

# Package ‘graphclust’

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**Type** Package

**Title** Hierarchical Graph Clustering for a Collection of Networks

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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 graph clustering of a collection of networks using an agglomerative algorithm” by T. Rebafka (2022) <[arXiv:2211.02314](https://arxiv.org/abs/2211.02314)>.

**License** GPL-2

**Encoding** UTF-8

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ARI	<i>Adjusted Rand index</i>
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### 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)

### Examples

```
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)
```

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degreeSort	<i>Sort stochastic block model parameter in a unique way using its graphon</i>
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---

### 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)
```

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fitSBMcollection	<i>Fit a unique stochastic block model to a collection of networks</i>
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### 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.

**Usage**

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

**Arguments**

allAdj            list of adjacency matrices  
hyperParam        hyperparameters of prior distributions  
nbCores           number of cores for parallelization. Default: detectCores().

**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)
```

---

fitSimpleSBM	<i>Fit a stochastic block model to every network in a collection of networks.</i>
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---

**Description**

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

**Usage**

```
fitSimpleSBM(
  allAdj,
  directed = TRUE,
  nbCores = detectCores(),
  outCountStat = TRUE
)
```

**Arguments**

allAdj	list of adjacency matrices
directed	Networks are directed (TRUE by default) or undirected (FALSE).
nbCores	number of cores for parallelization. Default: detectCores().
outCountStat	If TRUE (default), the output is a list of count statistics for every network. If 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)
```

---

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),
  nbCores = detectCores()
)
```

**Arguments**

allAdj	list of adjacency matrices
hyperParam	hyperparameters of prior distributions
nbCores	number of cores for parallelization. Default detectCores().

**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)
```

---

graphonL2norm	<i>(squared) L2-norm of the graphons associated with two stochastic block model parameters</i>
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---

**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	<i>Graph clustering using the pairwise graphon distances and spectral clustering</i>
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---

**Description**

Graph clustering using the pairwise graphon distances and spectral clustering

**Usage**

```
graphonSpectralClustering(allAdj, nbClusters, nbCores = detectCores())
```

**Arguments**

allAdj            list of adjacency matrices  
 nbClusters       number of clusters to be found  
 nbCores           number of cores for parallelization. Default: detectCores().

**Value**

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

**Examples**

```
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)
```

---

metagraph	<i>Plot the metagraph of the parameter of the stochastic block model associated with one of the estimated graph clusters</i>
-----------	--

---

**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)
```

---

permutParam	<i>Permute block labels of a stochastic block model parameter</i>
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---

**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

**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))
permutParam(theta1, 2:1)
permutParam(theta2, 2:1)
```

---

plotDendrogram	<i>Plot dendrogram to visualize the clustering obtained by the hierarchical clustering algorithm</i>
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---

**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)
```

---

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 connectivity parameters $\gamma$
directed	directed networks (TRUE by default) or undirected (FALSE)

**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	<i>Simulate a collection of networks of a mixture of stochastic block models</i>
---------	--

---

### 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 ( $\text{prop}$ )
directed	directed networks (TRUE by default) or undirected (FALSE)

### Value

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

### Examples

```
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	<i>Simulate a network of a stochastic block model</i>
------	---

---

### 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 connectivity 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)
```

---

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

**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))
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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