Package ‘pandaR’

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Title    PANDA Algorithm
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Description Runs PANDA, an algorithm for discovering novel network structure
by combining information from multiple complementary data sources.
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calcDegree ..................................................... 2
calcDegreeDifference ........................................ 3
importPandaMatlab ............................................. 3
multiplot ...................................................... 4
panda .......................................................... 4
pandaR .......................................................... 6
pandaResult .................................................... 6
pandaResultPairs .............................................. 7
pandaToyData ................................................... 7
plot.panda ...................................................... 8
plotCommunityDetection ................................. 8
plotGraph ...................................................... 9
plotZ .......................................................... 10
## Calculate regulatory network degree

### Description

Calculates the transcription factor out-degree or gene in-degree for the estimated panda regulatory network.

### Usage

```r
calcDegree(x, type = c("tf", "gene"), filter = FALSE, trim = FALSE, ...)
```

### Arguments

- `x`: An object of class "panda" or matrix
- `type`: Character string - 'tf' or 'gene'
- `filter`: Boolean to force negative degrees to zero
- `trim`: Boolean to trim using topedges or not at a cutoff (weights become binary 1,0)
- `...`: Options to be passed to topedges function

### Examples

```r
data(pandaToyData)
pandaRes <- panda(pandaToyData$motif, 
                   pandaToyData$expression, pandaToyData$ppi, hamming = .001, progress = TRUE)
calcDegree(pandaRes)
calcDegree(pandaRes, trim = TRUE, cutoff = 1.5)

data(pandaResult)
calcDegree(pandaResult, type = "tf", trim = TRUE, 1000)
calcDegree(pandaResult, type = "gene", trim = TRUE, 1000)
```
calcDegreeDifference  

Calculate difference in degrees

Description

Calculates the transcription factor out-degree or gene in-degree for two different panda regulatory networks. This is useful in comparing networks from two phenotypes.

Usage

calcDegreeDifference(x, y, type = c("tf", "gene"), filter = FALSE, trim = FALSE, ...)

Arguments

x  
An object of class "panda" or matrix

y  
A second object of class "panda" or matrix

type  
Character string - 'tf' or 'gene'

filter  
Boolean to force negative degrees to zero

trim  
Boolean to trim using topedges or not at a cutoff (weights become binary 1,0)

...  
Options to be passed to topedges function

Examples

data(pandaToyData)
pandaRes <- panda(pandaToyData$motif, pandaToyData$expression, pandaToyData$ppi, hamming=.001, progress=TRUE)
pandaRes2 <- panda(pandaToyData$motif, pandaToyData$expression, pandaToyData$ppi, hamming=.1, progress=TRUE)
calcDegreeDifference(pandaRes, pandaRes2)
calcDegreeDifference(pandaRes, pandaRes2, trim=TRUE, cutoff=1.5)

importPandaMatlab  
Panda Matlab importer

Description

Imports the files from the exportPanda.m file.

Usage

importPandaMatlab(dir = getwd(), celldata = "celldata.dat")

Arguments

dir  
Working directory to search for the numeric files.

celldata  
Name of the 'celldata.dat' file.
Value

Two column vector of "regulator" and "target"

Examples

```r
# determine gene degree
pandaFiles = importPandaMatlab()
indegree <- ddply(pandaFiles[,2:ncol(pandaFiles)],.(targer), numcolwise(sum))
row.names(indegree) <- indegree[,1]
indegree <- indegree[,,-1]
# to export the file
networkFiles = list.files(pattern="numeric")
write.table(indegree,paste("indegree_",networkFiles,sep=""),
  sep="\t",quote=F,row.names=T,col.names=T)
```

multiplot  

*Multiple plots*

Description

Multiple plot function as described in: http://www.cookbook-r.com/Graphs/Multiple_graphs_on_one_page_(ggplot2)/. If the layout is something like matrix(c(1,2,3,3), nrow=2, byrow=TRUE), then plot 1 will go in the upper left, 2 will go in the upper right, and 3 will go all the way across the bottom.

