Package 'graphclust'

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Description Graph clustering using an agglomerative algorithm to maximize the integrated classification likelihood criterion and a mixture of stochastic block models. The method is described in the article `Model-based clustering of multiple networks with a hierarchical algorithm" by T. Rebafka (2022) <arxiv:2211.02314>.</arxiv:2211.02314>	
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ARI

Adjusted Rand index

Description

ARI to compare two clusterings or to compare two entire lists of clusterings

Usage

```
ARI(x, y)
```

Arguments

x vector with clustering, matrix with hot-one-encoding of the clustering, or a list of clusterings (in vector or matrix form)

y as x

Value

ARI (scalar of vector)

```
x <- c(1,1,2,2,3,3)
y <- c(1,1,1,2,2,2)
ARI(x,y)

x <- matrix(0, 3, 6)
x[1,1] <- x[1,2] <- x[2,3] <- x[2,4] <- x[3,5] <- x[3,6] <- 1
y <- matrix(0, 2, 6)
y[1,1] <- y[1,2] <- y[1,3] <- y[2,4] <- y[2,5] <- y[2,6] <- 1
ARI(x,y)

X <- list(c(1,1,2,2,3,3), rep(1,10))
Y <- list(c(1,1,1,2,2,2), rep(1:2,each=5))
ARI(X,Y)</pre>
```

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degreeSort	Sort stochastic block model parameter in a unique way using its graphon

Description

Sort stochastic block model parameter in a unique way using its graphon

Usage

```
degreeSort(thetaInit, outTheta = TRUE, outPerm = FALSE)
```

Arguments

thetaInit stochastic block model parameter to be sorted

outTheta if TRUE returns the sorted stochastic block model parameter

outPerm if TRUE returns the permutation of the blocks of the stochastic block model to

provide the sorted stochastic block model parameter

Value

according to the values of outTheta and outPerm the function returns the sorted stochastic block model parameter or the associated permutation of the blocks of the stochastic block model or a list with both of them

Examples

```
theta1 <- list(pi=c(.5,.5), gamma=matrix((1:4)/8,2,2))
degreeSort(theta1)
theta2 <- list(pi=c(.5,.5), gamma=matrix(4:1/8,2,2))
degreeSort(theta2)</pre>
```

fitSBMcollection

Fit a unique stochastic block model to a collection of networks

Description

fitSBMcollection() is a subversion of graphClustering() where no stopping criterion is applied. So all networks are ultimately merged to a single cluster and considered as i.i.d realisations of a single stochastic block model.

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Usage

```
fitSBMcollection(
  allAdj,
  hyperParam = list(alpha = 0.5, eta = 0.5, zeta = 0.5, lambda = 0.5),
  nbCores = 1
)
```

Arguments

allAdj list of adjacency matrices

hyperParam hyperparameters of prior distributions nbCores number of cores for parallelization

Value

list with the following fields: \$nodeClusterings is a list with the node labels for each networks, \$theta contains the estimated SBM parameter, \$ICL is the value of the ICL criterion of the final clustering

Examples

```
theta <- list(pi=c(.5,.5), gamma=matrix((1:4)/8,2,2))
obs <- rCollectSBM(rep(10,4), theta)$listGraphs
res <- fitSBMcollection(obs, nbCores=1)</pre>
```

fitSimpleSBM

Fit a stochastic block model to every network in a collection of networks.

Description

Applies the variational EM-algorithm implemented in the package blockmodels to every network.

Usage

```
fitSimpleSBM(
  allAdj,
  directed = TRUE,
  nbSBMBlocks = Inf,
  nbCores = 1,
  outCountStat = TRUE
)
```

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Arguments

allAdj list of adjacency matrices

directed Networks are directed (TRUE by default) or undirected (FALSE).

nbSBMBlocks upper bound for the number of blocks in the SBMs of the mixture components.

Default is Inf

nbCores number of cores for parallelization.

FALSE, the output is a list of parameters of the stochastic block models fitted to

every network.

Value

list of count statistics for every network or list of parameters of the stochastic block models fitted to every network.

