Package 'mnem'

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Type Package

Title Mixture Nested Effects Models

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Description Mixture Nested Effects Models (mnem) is an extension of Nested Effects Models and allows for the analysis of single cell perturbation data provided by methods like Perturb-Seq (Dixit et al., 2016) or Crop-Seq (Datlinger et al., 2017). In those experiments each of many cells is perturbed by a knock-down of a specific gene, i.e. several cells are perturbed by a knock-down of gene B, ... and so forth. The observed read-out has to be multi-trait and in the case of the Perturb-/Crop-Seq gene are expression profiles for each cell. mnem uses a mixture model to simultaneously cluster the cell population into k clusters and and infer k networks causally linking the perturbed genes for each cluster. The mixture components are inferred via an expectation maximization algorithm.

Depends R (>= 3.6)

License GPL-3

Encoding UTF-8

LazyData true

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LinkingTo Rcpp, RcppEigen

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Processed scRNAsea from pooled CRISPR screens

Description

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Example data: mnem results for the Dixit et al., 2016 and Datlinger et al., pooled CRISPR screens. For details see the vignette or function createApp().

Usage

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References

Datlinger, P., Rendeiro, A., Schmidl, C., Krausgruber, T., Traxler, P., Klughammer, J., Schuster, L. C., Kuchler, A., Alpar, D., and Bock, C. (2017). Pooled crispr screening with single-cell transcriptome readout. Nature Methods, 14, 297-301.

Dixit, A., Parnas, O., Li, B., Chen, J., Fulco, C. P., Jerby-Arnon, L., Marjanovic, N. D., Dionne, D., Burks, T., Raychowdhury, R., Adamson, B., Norman, T. M., Lander, E. S., Weissman, J. S., Friedman, N., and Regev, A. (2016). Perturb-seq: Dissecting molecular circuits with scalable single-cell rna profiling of pooled genetic screens. Cell, 167(7), 1853-1866.e17.

Examples

data(app)

bootstrap 3

bootstrap	Bootstrap.
-----------	------------

Description

Run bootstrap simulations on the components (phi) of an object of class mnem.

Usage

```
bootstrap(x, size = 1000, p = 1, logtype = 2, complete = FALSE,
...)
```

Arguments

X	mnem object
size	size of the booststrap simulations
p	percentage of samples (e.g. for 100 E-genes p=0.5 means sampling 50)
logtype	logarithm type of the data (e.g. 2 for log2 data or exp(1) for natural)
complete	if TRUE, complete data log likelihood is considered (for very large data sets, e.g. 1000 cells and 1000 E-genes)
• • •	additional parameters for hte nem function

Value

returns bootstrap support for each edge in each component (phi); list of adjacency matrices

Author(s)

Martin Pirkl

Examples

```
 \begin{array}{l} sim <- simData(Sgenes = 3, Egenes = 2, Nems = 2, mw = c(0.4,0.6)) \\ data <- (sim\$data - 0.5)/0.5 \\ data <- data + rnorm(length(data), 0, 1) \\ result <- mnem(data, k = 2, starts = 1) \\ boot <- bootstrap(result, size = 2) \\ \end{array}
```

clustNEM

Cluster NEM.

Description

This function clusters the data and performs standard nem on each cluster.

Usage

```
clustNEM(data, k = 2:10, cluster = NULL, starts = 1, logtype = 2,
  nem = TRUE, getprobspars = list(), getaffinitypars = list(), ...)
```

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Arguments

data of log ratios with cells in columns and features in rows

k number of clusters to check

cluster given clustering has to correspond to the columns of data

starts number of random starts for the kmeans algorithm

logtype logarithm type of the data
nem if FALSE only clusters the data

getprobspars list of parameters for the getProbs function

getaffinitypars

list of parameters for the getAffinity function

additional arguments for standard nem function

Value

family of nems; the first k list entries hold full information of the standard nem search

comp list of all adjacency matrices phi

mw vector of mixture weights

probs fake cell probabilities (see mw: mixture weights)

Author(s)

Martin Pirkl

Examples

```
sim <- simData(Sgenes = 3, Egenes = 2, Nems = 2, mw = c(0.4,0.6)) data <- (sim$data - 0.5)/0.5 data <- data + rnorm(length(data), 0, 1) resulst <- clustNEM(data, k = 2:3)
```

createApp

Creating app data.