Usage

```r
multiplot(..., plotlist = NULL, cols = 1, layout = NULL)
```

Arguments

- `...` ggplot objects can be passed in or to plotlist (as a list of ggplot objects).
- `plotlist` NULL - if the plotlist is null, the function will figure out the panel dimensions.
- `cols` Number of columns in layout.
- `layout` A matrix specifying the layout. If present, 'cols' is ignored.

panda  

*Passing Messages between Biological Networks to Refine Predicted Interactions*

Description

This function runs the PANDA algorithm
panda

Usage

panda(motif, expr = NULL, ppi = NULL, alpha = 0.1, hamming = 0.001, iter = NA, output = c("regulatory", "coexpression", "cooperative"), zScale = TRUE, progress = FALSE, randomize = c("None", "within.gene", "by.gene"), cor.method = "pearson", scale.by.present = FALSE, edgelist = FALSE, remove.missing.ppi = FALSE, remove.missing.motif = FALSE, remove.missing.genes = FALSE)

Arguments

motif A motif dataset, a data.frame, matrix or exprSet containing 3 columns. Each row describes an motif associated with a transcription factor (column 1) a gene (column 2) and a score (column 3) for the motif.

expr An expression dataset, as a genes (rows) by samples (columns) data.frame

ppi A Protein-Protein interaction dataset, a data.frame containing 3 columns. Each row describes a protein-protein interaction between transcription factor 1(column 1), transcription factor 2 (column 2) and a score (column 3) for the interaction.

alpha value to be used for update variable, alpha (default=0.1)

hamming value at which to terminate the process based on hamming distance (default 10^-3)

iter sets the maximum number of iterations PANDA can run before exiting.

output a vector containing which networks to return. Options include "regulatory", "coregulatory", "cooperative".

zScale Boolean to indicate use of z-scores in output. False will use [0,1] scale.

progress Boolean to indicate printing of output for algorithm progress.

randomize method by which to randomize gene expression matrix. Default "None". Must be one of "None", "within.gene", "by.genes". "within.gene" randomization scrambles each row of the gene expression matrix, "by.gene" scrambles gene labels.

cor.method Correlation method, default is "pearson".

scale.by.present Boolean to indicate scaling of correlations by percentage of positive samples.

edgelist Boolean to indicate if edge lists instead of matrices should be returned.

remove.missing.ppi Boolean to indicate whether TFs in the PPI but not in the motif data should be removed.

remove.missing.motif Boolean to indicate whether genes targeted in the motif data but not the expression data should be removed.

remove.missing.genes Boolean to indicate whether genes in the expression data but lacking information from the motif prior should be removed.

Value

An object of class "panda" containing matrices describing networks achieved by convergence with PANDA algorithm.

"regNet" is the regulatory network
"coregNet" is the coregulatory network
"coopNet" is the cooperative network
References


Examples

data(pandaToyData)
pandaRes <- panda(pandaToyData$motif, pandaToyData$expression, pandaToyData$ppi, hamming=.1, progress=TRUE)

pandaResult

Analysis result from PANDA algorithm on toy data

Description

The PANDA approach is to model the regulatory network as a bipartite network and estimate edge weights based on the evidence that information from a particular transcription factor i is successfully being passed to a particular gene j. This package provides a straightforward tool for applying this established method.

References

pandaResultPairs

| pandaResultPairs | Analysis result from PANDA algorithm on toy data converted into pairs |

**Description**

This data panda object resulting from running the PANDA algorithm on the supplied toy dataset. The data consists of a matrix of TF, Gene, initial link, and final Z. data(pandaResultPairs)

**Usage**

pandaResultPairs

**Format**

A matrix

**Value**

A matrix

**References**


pandaToyData

| pandaToyData | Toy gene expression, motif, and ppi data |

**Description**

This data is a list containing three data.frames. The motif data.frame describes a set of pairwise connections where a specific known sequence motif of a transcription factor was found upstream of the corresponding gene. The expression data.frame is a set of 1000 gene expression levels measured across 50 samples. Finally, the ppi data.frame describes a set of known pairwise protein interactions.

**Usage**

pandaToyData

**Format**

A list containing 3 data.frames

**Value**

A list of length 3

**References**

plotCommunityDetection

Community detection plot

Description

This function performs community detection on an undirected PANDA network. The function optionally returns the graph and community.

Usage

plotCommunityDetection(x, scaleEdge = 5, verbose = TRUE, ...)

Arguments

x Toy PANDA output represented as a TF, Gene, and Score.
scaleEdge Visualization parameter for the edges.
verbose TRUE/FALSE - Report community structure.
... Options for the plot function.
plotGraph

Description

plotGraph plots a bipartite graph

Usage

plotGraph(x)

Arguments

x an object of class "panda"

Value

An matrix describing the subsetted bipartite network.

