Package ‘mixer’

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Maintainer Pierre Latouche <pierre.latouche@univ-paris1.fr>
Description Routines for the analysis (unsupervised clustering) of networks using MIXtures of Erdos-Renyi random graphs
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Author Christophe Ambroise [aut],
Gilles Grasseau [aut],
Mark Hoebke [aut],
Pierre Latouche [aut, cre],
Vincent Miele [aut],
Franck Picard [aut],
LAPACK authors [aut] (copyrights apply to src/*.f)

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French Political Blogosphere network dataset consists of a single day snapshot of over 200 political blogs automatically extracted the 14 October 2006 and manually classified by the "Observatoire Presidentielle" project. This project is the result of a collaboration between RTGI SAS and Exalead and aims at analyzing the French presidential campaign on the web.

Usage

data(blog)

Format

196 political blogs described by their political origin and connections. There are 2 items in the list blog:

- **links**: adjacency matrix describing if there exist a hyperlink between two blogs.
- **politicalParty**: political group of the blogs.

Details

In this data set, nodes represent hostnames (a hostname contains a set of pages) and edges represent hyperlinks between different hostnames. If several links exist between two different hostnames, they were collapsed into a single one. Note that intra domain links can be considered if hostnames are not identical. Finally, in this experimentation we consider that edges are not oriented which is not realistic but which does not affect the interpretation of the groups. This network presents an interesting communities organization due to the existence of several political parties and commentators. We assume that authors of these blogs tend to link, by political affinities, blogs with similar political positions. Six known communities compose this network: Gauche ("French democrat"), Divers Centre (Moderate party), Droite (French republican), Ecologiste (green), Liberal (supporters of economic-liberalism) and finally Analysts.

References

http://www.observatoire-presidentielle.fr/

Examples

data(blog)
mixer(x=blog$links,qmin=2,qmax=12)->xout
## Not run: plot(xout)
getModel

Get the parameters of a model

Description

Generic method to get the parameters of a model. Type `getModel.mixer` for more details.

Usage

`getModel( object, ...)`

Arguments

- `object` an object representing a model
- `...` adding arguments (depending on the object type)

Value

Return parameters of the model.

Examples

```r
graph.affiliation(n=100,c(1/3,1/3,1/3),0.8,0.2) -> g
mixer(g$x,qmin=2,qmax=6) -> xout
m <- getModel( xout )
m <- getModel( xout, q=5 )
```

getModel.mixer

Get the model parameters

Description

Given a number of classes, get the model parameters (`alphas`, `pis`, `taus`).

Usage

```r
## S3 method for class 'mixer'
getModel( object, ...)
```
Arguments

object

... adding arguments (depending on the object type)

q selects a q-class model. If NULL, q value is chosen to maximise the criterion (ICL criterion or ILvb criterion) (see plot.mixer and mixer). Default NULL.

Value

Return a list with the following attributes:

q q-class model selected.
criterion criterion value (ICL criterion or ILvb criterion).
alphas vector of class proportion.
Pis connectivity matrix of classes.
Taus matrix of posterior probabilities (of the hidden colour knowing the graph structure).

Author(s)

G. Grasseau

Examples

```r
graph.affiliation(n=100,c(1/3,1/3,1/3),0.8,0.2) -> g
mixer(g$x,qmin=2,qmax=6) -> xout
m <- getModel( xout )
m <- getModel( xout, q=5 )
```

Description

Simulate an affiliation network with a given number of clusters, specific class proportions and within/between connection probabilities.

Usage

```r
graph.affiliation( n=100, alphaVect=c(1/2,1/2), lambda=0.7, epsilon=0.05, directed=FALSE )
```
Arguments

- n: number of nodes of the simulated network.
- alphaVect: vector of cluster proportions.
- lambda: within-cluster probability of edge.
- epsilon: between-clusters probability of edge.
- directed: TRUE for directed graphs.

Details

graph.affiliation simulates a simple Erdös-Rényi Mixture of Graph model, using the same within-cluster edge probability for all clusters and a unique between-cluster edge probability.

Value

graph.affiliation returns a list of 2 objects:

- x: an adjacency matrix of size n by n,
- cluster: a vector of integers indicating the cluster to which each node is allocated.

Author(s)

Christophe Ambroise

Examples

```r
g = graph.affiliation(n=100,c(1/3,1/3,1/3),0.8,0.2)
str(g)
```

---

macaque

Connection of macaque brain cortical regions

Description

The dataset consists in 47 brain cortical regions connected by 505 inter-regional pathways in the Macaque Cortex.