Description

This function is for the reproduction of the application results in the vignette and publication. See the publication Pirkl & Beerenwinkel (2018) on how to download the data files: GSE92872_CROP-seq_Jurkat_TCR.digital_expression.csv k562_both_filt.txt GSM2396861_k562_ccycle_cbc_gbc_dict.csv GSM2396858_k562_tfs_7_cbc_gbc_dict.csv

Usage

```
createApp(sets = seq_len(3), m = NULL, n = NULL, o = NULL,
maxk = 5, parallel = NULL, path = "", dataonly = FALSE,
allcrop = FALSE, multi = FALSE, ...)
```

fitacc 5

Arguments

sets	numeric vector with the data sets: 1 (CROPseq), 2, 3 (both PERTURBseq); default is all three
m	number of Sgenes (for testing)
n	number of most variable E-genes (for testing)
0	number of samples per S-gene (for testing)
maxk	maximum number of component in mnem inference (default: 5)
parallel	number of threads for parallelisation
path	path to the data files path/file.csv
dataonly	if TRUE, only fetches and normalizes the data and computes the log odds
allcrop	if TRUE, does not restrict and uses the full CROPseq dataset
multi	if TRUE, includes cells with more than one perturbed gene
	additional parameters for the mixture nem function

Value

app data object

Author(s)

Martin Pirkl

Examples

```
## recreate the app data object (takes very long, i.e. days)
## Not run:
createApp()
## End(Not run)
data(app)
```

fitacc

Simulation accuracy.

Description

Computes the accuracy of the fit between simulated and inferred mixture.

Usage

```
fitacc(x, y, strict = FALSE, unique = TRUE, type = "ham")
```

Arguments

X	mnem object
У	simulation object or another mnem object
strict	if TRUE, accounts for over/underfitting, i.e. the number of components
unique	if TRUE, phis of x and y are made unique each (FALSE if strict is TRUE)
type	type of accuracy. "ham" for hamming, "sens" for sensitivity and "spec for Speci-
	ficity"

6 fuzzyindex

Value

plot of EM convergence

Author(s)

Martin Pirkl

Examples

fuzzyindex

Calculate fuzzy ground truth.

Description

Calculates responsibilities and mixture weights based on the ground truth and noisy data.

Usage

```
fuzzyindex(x, data, logtype = 2, complete = FALSE, ...)
```

Arguments

x mnemsim objectdata noisy data matrix

logtype logarithm type of the data

complete if TRUE, complete data log likelihood is considered (for very large data sets,

e.g. 1000 cells and 1000 E-genes)

... additional parameters for the function getAffinity

Value

list with cell log odds mixture weights and log likelihood

Author(s)

Martin Pirkl

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Examples

getAffinity

Calculate responsibilities.

Description

This function calculates the responsibilities of each component for all cells from the expected log distribution of the hidden data.

Usage

```
getAffinity(x, affinity = 0, norm = TRUE, logtype = 2, mw = NULL,
  data = matrix(0, 2, ncol(x)), complete = FALSE)
```

Arguments

x log odds for l cells and k components as a kxl matrix

affinity 0 for standard soft clustering, 1 for hard clustering during inference (not recommended)

norm if TRUE normalises to probabilities (recommended)

logtype logarithm type of the data (e.g. 2 for log2 data or exp(1) for natural)

mw mixture weights of the components

data in log odds

complete if TRUE, complete data log likelihood is considered (for very large data sets,

e.g. 1000 cells and 1000 E-genes)

Value

responsibilities as a kxl matrix (k components, l cells)

Author(s)

Martin Pirkl

```
sim <- simData(Sgenes = 3, Egenes = 2, Nems = 2, mw = c(0.4,0.6)) data <- (sim\$data - 0.5)/0.5 data <- data + rnorm(length(data), 0, 1) result <- mnem(data, k = 2, starts = 1) resp <- getAffinity(result$probs, mw = result$mw, data = data)
```

8 getIC

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Calculate negative penalized log likelihood.