Examples

data(pandaToyData)
pandaRes <- panda(pandaToyData$motif,
 pandaToyData$expression,pandaToyData$ppi,hamming=.001,progress=TRUE)
topPandaRes <- topedges(pandaRes,1000)
subnet.pandaRes <- subnetwork(topPandaRes,c("AR","ARID3A","ELK1"))
plotGraph(subnet.pandaRes)

data(pandaResult)
topPandaRes <- topedges(pandaResult, 1000)
subnet.pandaRes <- subnetwork(topPandaRes,c("AR","ARID3A","ELK1"))
plotGraph(subnet.pandaRes)

Value

Optionally return a list with the graph and community.

Examples

# start with some toy PANDA output
mat <- cbind(rep(1:5, each=10), rep(seq(11,20),5), sample(100, 50)/100)
x =plotCommunityDetection(mat)
str(x)

# example of very different edges
set.seed(1)
subst <- sample(50,10)
mat[subst, 3] <- subst
plotCommunityDetection(mat,scaleEdge=0.5)
**plotZ**  
*Comparison of Z scores between two PANDA runs*

**Description**

Given two PANDA objects with the same network structure, plot the Z-score comparison. The two PANDA objects should only differ in the gene expression used for the network constructions or other parameters.

**Usage**

```
plotZ(x, y, hex = TRUE, bins = 200, addLine = TRUE, rank = FALSE)
```

**Arguments**

- `x`: PANDA object - output of the `panda` function.
- `y`: PANDA object - second PANDA object.
- `hex`: TRUE/FALSE - If TRUE, bin data points to avoid over plotting.
- `bins`: Number of bins to use for plotting.
- `addLine`: TRUE/FALSE - to add y=x line.
- `rank`: TRUE/FALSE - If TRUE, plot rank of edge weights rather than weight values.

**Value**

`ggplot` comparing the Z-scores between the two networks.

**Examples**

```
data(pandaResult)
data(pandaToyData)
pandaRes <- pandaRes2 <- pandaResult
plotZ(pandaRes, pandaRes2)
```

```
panda.res1 <- with(pandaToyData, panda(motif, expression, ppi, hamming=1))
panda.res2 <- with(pandaToyData, panda(motif, expression +
    rnorm(prod(dim(expression)),sd=5), ppi, hamming=1))
plotZ(panda.res1, panda.res2, addLine=FALSE)
```

**plotZbyTF**  
*Plot Z by TF out-degree quantiles*

**Description**

Generates a Z-score scatterplot for edges according to the TF outdegree in prior. The two PANDA objects should only differ in the gene expression used for the network constructions or other parameters.
Usage

plotZbyTF(x, y, motif, hasPrior = TRUE, cuts = 1, cols = 2)

Arguments

x       PANDA object - output of the panda function.
y       PANDA object - second PANDA object.
motif       Motif used to construct the networks.
hasPrior       TRUE/FALSE. If TRUE plots the edges that are given a weight > 0 in the motif, else plot those given a weight of 0
cuts       either a numeric vector of two or more unique cut points or a single number (greater than or equal to 2) giving the number of intervals into which 'x' is to be cut.
cols       Number of columns in layout.

Value

ggplot heatmap for each TF, get outdegree in regulatory prior

Examples

data(pandaResult)
data(pandaToyData)
plotZbyTF(pandaResult,pandaResult, pandaToyData$motif, hasPrior=TRUE)
plotZbyTF(pandaResult,pandaResult, pandaToyData$motif, cuts=2)

Description

summarizes the results of a PANDA analysis

Usage

## S3 method for class 'panda'
print(x, ...)

Arguments

x       an object of class "panda"
...

Value

Summary description of panda S4 object
Examples

data(pandaToyData)
pandaRes <- panda(pandaToyData$motif,
                  pandaToyData$expression,pandaToyData$ppi,hamming=.001,progress=TRUE)
print(pandaRes)

data(pandaResult)
print(pandaResult)

---

subnetwork

Subnetwork

Description

subnetwork gets a bipartite network containing only the transcription factors or genes and their respective connections

Usage

subnetwork(x, nodes, subTf = TRUE)

Arguments

x  
an object of class "panda"

nodes  
character vector containing the transcription factor or gene labels to subset

subTf  
an optional logical indicating whether to subset by transcription factor. Default is TRUE.

Value

An matrix describing the subsetted bipartite network.

Examples

data(pandaToyData)
pandaRes <- panda(pandaToyData$motif,
                  pandaToyData$expression,pandaToyData$ppi,hamming=.001,progress=TRUE)
topPandaRes <- topedges(pandaRes,1000)
subnet.pandaRes <- subnetwork(topPandaRes,c("AR","ARID3A","ELK1"))

data(pandaResult)
topPandaRes <- topedges(pandaResult,1000)
subnetwork(topPandaRes,c("AR","ARID3A","ELK1"))
**summary.panda**

**Description**

summarizes the results of a PANDA analysis

**Usage**

```r
## S3 method for class 'panda'
summary(object, ...)
```

**Arguments**

- `object` an object of class "panda"
- `...` further arguments passed to or from other methods.