Usage

data(macaque)

Format

A data frame describing the adjacency matrix of the connection of the 47 brain cortical regions of the Macaque Cortex.
Details

As brain function is based on inter-regional connections, studying the way cortical regions interact may offer new perspectives in the comprehension of information flows within the brain. It appears that particular brain regions may play different roles: some regions can be at the "center" of a particular part of the network, meaning that a lot of information will pass through them, whereas other parts of the network may be more "peripherica". Consequently, identifying central zones would be important, as their lesion may compromise the integrity of the whole network. From a topological view, finding those "hubs" as focused much attention, with a definition based on degree only. However, there exists many ways for a node to be a hub, and degree is one criteria. As there is no definition of what a hub is, there are many different hubs (provincial and central). This is why [1] developed a multi-criteria strategy to find nodes that can be called "hubs". From a methodological point of view, this approach seems to be limited as the resuting hubs will be criteria-dependent. The gain of Mixer is that the model can be used to find those hubs. Indeed, using the underlying missing data framework, MixNet will find nodes that connect heavily to other nodes in the network, and that share this connectivity pattern (a class of hubs for instance).

References


Examples

data(macaque)
mixer(macaque,qmin=8)->xout
## Not run: plot(xout)

mixer

MIXtures of Erdos-Renyi random graphs

Description

Estimate the parameters, the hidden class variables, as well as the number of classes q of a MIXture of Erdős Rényi random graphs. The estimation is performed for binary graphs (edges are assumed to be drawn from Bernoulli distributions).

Usage

mixer( x, qmin=2, qmax=NULL, method="variational", directed = NULL, nbiter=10, fpnbiter=5, improve=FALSE, verbose=TRUE)

Arguments

x an adjacency matrix or a matrix of edges (each column gives the two node indexes defining an edge) or a spm file name (a .spm file describes the network as a sparse matrix).
qmin minimum number of classes.
mixer

qmax  maximum number of classes (if NULL, only q=qmin is considered).
method strategy used for the estimation: "variational", "classification", or "bayesian"
directed TRUE/FALSE for directed/undirected graph. Default is NULL, i.e. according to the input array x, mixer identifies whether the graph is directed or undirected.
nbiter maximum number of EM iterations (default: 10).
fpnbiter maximum number of internal iterations for the E step (default: 5).
improve selects between improved or basic strategies (default: FALSE).
verbose display warning messages (default: TRUE).

Details
mixer implements inference methods for the MixNet model (sometimes referred to as Erdős-Rényi mixture model for graphs) which is described in Daudin et. al (2008). Please note that the MixNet model is a special case of binary stochastic block models (Nowicki and Snijders, 2001). The inference allows to uncover clusters of vertices sharing homogeneous connection profiles. In particular, the package can be used to look for specific clusters, namely communities, where nodes of a community are more likely to connect to nodes of the same community.

MixNet must not be confused with Exponential Random Graph Models for network data (ERGM). The mixer package implements three different estimation strategies which were developed to deal with directed and undirected graphs:

variational refers to the paper of Daudin et. al (2008). It is the default method.

classification implements the method described in Zanghi et. al (2008). This method is faster than the variational approach and is able to deal with bigger networks but can produce biased estimates.

bayesian implements the method described in Latouche et. al (2012).

The implementation of the two first methods consists of an R wrapper of the c++ software package mixnet developed by Vincent Miele (2006).

The mixer routine uses the estimation strategy described in method and computes a model selection criterion for each value of q (the number of classes) between qmin and qmax. The ICL criterion is used for the variational and classification methods. It corresponds to an asymptotic approximation of the Integrated Classification Likelihood. The other criterion, so called ILvb (Integrated Likelihood variational Bayes), is used for the bayesian method. It is based on a variational (non-asymptotic) approximation of the Integrated observed Likelihood.

mixer is an user-friendly package with a reduced number of functions. For R-developers in statistical networks a more complete set, called mixer-dev, is provided (see below).

Value
mixer returns an object of class mixer. Below the main attributes of this class:

nnodes number of connected nodes.
map mapping from connected nodes to the whole set of nodes.
edges edge list.
mixer

qmin, qmax  number of classes.
output  output list of qmax-qmin+1 items. Each item contains the result of the estimation for a given number of class q. Details of output field:
output[[i]]$criterion  ICL criterion or ILvb criterion used for model selection (see details section for more).
output[[i]]$alphas  vector of proportion, whose length is the number of component.
output[[i]]$pis  class connectivity matrix.
output[[i]]$taus  matrix of posterior probabilities (of the hidden color knowing the graph structure).