Description

This function calculates a negative penalized log likelihood given a object of class mnem. This penalized likelihood is based on the normal likelihood and penalizes complexity of the mixture components (i.e. the networks).

Usage

```
getIC(x, man = FALSE, degree = 4, logtype = 2, pen = 2,
  useF = FALSE, Fnorm = FALSE)
```

Arguments

Χ	mnem object

man logical. manual data penalty, e.g. man=TRUE and pen=2 for an approximation

of the Akaike Information Criterion

degree different degree of penalty for complexity: positive entries of transitively re-

duced phis or phi^r (degree=0), phi^r and mixture components minus one k-1 (1), phi^r, k-1 and positive entries of thetas (2), positive entries of transitively closed phis or phi^t, k-1 (3), phi^t, theta, k-1 (4, default), all entries of phis,

thetas and k-1 (5)

logtype logarithm type of the data (e.g. 2 for log2 data or exp(1) for natural)

pen penalty weight for the data (e.g. pen=2 for approximate Akaike Information

Criterion)

use F (see publication) as complexity instead of phi and theta

Fnorm normalize complexity of F, i.e. if two components have the same entry in F, it is

only counted once

Value

penalized log likelihood

Author(s)

Martin Pirkl

```
sim <- simData(Sgenes = 3, Egenes = 2, Nems = 2, mw = c(0.4,0.6))
data <- (sim$data - 0.5)/0.5
data <- data + rnorm(length(data), 0, 1)
pen <- numeric(3)
result <- list()
for (k in seq_len(2)) {
    result[[k]] <- mnem(data, k = k, starts = 1)
    pen[k] <- getIC(result[[k]])
}
print(pen)</pre>
```

hamSim 9

Accuracy for two phis.

Description

This function uses the hamming distance to calculate an accuracy for two networks (phi).

Usage

```
hamSim(a, b, diag = 1, symmetric = TRUE)
```

Arguments

a adjacency matrix (phi)
b adjacency matrix (phi)

diag if 1 includes diagonal in distance, if 0 not

symmetric comparing a to b is asymmetrical, if TRUE includes comparison b to a

Value

normalized hamming accuracy for a and b

Author(s)

Martin Pirkl

Examples

```
sim <- simData(Sgenes = 3, Egenes = 2, Nems = 2, mw = c(0.4,0.6)) similarity <- hamSim(sim$Nem[[1]], sim$Nem[[2]])
```

mnem

Mixture NEMs - main function.

Description

This function simultaneously learns a mixture of causal networks and clusters of a cell population from single cell perturbation data (e.g. log odds of fold change) with a multi-trait readout. E.g. Pooled CRISPR scRNA-Seq data (Perturb-Seq. Dixit et al., 2016, Crop-Seq. Datlinger et al., 2017).

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Usage

```
mnem(D, inference = "em", search = "greedy", phi = NULL,
    theta = NULL, mw = NULL, method = "llr", parallel = NULL,
    reduce = FALSE, runs = 1, starts = 3, type = "random",
    complete = FALSE, p = NULL, k = NULL, kmax = 10,
    verbose = FALSE, max_iter = 100, parallel2 = NULL,
    converged = -Inf, redSpace = NULL, affinity = 0,
    evolution = FALSE, lambda = 1, subtopoX = NULL, ratio = TRUE,
    logtype = 2, domean = TRUE, modulesize = 5, compress = FALSE,
    increase = TRUE, fpfn = c(0.1, 0.1), multi = FALSE,
    ksel = c("kmeans", "silhouette", "cor"))
```

Arguments

D data with cells indexing the columns and features (E-genes) indexing the rows

inference inference method "em" for expectation maximization

search search method for single network inference "greedy", "exhaustive" or "modules"

