolic network of Carbohydrate metabolism.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)

## Assign edge weights based on Affymetrix attributes and microarray dataset.
# Calculate Pearson's correlation.
data(ex_microarray) # Part of ALL dataset.
rgraph <- assignEdgeWeights(microarray = ex_microarray, graph = rgraph,
    weight.method = "cor", use.attr="miriam.uniprot", bootstrap = FALSE)

## Get ranked paths using probabilistic shortest paths.</pre>
```

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```
ranked.p <- pathRanker(rgraph, method="prob.shortest.path",
    K=20, minPathSize=8)

## Convert paths to binary matrix.
ybinpaths <- pathsToBinary(ranked.p)
p.cluster <- pathCluster(ybinpaths, M=2)

## just an example of how to predict cluster membership.
pclust.pred <- predictPathCluster(p.cluster,ybinpaths$paths)</pre>
```

registerMemoryErr

Internal method to register memery errors.

Description

Internal method to register memery errors, caused by compiled code. This method is used only by the package, and should not be invoked by users.

Usage

```
registerMemoryErr(method)
```

Arguments

method

The mathod which generated the error.

Author(s)

Ahmed Mohamed

reindexNetwork

Replaces current vertex ids with chosen attribute.

Description

This function allows users to replace vertex ids with another attribute, calculating connectivities based on the new attribute.

Usage

```
reindexNetwork(graph, v.attr)
```

Arguments

graph An annotated igraph object.

v.attr Name of the attribute to use as vertex ids.

rmSmallCompounds

Details

This functions can be very useful when merging networks constructed from different databases. For example, to match a network created from Reactome to a KEGG network, you can reindex the vertices by "miriam.kegg.compound" attribute. Another usage is to remove duplicated vertices (in case of different subcellular compartments, for example). if a network has ATP_membrane & ATP_cytoplasm vertices, reindexing by chemical name will collapse them into one 'ATP' vertex.

Value

A new graph with vertices expanded.

Author(s)

Ahmed Mohamed

See Also

Other Network processing methods: expandComplexes(), makeMetaboliteNetwork(), makeReactionNetwork(), rmSmallCompounds(), simplifyReactionNetwork(), vertexDeleteReconnect()

```
## Make a gene network from a reaction network.
data(ex_sbml) # A bipartite metbaolic network.
rgraph <- makeReactionNetwork(ex_sbml, simplify=TRUE)</pre>
ggraph <- makeGeneNetwork(rgraph)</pre>
## Expand vertices into their contituent genes.
data(ex_kgml_sig) # Ras and chemokine signaling pathways in human
ggraph <- expandComplexes(ex_kgml_sig, v.attr = "miriam.ncbigene",</pre>
     keep.parent.attr= c("^pathway", "^compartment"))
## Create a separate vertex for each compartment. This is useful in duplicating
## metabolite vertices in a network.
## Not run:
graph <- expandComplexes(graph, v.attr = "compartment",</pre>
        keep.parent.attr = "all",
        expansion.method = "duplicate",
        missing.method = "keep")
## End(Not run)
```

rmSmallCompounds 45

Description

This function removes uniquitous compounds (metabolites connected to numerous reactions) from a metabolic network. These compounds are reaction cofactors and currency compounds, such as ATP, CO2, etc. A path through these metabolites may not be bioloigically meaningful. The defualt small compound list is derived from Reactome, containing keeg.compound, pubchem.compound, ChEBI and CAS identifiers.

Usage

```
rmSmallCompounds(
  graph,
  method = c("remove", "duplicate"),
  small.comp.ls = NPMdefaults("small.comp.ls")
)
```

Arguments

graph A metabolic network.

method How to handle small compounds. Either simply delete these vertices "remove"

(default), or make a separate vertex for each reaction they participate in "duplicate".

small.comp.ls A list of small compounds to be used.

Value

A modified graph, with the small compounds removed or duplicated.

Author(s)

Ahmed Mohamed

See Also

Other Network processing methods: expandComplexes(), makeMetaboliteNetwork(), makeReactionNetwork(), reindexNetwork(), simplifyReactionNetwork(), vertexDeleteReconnect()

```
data(ex_sbml)
sbml.removed <- rmSmallCompounds(ex_sbml, method="remove")</pre>
```

46 SBML2igraph

SBML2igraph

Processes SBML files into igraph objects

Description

This function takes SBML files as input, and returns either a metabolic or a signaling network as output.

Usage

```
SBML2igraph(
  filename,
  parse.as = c("metabolic", "signaling"),
  miriam.attr = "all",
  gene.attr,
  expand.complexes,
  verbose = TRUE
)
```

Arguments

A character vector containing the SBML files to be processed. If a directory path is provided, all *.xml and *.sbml files in it and its subdirectories are included.

parse.as Whether to process file into a metabolic or a signaling network.

Miriam.attr A list of annotation attributes to be extracted. If "all", then all attibutes written in MIRIAM guidelines (see Details) are extracted (Default). If "none", then no attributes are extracted. Otherwise, only attributes matching those specified are extracted.

An attribute to distinguish species representing genes from those representing small molecules (see Details). Ignored if parse.as="metabolic".

expand.complexes

Split protein complexes into individual gene nodes. Ignored if parse.as="metabolic", or when gone attrib not provided

or when gene.attr is not provided.

verbose Whether to display the progress of the function.

Details

Users can specify whether files are processes as metabolic or signaling networks.

Metabolic networks are given as bipartite graphs, where metabolites and reactions represent vertex types. This is constructed from ListOfReactions in SBML file, connecting them to their corresponding substrates and products (ListOfSpecies). Each reaction vertex has genes attribute, listing all modifiers of this reaction. As a general rule, reactions inherit all annotation attributes of its catalyzig genes.