Author(s)

Christophe Ambroise, Gilles Grasseau, Mark Hoebeke, Pierre Latouche, Vincent Miele, Franck Picard

References

Hugo Zanghi, Christophe Ambroise and Vincent Miele (2008), Fast online graph clustering via Erdős-Rényi mixture. Pattern Recognition, 41, 3592-3599.
mixer-dev tool: see http://ssbgroup.fr/mixnet/mixer.html

Examples

```r
g <- graph.affiliation(n=100,c(1/3,1/3,1/3),0.8,0.2)
mixer(g,x,qmin=2,qmax=6)->xout
## Not run: plot(xout)
```

```r
g <- graph.affiliation(n=50,c(1/3,1/3,1/3),0.8,0.2)
mixer(g,x,qmin=2,qmax=5, method="bayesian")->xout
## Not run: plot(xout)
```
## data(blog)

## set the seed to replicate results
setSeed(777)

## Not run: plot(xout)

## get best run
m <- getModel(xout)

## get run for q=5
m <- getModel(xout, q=5)

---

### plot.mixer

**Plot of mixer object**

#### Description

`plot.mixer` can display five kinds of figure: model selection criterion curve, the adjacency matrix map, the degree distribution histogram, the connectivity matrix graph and the adjacency matrix graph. By default the four first plots are displayed.

#### Usage

```r
## S3 method for class 'mixer'
plot(x, q=NULL, frame=1:4, classes=NULL, classes.col=NULL, quantile.val=0.1, ...)
```

#### Arguments

- **x**: a mixer object (output of the mixer function).
- **q**: the q-class model to display. By default, the q is set to the value which maximizes the criterion (see frame 1).
- **frame**: a vector of frame numbers to display (5 kinds of plots, see details section for more).
- **classes**: an external classification used for frame 4 (pie chart): vector as factor of node elements (the number of external class levels corresponds to the number of levels).
- **classes.col**: a vector of user colors used to identify the classes.
quantile.val filters the connectivity matrix values (Pis) in frame 4. Display the upper part (specified by quantile.val) of the distribution.

... further graphical arguments.

Details

Frame values:

1 criterion (ICL or ILvb) versus the number of classes (see mixer).
2 adjacency matrix reorganized according to the estimated partition for a given number of classes q.
3 degree distribution (histogram) and theoretical degree distribution (blue curve) computed from the q-class model parameters (alphas, Pis).
4 matrix connectivity between classes (Pis) given a number of classes q. The thickest edges identify the highest values of the connectivity probabilities and the largest nodes point out the most populated classes.

Providing external classes (see classes argument) each node displays a pie chart pointing out the classification relevance.
5 graph display of the adjacency matrix.

Author(s)

C. Ambroise, G. Grasseau

See Also

mixer, getModel

Examples

# Simple example: display the 4 frames for the best class number estimation
#
g <- graph.affiliation(n=100, c(1/3, 1/3, 1/3), 0.8, 0.2)
xout <- mixer(g$x, qmin=2, qmax=6)
## Not run: plot(xout)

# Display the same for 4 classes with no filtering
#
## Not run: plot(xout, q=4, quantile.val=0)

# Display a pie chart for 4 classes
#
data(blog)
xout <- mixer(x=blog$links, qmin=2, qmax=12)
# Unconnected nodes have been removed by mixer.
# xout$map contains the mapping from connected nodes to the whole set
ext.classes <- blog$politicalParty
## Not run: plot( xout, frame=4, classes=ext.classes )
**setSeed**

*Set internal seed*

**Description**

This utility sets the internal random seed used by mixer.

**Usage**

```
setSeed( seed = 1 )
```

**Arguments**

- **seed**: sets the seed of the internal random generator in the mixer C/C++ libraries (integer value).

**Details**

Sets the seed of the random generator (`srand`) in the C standard library. This random generator is used inside the mixer initialization stage. This function is useful to generate exactly the same initial conditions before two mixer runs.

**Author(s)**

G. Grasseau

**Examples**

```r
graph.affiliation(n=100,c(1/3,1/3,1/3),0.8,0.2)->g
setSeed(777)
mixer(g$x,qmin=2,qmax=6)->xout
## Not run: plot(xout)

# Produce strictly the same result
setSeed(777)
mixer(g$x,qmin=2,qmax=6)->xout
## Not run: plot(xout)
```
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