(also possible: "small", which is greedy with only one edge change per M-step

to make for a smooth convergence)

phi a list of n lists of k networks for n starts of the EM and k components

theta a list of n lists of k attachment vector for the E-genes for n starts of the EM and

k components

mw mixture weights; if NULL estimated or uniform

method "llr" for log ratios or foldchanges as input (see ratio)

parallel number of threads for parallelization of the number of em runs

reduce logical - reduce search space for exhaustive search to unique networks

runs number of runs for greedy search

starts number of starts for the em

type initialize with responsibilities either by "random", "cluster" (each S-gene is clus-

tered and the different S-gene clustered differently combined for several starts), "cluster2" (clustNEM is used to infer reasonable phis, which are then used as a start for one EM run), "cluster3" (global clustering as a start), or "networks"

(initialize with random phis)

complete if TRUE, optimizes the expected complete log likelihood of the model, other-

wise the log likelihood of the observed data

p initial probabilities as a k (components) times l (cells) matrix

k number of components

kmax maximum number of components when k=NULL is inferred

verbose verbose output

max_iter maximum iteration, if likelihood does not converge

parallel2 if parallel=NULL, number of threads for single component optimization

converged absolute distance for convergence between new and old log likelihood; if set to

-Inf, the EM stops if neither the phis nor thetas were changed in the most recent

iteration

redSpace space for "exhaustive" search

affinity 0 is default for soft clustering, 1 is for hard clustering

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evolution logical. If TRUE components are penelized for being different from each other. lambda smoothness value for the prior put on the components, if evolution set to TRUE subtopoX hard prior on theta as a vector with entry i equal to j, if E-gene i is attached to ratio logical, if true data is log ratios, if false foldchanges logarithm type of the data (e.g. 2 for log2 data or exp(1) for natural) logtype domean average the data, when calculating a single NEM (speed improvment)

modulesize max number of S-genes per module in module search

compress compress networks after search (warning: penelized likelihood not interpretable)

increase if set to FALSE, the algorithm will not stop if the likelihood decreases

fpfn numeric vector of length two with false positive and false negative rates for

discrete data

multi set to TRUE if the data contains multiple perturbation per sample; make sure the

samples are reasonably named, e.g. "IRF1_CTNNB1"

ksel character vector of methods for the inference of k; can combine "hc" (hierarchi-

cal clustering) or "kmeans" with "silhouette", "BIC" or "AIC"; can also include

"cor" for correlation distance (preferred) instead of euclidean

Value

object of class mnem

list of the component with each component being a list of the causal network comp

phi and the E-gene attachment theta

data input data matrix

limits list of results for all indpendent searches

11 log likelihood of the best model

11s log likelihood ascent of the best model search

vector with mixture weights mw

kxl matrix containing the cell log likelihoods of the model probs

Author(s)

Martin Pirkl

```
sim \leftarrow simData(Sgenes = 3, Egenes = 2, Nems = 2, mw = c(0.4, 0.6))
data <- (sim data - 0.5)/0.5
data <- data + rnorm(length(data), 0, 1)</pre>
result <- mnem(data, k = 2, starts = 1)</pre>
```

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mnemh

Hierarchical mixture.

Description

This function does a hierarchical mixture. That means it uses the approximate BIC to check, if there are more than one component. It recursively splits the data if there is evidence for k > 1 components.

Usage

```
mnemh(data, k = 2, logtype = 2, getprobspars = list(), ...)
```

Arguments

data data matrix either binary or log odds
k number of maximal components for each hierarchy leaf

log type of the data

getprobspars list of parameters for the getProbs function
... additional parameters for the mnem function

Value

object of class mnem

Author(s)

Martin Pirkl

Examples

```
sim <- simData(Sgenes = 3, Egenes = 2, Nems = 2, mw = c(0.4,0.6)) data <- (sim$data - 0.5)/0.5 data <- data + rnorm(length(data), 0, 1) result <- mnemh(data, starts = 1, k = 1)
```

mnemk

Learn the number of components K and optimize the mixture.