Signaling network have genes as vertices and edges represent interactions. Since SBML format may represent singling events as reaction, all species are assumed to be genes (rather than small

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molecules). For a simple path S0 -> R1 -> S1, in signaling network, the path will be S0 -> M(R1) -> S1 where M(R1) is R1 modifier(s). To ditiguish gene species from small molecules, user can provide gene.attr (for example: miriam.uniprot or miriam.ncbigene) where only annotated species are considered genes.

All annotation attributes written according to MIRIAM guidlines (either urn:miriam:xxx:xxx or http://identifiers.org/xxx/xxx) are etxracted by default. Non-conforming attributes can be extracted by specifying miriam.attr.

To generate a genome scale network, simply provide a list of files to be parsed, or put all file in a directory, as pass the directory path as filename

Note: This function requires libSBML installed (Please see the installation instructions in the Vignette). Some SBML level-3 files may requires additional libraries also (An infomative error will be displayed when parsing such files). Please visit http://sbml.org/Documents/Specifications/SBML_Level_3/Packages for more information.

Value

An igraph object, representing a metbolic or a signaling network.

Author(s)

Ahmed Mohamed

See Also

Other Database extraction methods: KGML2igraph(), biopax2igraph()

```
simplifyReactionNetwork
```

Removes reactions with no gene annotations

Description

This function removes reaction vertices with no gene annotations as indicated by the parameter gene.attr, and connect their neighbour vertices to preserve graph connectivity. This is particularly meaningful when reactions are translocation or spontaneous reactions, which are not catalysed by genes.

Usage

```
simplifyReactionNetwork(
  reaction.graph,
  gene.attr = "genes",
  remove.missing.genes = TRUE,
  reconnect.threshold = vcount(reaction.graph)
)
```

Arguments

```
reaction.graph A reaction network.

gene.attr The attribute to be considered as "genes". Reactions missing this annotation, will be removed.

remove.missing.genes

If FALSE, only transocation and spontaneous reactions are removed, otherwise all rections with no gene annotations are removed.

reconnect.threshold

An argument passed to vertexDeleteReconnect
```

Value

A simplified reaction network.

Author(s)

Ahmed Mohamed

See Also

Other Network processing methods: expandComplexes(), makeMetaboliteNetwork(), makeReactionNetwork(), reindexNetwork(), rmSmallCompounds(), vertexDeleteReconnect()

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Examples

```
data(ex_sbml)
rgraph <- makeReactionNetwork(ex_sbml, simplify=FALSE)
## Removes all reaction nodes with no annotated genes.
rgraph <- simplifyReactionNetwork(rgraph, remove.missing.genes=TRUE)</pre>
```

stdAttrNames

MIRIAM annotation attributes

Description

These functions deals with conforming with MIRIAM annotation guidelines, conversion and mapping between MIRIAM identifiers.

Usage

```
stdAttrNames(graph, return.value = c("matches", "graph"))
fetchAttribute(
  graph,
  organism = "Homo sapiens",
  target.attr,
  source.attr,
  bridge.web = NPMdefaults("bridge.web")
)
```

Arguments

graph	An annotated igraph object.
return.value	Specify whether to return the names of matched standard annotations, or modify the graph attribute names to match the standards.
organism	The latin name of the organism (Case-sensitive).
target.attr	The target annotation, given as MIRIAM standard in the format miriam.xxx
source.attr	The source annotation attribute from graph
bridge.web	The base URL for Brigde Database webservices.

Value

For stdAttrNames, matches gives the original attribute names and their MIRIAM version. Since this is done by simple text matching, mismatches may occur for ambiguous annotations (such as GO, EC number). graph returns the input graph with attribute names standardized.

For fetchAttribute, the input graph with the fetched attribute mapped to vertices.

50 toGraphNEL

Author(s)

Ahmed Mohamed

See Also

Other Attribute handling methods: getAttrStatus()

Examples

toGraphNEL

Converts an annotated igraph object to graphNEL

Description

Converts an annotated igraph object to graphNEL

Usage

```
toGraphNEL(graph, export.attr = "")
```

Arguments

graph An annotated igraph object..

export.attr A regex experssion representing vertex attributes to be exported to the new

graphNEL object. Supplying an empty string "" (default) will export all at-

tributes.

Value

A graphNEL object.

Author(s)

Ahmed Mohamed

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Examples

```
data(ex_kgml_sig) # Ras and chemokine signaling pathways in human
graphNEL <- toGraphNEL(ex_kgml_sig, export.attr="^miriam.")</pre>
```

vertexDeleteReconnect Network editing: removing vertices and connecting their neighbours

Description

This function removes vertices given as vids and connects their neighbours as long as the shortest path beween the neighbours are below the reconnect.threshold.

Usage

```
vertexDeleteReconnect(
  graph,
  vids,
  reconnect.threshold = vcount(graph),
  copy.attr = NULL
)
```

Arguments

graph A reaction network.

vids Vertex ids to be removed.

reconnect.threshold

If the shortest path between vertices is larger than this threshold, they are not

reconnected.

copy.attr A function, or a list of functions, combine edge attributes. Edge attributes of new

edges (between reconnected neighbours) are obtained by combining original

edges attributes along the shortest path between reconnected neighbors.

Value

A modified graph.

Author(s)

Ahmed Mohamed

See Also

Other Network processing methods: expandComplexes(), makeMetaboliteNetwork(), makeReactionNetwork(), reindexNetwork(), rmSmallCompounds(), simplifyReactionNetwork()

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```
## Remove all reaction vertices from a bipartite metabolic network
## keeping only metabolite vertices.
data(ex_sbml)
graph <- vertexDeleteReconnect(ex_sbml, vids=which(V(ex_sbml)$reactions))</pre>
```

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