**Value**

Summary description of panda S4 object

**Examples**

```r
data(pandaToyData)
panda.res <- panda(pandaToyData$motif, pandaToyData$expression, pandaToyData$ppi, hamming=.001, progress=TRUE)
summary(panda.res)
data(pandaResult)
summary(pandaResult)
```

---

**targetedGenes**

**Description**

Gets a set of genes targeted by a specified transcription factor. This function can be applied to a graph that is not complete, subsetting the edges which have non-zero edge weight. See function `topEdges` for dichotomizing edgeweights.

**Usage**

```r
targetedGenes(x, tfs)
```

**Arguments**

- `x` an object of class "panda"
- `tfs` transcription factors to query
testMotif

Value

A vector of targeted genes

Examples

```r
data(pandaToyData)
pandaRes <- panda(pandaToyData$motif,
pandaToyData$expression,pandaToyData$ppi,hamming=.001)
topPandaRes <- topedges(pandaRes,1000)
targetedGenes(topPandaRes,c("AR","ELK1"))
data(pandaResult)
topPandaRes <- topedges(pandaResult,1000)
```

data(pandaToyData)
data(pandaResult)
with(pandaToyData, testMotif(regnet, motif, mode="augment", expression, ppi, hamming=1))
```

testMotif  

Check motif

Description

This function adds random false positive edges to the regulatory prior and will check if they become pruned.

Usage

```r
testMotif(x, motif, expr, ppi, mode = c("augment", "remove"), prop = 0.05, seed = 1, ...)
```

Arguments

- `x` : Model regulatory network.
- `motif` : Motif used to construct the model regulatory network.
- `expr` : Expression matrix used to construct model network.
- `ppi` : PPI used to construct model regulatory network.
- `mode` : a character string - either "augment" to add random edges or "remove" to remove random edges.
- `prop` : numeric specifying number of edges to augment or remove from regulatory prior, as a proportion of the number of edges in the regulatory prior.
- `seed` : Random seed.
- `...` : Options for the panda function.

Value

ggplot heatmap list of indices of net corresponding to each TF

Examples

```r
data(pandaToyData)
data(pandaResult)
regnet = slot(pandaResult,"regNet")
with(pandaToyData, testMotif(regnet, motif, mode="augment", expression, ppi, hamming=1))
```

Description

topedges gets a network from a panda obj with a specified cutoff based on magnitude of edgeweight.

Usage

topedges(x, count = NA, cutoff = 2, networks = c("coregulation", "cooperation", "regulatory"))

Arguments

x             an object of class "panda"
count         an optional integer indicating number of top edges to be included in regulatory network.
cutoff        an optional numeric indicating the z-score edge weight cutoff to be used to identify edges. Default is 2.0. Not used if count is not NA.
networks      an optional vector specifying which networks to be included in output. May be any combination of c("coregulation"."cooperation"."regulatory").

Value

An object of class "panda" containing binary matrices indicating the existence of an edge between two nodes. For regulatory network the matrix indicates an edge between a transcription factor (row) and a gene (column)

Examples

data(pandaToyData)
pandaRes <- panda(pandaToyData$motif,
                 pandaToyData$expression, pandaToyData$ppi, hamming=.001, progress=TRUE)
topPandaRes <- topedges(pandaRes,1000)

data(pandaResult)
topPandaRes <- topedges(pandaResult,1000)
Index

*Topic datasets
  pandaResult, 6
  pandaResultPairs, 7
  pandaToyData, 7
*Topic keywords
  panda, 4
  plot.panda, 8
  plotGraph, 9
  print.panda, 11
  subnetwork, 12
  summary.panda, 13
  targetedGenes, 13
  topedges, 15

calcDegree, 2
calcDegreeDifference, 3
importPandaMatlab, 3
multiplot, 4
panda, 4
pandaR, 6
pandaR-package (pandaR), 6
pandaResult, 6
pandaResultPairs, 7
pandaToyData, 7
plot.panda, 8
plotCommunityDetection, 8
plotGraph, 9
plotZ, 10
plotZbyTF, 10
print.panda, 11

subnetwork, 12
summary.panda, 13

targetedGenes, 13
testMotif, 14
topedges, 15