Description

High level function for learning the number of components k, if unknown.

Usage

```
mnemk(D, ks = seq_len(5), man = FALSE, degree = 4, logtype = 2,
    pen = 2, useF = FALSE, Fnorm = FALSE, ...)
```

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Arguments

D	data with cells indexing the columns and features (E-genes) indexing the rows
ks	vector of number of components k to test
man	logical. manual data penalty, e.g. man=TRUE and pen=2 for an approximation of the Akaike Information Criterion
degree	different degree of penalty for complexity: positive entries of transitively reduced phis or phi^r (degree=0), phi^r and mixture components minus one k-1 (1), phi^r, k-1 and positive entries of thetas (2), positive entries of transitively closed phis or phi^t, k-1 (3), phi^t, theta, k-1 (4, default), all entries of phis, thetas and k-1 (5)
logtype	logarithm type of the data (e.g. 2 for log2 data or exp(1) for natural)
pen	penalty weight for the data (e.g. pen=2 for approximate Akaike Information Criterion)
useF	use F (see publication) as complexity instead of phi and theta
Fnorm	normalize complexity of F, i.e. if two components have the same entry in F, it is only counted once
	additional parameters for the mnem main function

Value

list containing the result of the best k as an mnem object and the raw and penalized log likelihoods

Author(s)

Martin Pirkl

Examples

```
 \begin{array}{l} sim <- sim Data(Sgenes = 3, Egenes = 2, Nems = 2, mw = c(0.4,0.6)) \\ data <- (sim data - 0.5)/0.5 \\ data <- data + rnorm(length(data), 0, 1) \\ result <- mnemk(data, ks = seq_len(2), starts = 1) \\ \end{array}
```

plot.bootmnem

Plot bootstrap mnem result.

Description

Plot bootstrap mnem result.

Usage

```
## S3 method for class 'bootmnem' plot(x, reduce = TRUE, ...)
```

Arguments

x bootmnem object

reduce if TRUE transitively reduces the graphs

... additional parameters for the plotting function plotDNF

plot.mnem

Value

visualization of bootstrap mnem result with Rgraphviz

Author(s)

Martin Pirkl

Examples

```
 \begin{array}{l} sim <- \ sim Data(Sgenes = 3, Egenes = 2, Nems = 2, mw = c(0.4,0.6)) \\ data <- \ (sim data - 0.5)/0.5 \\ data <- \ data + rnorm(length(data), 0, 1) \\ result <- \ mnem(data, k = 2, starts = 1) \\ boot <- \ bootstrap(result, size = 2) \\ plot(boot) \\ \end{array}
```

plot.mnem

Plot mnem result.

Description

Plot mnem result.

Usage

```
## S3 method for class 'mnem'
plot(x, oma = c(3, 1, 1, 3), main = "M&NEM",
  anno = TRUE, cexAnno = 1, scale = NULL, global = TRUE,
  egenes = TRUE, sep = FALSE, tsne = FALSE, affinity = 0,
  logtype = 2, cells = TRUE, pch = ".", legend = FALSE,
  showdata = FALSE, bestCell = TRUE, showprobs = FALSE,
  shownull = TRUE, ratio = TRUE, method = "llr",
  showweights = TRUE, ...)
```

Arguments

X	mnem object
oma	outer margin
main	main text
anno	annotate cells by their perturbed gene
cexAnno	text size of the cell annotations
scale	scale cells to show relative and not absolute distances
global	if TRUE clusters all cells, if FALSE clusters cells within a component
egenes	show egene attachments, i.e. number of E-genes assigned to each S-gene
sep	seperate clusters and not put them on top of each other for better visualization
tsne	if TRUE use tsne instead of pca
affinity	use hard clustering if TRUE
logtype	logarithm type of the data (e.g. 2 for log2 data or exp(1) for natural)

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cells show cell attachments, i.e how many cells are assigned to each S-gene cell symbol pch legend show legend showdata show data if TRUE bestCell show probability of best fitting cell for each S-gene showprobs if TRUE, shows responsibilities for all cells and components shownull if TRUE, shows the null node ratio use log ratios (TRUE) or foldchanges (FALSE) "llr" for ratios method if TRUE, shows mixture weights for all components showweights additional parameters . . .