Examples

```
theta <- list(pi=c(.5,.5), gamma=matrix((1:4)/8,2,2))
obs <- rCollectSBM(rep(10,4), theta)$listGraphs
res <- fitSimpleSBM(obs, outCountStat=FALSE, nbCores=2)</pre>
```

graphClustering

Hierarchical graph clustering algorithm

Description

Applies the hierarchical graph clustering algorithm to a collection of networks and fits a finite mixture model of stochastic block models to the data

Usage

```
graphClustering(
  allAdj,
  hyperParam = list(alpha = 0.5, eta = 0.5, zeta = 0.5, lambda = 0.5),
  returnInitial = FALSE,
  nbClust = NULL,
  nbSBMBlocks = Inf,
  initCountStat = NULL,
  initDeltaICL = NULL,
  nbCores = 1
)
```

Arguments

allAdj list of adjacency matrices

hyperParam hyperparameters of prior distributions

returnInitial Boolean. Return SBM parameters from initialization or not. Default is FALSE.

nbClust desired number of clusters. Default NULL, which means that the number of

clusters is chosen automatically via the ICL criterion

nbSBMBlocks upper bound for the number of blocks in the SBMs of the mixture components.

Default is Inf

initCountStat initial count statistics may be provided to the method. Default is NULL. initial deltaICL-matrix may be provided to the method. Default is NULL.

nbCores number of cores for parallelization

Value

list with the following fields: \$graphGroups is the graph clustering, \$nodeClusterings is a list with the node labels for each networks, \$thetaMixSBM contains the estimated parameter of the mixture of SBMs, \$ICL is the value of the ICL criterion of the final clustering, \$histGraphGroups traces the history of the cluster aggregations, \$histDeltaICL traces the evolution of the deltaICL value, \$histFusedClusters traces the history of the aggregated cluster numbers

Examples

```
theta <- list(pi=c(.5,.5), gamma=matrix((1:4)/8,2,2))
obs <- rCollectSBM(rep(10,4), theta)$listGraphs
res <- graphClustering(obs, nbCores=1)</pre>
```

graphMomentsClustering

Graph clustering method using graph moments

Description

Graph clustering method based on graph moments by Mukherjee et al. (2017)

Usage

```
graphMomentsClustering(Networks, nbMoments = 3, nbClusters)
```

Arguments

Networks list of adjacency matrices

nbMoments order of the largest graph moments to be considered

nbClusters desired number of clusters

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Value

vector with the clustering of the networks

Examples

```
param <- vector('list', 3)</pre>
param[[1]] <- list(prop = 1/3, # component 1 : alpha > beta
                    alpha = .04,
                    beta = .02,
                    deltaIn = 100
                    deltaOut = 100,
                    R = 500
)
param[[2]] <- list(prop = 1/3, # component 2 : just permute alpha and beta ;</pre>
                    alpha = .01,
                    beta = .02,
                    deltaIn = 100,
                    deltaOut = .1,
                    R = 1000
)
param[[3]] <- list(prop = 1/3, # component 3 : alpha=beta</pre>
                    alpha = .015,
                    beta = .015,
                    deltaIn = .1,
                    deltaOut = .1,
                    R = 1000
)
obs <- sampleDPAMixture(M=20, param)</pre>
res <- graphMomentsClustering(obs$listAdj, 3, 3)</pre>
table(res, obs$graphGroups)
```

graphonL2norm

(squared) L2-norm of the graphons associated with two stochastic block model parameters

Description

(squared) L2-norm of the graphons associated with two stochastic block model parameters

Usage

```
graphonL2norm(theta1, theta2)
```

Arguments

```
theta1 a stochastic block model parameter
theta2 a stochastic block model parameter
```

Value

(squared) L2-norm of the graphons associated with two stochastic block model parameters

Examples

```
theta1 <- list(pi=c(.5,.5), gamma=matrix((1:4)/8,2,2)) theta2 <- list(pi=c(.5,.5), gamma=matrix(4:1/8,2,2)) graphonL2norm(theta1, theta2)
```

graphonSpectralClustering

Graph clustering using the pairwise graphon distances and spectral clustering

Description

Graph clustering using the pairwise graphon distances and spectral clustering

Usage

```
graphonSpectralClustering(allAdj, nbClusters, sig = 0.1, nbCores = 1)
```

Arguments

allAdj list of adjacency matrices

nbClusters number of clusters to be found

sig parameter for Gaussian kernel used for the similarity matrix

nbCores number of cores for parallelization.