Value

visualization of mnem result with Rgraphviz

Author(s)

Martin Pirkl

Examples

```
 \begin{array}{l} sim <- simData(Sgenes = 3, Egenes = 2, Nems = 2, mw = c(0.4,0.6)) \\ data <- (sim\$data - 0.5)/0.5 \\ data <- data + rnorm(length(data), 0, 1) \\ result <- mnem(data, k = 2, starts = 1) \\ plot(result) \\ \end{array}
```

plot.mnemsim

Plot simulated mixture.

Description

Plot simulated mixture.

Usage

```
## S3 method for class 'mnemsim'
plot(x, data = NULL, logtype = 2,
  fuzzypars = list(), ...)
```

Arguments ×

data noisy data matrix (optional)
logtype logarithm type of the data
fuzzypars list of parameters for the function fuzzyindex

mnemsim object

... additional parameters for the plotting function plotDNF

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Value

visualization of simulated mixture with Rgraphviz

Author(s)

Martin Pirkl

Examples

```
sim <- simData(Sgenes = 3, Egenes = 2, Nems = 2, mw = c(0.4,0.6)) plot(sim)
```

plotConvergence

Plot convergence of EM

Description

This function plots the convergence of the different EM iterations (four figures, e.g. par(mfrow=(2,2))).

Usage

```
plotConvergence(x, col = NULL, type = "b", convergence = 0.1, ...)
```

Arguments

x mnem object

col vector of colors for the iterations

type see ?plot.default

convergence difference of when two log likelihoods are considered equal; see also conver-

gence for the function mnem()

... additional parameters of the plots/lines functions

Value

plot of EM convergence

Author(s)

Martin Pirkl

```
 \begin{aligned} & sim <- simData(Sgenes = 3, Egenes = 2, Nems = 2, mw = c(0.4,0.6)) \\ & data <- (sim\$data - 0.5)/0.5 \\ & data <- data + rnorm(length(data), 0, 1) \\ & result <- mnem(data, k = 2, starts = 1) \\ & par(mfrow=c(2,2)) \\ & plotConvergence(result) \end{aligned}
```

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plotDnf Plot disjunctive normal form.			
	plotDnf	Plot disjunctive normal form.	
	p======	J	

Description

This function visualizes a graph encoded as a disjunctive nromal form.

Usage

```
plotDnf(dnf = NULL, freq = NULL, stimuli = c(), signals = c(),
   inhibitors = c(), connected = TRUE, CNOlist = NULL, cex = NULL,
   fontsize = NULL, labelsize = NULL, type = 2, lwd = 1,
   edgelwd = 1, legend = 0, x = 0, y = 0, xjust = 0, yjust = 0,
   width = 1, height = 1, layout = "dot", main = "", sub = "",
   cex.main = 1.5, cex.sub = 1, col.sub = "grey", fontcolor = NULL,
   nodestates = NULL, simulate = NULL, edgecol = NULL,
   labels = NULL, labelcol = "blue", nodelabel = NULL,
   nodecol = NULL, bordercol = NULL, nodeshape = NULL,
   verbose = FALSE, edgestyle = NULL, nodeheight = NULL,
   nodewidth = NULL, edgewidth = NULL, lty = NULL, hierarchy = NULL,
   showall = FALSE, edgehead = NULL, edgelabel = NULL,
   edgetail = NULL, bool = TRUE, draw = TRUE, ...)
```

Arguments

dnf	Hyper-graph in disjunctive normal form, e.g. $c("A=B", "A=C+D", "E=!B")$ with the child on the left and the parents on the right of the equation with "A=C+D" for A = C AND D. Alternatively, dnf can be an adjacency matrix, which is converted on the fly to a disjunctive normal form.
freq	Frequency of hyper-edges which are placed on the edges.
stimuli	Highlights vertices which can be stimulated.
signals	Highlights vertices which regulate E-genes.
inhibitors	Highlights vertices which can be inhibited.
connected	If TRUE, only includes vertices which are connected to other vertices.
CNOlist	CNOlist object. Optional instead of stimuli, inhibitors or signals. See package CellNOptR.
cex	Global font size.
fontsize	Vertice label size.
labelsize	Edge label size.
type	Different plot types. 2 for Rgraphviz and 1 for graph.
lwd	Line width.
edgelwd	Edgeline width.
legend	0 shows no legend. 1 shows legend as a graph. 2 shows legend in a standard box.
X	x coordinate of box legend.
у	y coordinate of box legend.

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xjust Justification of legend box left, right or center (-1,1,0). yjust Justification of legend box top, bottom or middle (-1,1,0).

width Vertice width. height Vertice height.

layout Graph layout. See graphvizCapabilities()\$layoutTypes.

main Main title. sub Subtitle.

cex.main Main title font size.
cex.sub Subtitle font size.
col.sub Font color of subtitle.
fontcolor Global font color.

nodestates Binary state of each vertice.

simulate Simulate stimulation and inhibition of a list of vertices. E.g. simulate = list(stimuli

= c("A", "B"), inhibitors = c("C", "D")).

edgecol Vector with colors for every edge of the graph (not hyper-graph). E.g. an AND

gate consists of three distinct edges.

labels Vector with labels for the edges.

labelcol Vector with label colors for the edges.

nodelabel List of vertices with labels as input. E.g. labels = list(A="test", B="label for

B").

nodecol List of vertices with colors as input.
bordercol List of vertices with colors as input.

nodeshape List of vertices with shapes (diamond, box, square,...).

verbose Verbose output.

edgestyle set the edge style like dashed, can be numerical

nodeheight List of vertices with height as input.

List of vertices with width as input.

edgewidth Vector with edge widths.

1ty Vector with edge styles (line, dotted,...).

hierarchy List with the hierarchy of the vertices. E.g. list(top = c("A", "B"), bottom =

c("C", "D")).

showall See "connected" above.
edgehead Vector with edge heads.
edgelabel Vector with edge labels.
edgetail Vector with edge tails.

bool If TRUE, only shows normal graph and no AND gates.

draw Do not plot the graph and only output the graphNEL object.

... additional arguments

Value

Rgraphviz object

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Author(s)

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Examples

```
g \leftarrow c("!A+B+C=G", "C=G", "!D=G")
plotDnf(g)
```

simData

Simulate data.

Description

This function simulates single cell data from a random mixture of networks.

Usage

```
simData(Sgenes = 5, Egenes = 1, Nems = 2, reps = NULL, mw = NULL,
evolution = FALSE, nCells = 1000, uninform = 0, unitheta = FALSE,
edgeprob = 0.25, multi = FALSE, subsample = 1, scalefree = FALSE,
...)
```

Arguments

Sgenes number of Sgenes

Egenes number of Egenes

Nems number of components

reps number of replicates, if set (not realistic for cells)

mw mixture weights (has to be vector of length Nems)

evolution evolving and not purely random network, if set to TRUE

evolving and not parely random network

nCells number of cells

uninform number of uninformative Egenes

unitheta uniform theta, if TRUE

edge probability, value between 0 and 1 for sparse or dense networks

multi a vector with the percentages of cell with multiple perturbations, e.g. c(0.2,0.1,0)

for 20 no quadruple knock-downs

subsample range to subsample data. 1 means the full simulated data is used

scalefree if TRUE, graph is scale free

... additional parameters for the scale free network sampler (see 'nem' package)

Value

simulation object with meta information and data

Nem list of adjacency matrixes generatign the data

theta E-gene attachaments

data data matrix

index index for which Nem generated which cell (data column)

mw vector of input mixture weights

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```
sim <- simData(Sgenes = 3, Egenes = 2, Nems = 2, mw = c(0.4,0.6))
```

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