Value

list with the obtained graph clusteirng (\$clust) and the matrix with the pairwise graphon distances between all pairs of networks

```
theta <- list(pi=c(.5,.5), gamma=matrix((1:4)/8,2,2))
obs <- rCollectSBM(rep(10,4), theta)$listGraphs
res <- graphonSpectralClustering(obs, 2, nbCores=1)</pre>
```

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metagraph	Plot the metagraph of the parameter of the stochastic block model
048. ap	associated with one of the estimated graph clusters

Description

Plot the metagraph of the parameter of the stochastic block model associated with one of the estimated graph clusters

Usage

```
metagraph(nb, res, title = NULL, edge.width.cst = 10)
```

Arguments

nb number of the cluster we are interested in

res output of graphClustering()

title title of the figure

edge.width.cst width of edges in the metagraph

Value

none

Examples

```
theta <- list(pi=c(.5,.5), gamma=matrix((1:4)/8,2,2))
obs <- rCollectSBM(rep(10,4), theta)$listGraphs
res <- graphClustering(obs, nbCores=2)
metagraph(1, res)</pre>
```

moments

Computation of graph moments of a network

Description

Computation of graph moments of a network

Usage

```
moments(A, k = 3)
```

Arguments

A adjacency matrix

k order of the largest graph moments to be considered

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Value

vector with the first k (normalized) graph moments of the network A

Examples

```
param <- list(R = 500, alpha = .04, beta = .02, deltaIn = 100, deltaOut = 100) A <- sampleDPA(param) moments(A)
```

permutParam

Permute block labels of a stochastic block model parameter

Description

Permute block labels of a stochastic block model parameter

Usage

```
permutParam(theta, permut)
```

Arguments

theta a SBM parameter with say K blocks
permut a permutation of the block labels 1,2,...,K

Value

stochastic block model parameter with permuted block labels

```
theta1 <- list(pi=c(.5,.5), gamma=matrix((1:4)/8,2,2)) theta2 <- list(pi=c(.5,.5), gamma=matrix(4:1/8,2,2)) permutParam(theta1, 2:1) permutParam(theta2, 2:1)
```

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plotDendrogram	Plot dendrogram to visualize the clustering obtained by the hierarchical clustering algorithm

Description

Plot dendrogram to visualize the clustering obtained by the hierarchical clustering algorithm

Usage

```
plotDendrogram(res, labels = NULL, labcex = 0.5)
```

Arguments

res output of graphClustering()

labels network labels, default (NULL) network number.

labcex size of labels in the figure

Value

dendrogram

Examples

```
theta <- list(pi=c(.5,.5), gamma=matrix((1:4)/8,2,2))
obs <- rCollectSBM(rep(10,4), theta)$listGraphs
res <- graphClustering(obs, nbCores=2)
plotDendrogram(res)</pre>
```

rCollectSBM

Simulate a sample of networks of a stochastic block model

Description

Simulate a sample of networks of a stochastic block model

Usage

```
rCollectSBM(vec_n, theta, directed = TRUE)
```

Arguments

vec_n vector with number of vertices

theta stochastic block model parameter with latent group probabilities \$pi and con-

nectivy parameters \$gamma

directed directed networks (TRUE by default) or undirected (FALSE)

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Value

list with a list of adjacency matrices (\$listGraphs) and a list of node labels (\$listLatentZ)

Examples

```
theta1 <- list(pi=c(.5,.5), gamma=matrix((1:4)/8,2,2)) rCollectSBM(2:4, theta1)
```

rMixSBM

Simulate a collection of networks of a mixture of stochastic block mod-

Description

Simulate a collection of networks of a mixture of stochastic block models

Usage

```
rMixSBM(vec_n, thetaMixSBM, directed = TRUE)
```

Arguments

vec_n vector with number of vertices

thetaMixSBM K-list for a mixture with K components. Each field is a list with the stochastic

block model parameter (\$pi and \$gamma) and a cluster proportion (\$prop)

directed directed networks (TRUE by default) or undirected (FALSE)

Value

list with a list of adjacency matrices (\$listGraphs), a list of node labels (\$listLatentZ) and a vector with the graph clustering (\$label)

```
theta1 <- list(prop=.2, pi=c(.5,.5), gamma=matrix((1:4)/8,2,2)) theta2 <- list(prop=.8, pi=c(.5,.5), gamma=matrix(4:1/8,2,2)) thetaMixSBM <- list(NULL) thetaMixSBM[[1]] <- theta1 thetaMixSBM[[2]] <- theta2 obs <- rMixSBM(vec_n=rep(10,3), thetaMixSBM)
```

rsbm 13

r	s	h	m

Simulate a network of a stochastic block model

Description

Simulate a network of a stochastic block model

Usage

```
rsbm(n, theta, directed = TRUE)
```

Arguments

n number of vertices

theta stochastic block model parameter with latent group probabilities \$pi and con-

nectivy parameters \$gamma

directed directed network (TRUE by default) or undirected (FALSE)

Value

list with simulated adjacency matrix (\$adj) and node labels (\$Z)

Examples

```
theta1 <- list(pi=c(.5,.5), gamma=matrix((1:4)/8,2,2)) rsbm(10, theta1)
```

sampleDPA

generation of a network of the directed preferential attachment (DPA) model

Description

generation of a network of the directed preferential attachment (DPA) model

Usage

```
sampleDPA(param)
```

Arguments

param

list with the following elements: R = 1 (= number of iterations), α , β

\$deltaIn, \$deltaOut (parameters of the DPA model)

Value

adjacency matrix of generated network

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Examples

```
param <- list(R = 500, alpha = .04, beta = .02, deltaIn = 100, deltaOut = 100) A <- sampleDPA(param) A
```

sampleDPAMixture

Generation of a mixture of directed preferential attachment (DPA) models

Description

Generation of a mixture of directed preferential attachment (DPA) models

Usage

```
sampleDPAMixture(M, param)
```

Arguments

М

number of desired networks

param

list of list of parameters of the DPA models. Each element of param is a list with the following elements: prop (weight of the mixture component), R (= number of iterations), heat , heat , heat , heat (parameters of the DPA model)

Value

list of 2 lists: the first (\$listAdj) is a list of M adjacency matrices, the second a list (\$graphGroups) contains the true cluster labels

```
param <- vector('list', 3)</pre>
param[[1]] <- list(prop = 1/3, # component 1 : alpha > beta
                    alpha = .04,
                    beta = .02,
                    deltaIn = 100,
                    deltaOut = 100,
                    R = 500
)
param[[2]] <- list(prop = 1/3, # component 2 : just permute alpha and beta ;</pre>
                    alpha = .01,
                    beta = .02,
                    deltaIn = 100,
                    deltaOut = .1,
                    R = 1000
)
param[[3]] <- list(prop = 1/3, # component 3 : alpha=beta</pre>
                    alpha = .015,
```

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sbmNorm

(squared) norm between two stochastic block models

Description

the norm is the minimal graphon distance between two stochastic block model parameters obtained with the best permutations of the parameters

Usage

```
sbmNorm(theta1, theta2)
```

Arguments

theta1 a stochastic block model parameter theta2 a stochastic block model parameter

Value

(squared) norm between two stochastic block models

```
theta1 <- list(pi=c(.5,.5), gamma=matrix((1:4)/8,2,2)) theta2 <- list(pi=c(.5,.5), gamma=matrix(4:1/8,2,2)) theta3 <- list(pi=c(.5,.5), gamma=matrix(1:4/4,2,2)) sbmNorm(theta1, theta2) sbmNorm(theta1, theta3) sbmNorm(theta2, theta3)
```

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