# Package 'MBCbook'

May 8, 2024

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# **Description**

The companion package provides all original data sets and functions that are used in the book "Model-Based Clustering and Classification for Data Science" by Charles Bouveyron, Gilles Celeux, T. Brendan Murphy and Adrian E. Raftery (2019, ISBN:9781108644181).

## **Details**

# The DESCRIPTION file:

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License: GPL (>= 2)

NeedsCompilation: no

URL: https://github.com/cbouveyron/MBCbook

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BugReports: https://github.com/cbouveyron/MBCbook/issues

Author: Charles Bouveyron [cre, aut], Gilles Celeux [aut], T. Brendan Murphy [aut], Adrian Raftery [aut]

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#### Author(s)

Charles Bouveyron [cre, aut], Gilles Celeux [aut], T. Brendan Murphy [aut], Adrian Raftery [aut]

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# References

Charles Bouveyron and Gilles Celeux and T. Brendan Murphy and Adrian E. Raftery, Model-Based Clustering and Classification for Data Science: with Applications in R, Cambridge University Press, 2019.

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Advice

The Advice data set from Lazega (2001)

## **Description**

Lazega (2001) <doi:10.2307/3556688> collected a network data set detailing interactions between a set of 71 lawyers in a corporate law firm in the USA. The data include measurements of the advice network, friendship network and co-worker network between the lawyers within the firm. Further covariates associated with each lawyer in the firm are also available including age, seniority, college education and office location.

### Usage

```
data("Advice")
```

#### **Format**

A large network object, which can be managed with the network library, with 71 nodes.

## References

Lazega, E., The Collegial Phenomenon: The Social Mechanisms of Cooperation Among Peers in a Corporate Law Partnership, Oxford University Press, 2001 <doi:10.2307/3556688>.

# Examples

data(Advice)

AIDSBlogs

The AIDSBlogs data set

#### **Description**

The AIDS blog data set records the pattern of citation among 146 unique blogs related to AIDS patients and their support networks. The data were originally collected by Gopal (2007) <doi:10.1007/1-4020-5427-0\_18> over a randomly selected three-day period in August 2005. The nodes in the network correspond to blogs and a directed edge from one blog to another indicates that the former had a link to the latter in their web page.

## Usage

```
data("AIDSBlogs")
```

# Format

A large network object, which can be managed with the network library, with 146 nodes.

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## References

Gopal, S., The evolving social geography of blogs, in Miller, H. J. (ed.), Societies and Cities in the Age of Instant Access, The GeoJournal Library, vol. 88., pp. 275–293, 2007 <doi:10.1007/1-4020-5427-0\_18>.

### **Examples**

data(AIDSBlogs)

amazonFineFoods

The Amazon Fine Foods data set

## **Description**

The Amazon Fine Foods data set has 1646 rows and 1735 columns, describing whether an user (row) has noted and reviewed a product (column) or not.

## Usage

```
data("amazonFineFoods")
```

## **Format**

A data frame with binary values indicating whether an user (row) has noted and reviewed a product (column) or not.

## Source

https://snap.stanford.edu/data/web-FineFoods.html.

# **Examples**

data(amazonFineFoods)

constrEM

Semi-supervised clustering with must-link constraints

# **Description**

Semi-supervised clustering with must-link constraints allows to cluster data for which must-link constraints are available. This function implements the method described in Shental et al. (2003, ISBN:9781615679119).

# Usage

```
constrEM(X, K, C, maxit = 30)
```

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# **Arguments**

Κ

Χ	a data frame of observations, assuming the rows are the observations and the
	columns the variables. Note that NAs are not allowed.

the number of desired groups.

C a vector encoding the must-link constraints through chuncklets. This vector has

to be of the length of the number of observations. Two observations that have to be in the same group must be in the same chuncklet. For instance, the chuncklet vector (1,2,3,4,3,5) indicate that 3rd and the 5th observations have a must-link constraint. If there is no must-link constraints, this vector should be simply

1:nrow(X).

maxit the maximum number of iterations.

#### Value

A list is returned with the following fields:

cls a vector containg the group memberships of the observations.

T the posterior probabilities that the observations belong to the K groups.

prop the estimated mixture proportions.

mu the estimated mixture means.

S the estimated mixture covariance matrices.

the log-likelihood value at convergence.

## Author(s)

C. Bouveyron

# References

This function implements the method described in Shental, N., Bar-Hillel, A., Hertz, T., and Weinshall, D., Computing Gaussian mixture models with EM using equivalence constraints, Proceedings of the 16th International Conference on Neural Information Processing Systems, pages 465–472, 2003 (ISBN:9781615679119).

# **Examples**

```
# Simulation of some data set.seed(123)  
n = 200  
m1 = c(0,0); m2 = 4*c(1,1); m3 = 4*c(1,1)  
S1 = diag(2); S2 = rbind(c(1,0),c(0,0.05))  
S3 = rbind(c(0.05,0),c(0,1))  
X = rbind(mvrnorm(n,m1,S1),mvrnorm(n,m2,S2),mvrnorm(n,m3,S3))  
cls = rep(1:3,c(n,n,n))  
# Encoding the constraints through chunklets  
# Observations 397 and 408 are in the same chunklet  
a = 398
```

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```
b = 430
C = c(1:(b-1),a,b:(nrow(X)-1))
# Clustering with constrEM
res = constrEM(X,K=3,C,maxit=20)
```

Coworker

The Coworker data set from Lazega (2001)

## **Description**

Lazega (2001) <doi:10.2307/3556688> collected a network data set detailing interactions between a set of 71 lawyers in a corporate law firm in the USA. The data include measurements of the advice network, friendship network and co-worker network between the lawyers within the firm. Further covariates associated with each lawyer in the firm are also available including age, seniority, college education and office location.

## Usage

```
data("Coworker")
```

#### **Format**

A large network object, which can be managed with the network library, with 71 nodes.

#### References

Lazega, E., The Collegial Phenomenon: The Social Mechanisms of Cooperation Among Peers in a Corporate Law Partnership, Oxford University Press, 2001 <doi:10.2307/3556688>.

# **Examples**

```
data(Coworker)
```

credit

The Credit data set

# **Description**

The Credit data set has 66 rows and 11 columns, describing customers who took out loans from a credit company described with 11 categorical or ordinal variables.

# Usage

```
data("credit")
```

8 denoisePatches

## **Format**

A data frame with 66 observations and 11 categorical or ordinal variables.

#### Source

https://husson.github.io/data.html

# **Examples**

```
data(credit)
```

denoisePatches

Denoising of image patches

# Description

Denoising of image patches based on the clustering of patches.

# Usage

```
denoisePatches(Y,out,P,sigma=10)
```

# **Arguments**

Y a data frame containing as rows the image patches to denoise out the mixmodCluster object that contains mixture parameters
P the posterior probabilities that patches belong to the clusters sigma the noise standard deviation

# Value

A data fame of the denoised patches is returned.

#### Note

C. Bouveyron & J. Delon

# **Examples**

```
Im = diag(16)
ImNoise = Im + rnorm(256,0,0.1)
X = imageToPatch(ImNoise,4)
out = mixmodCluster(X,10,model=mixmodGaussianModel(family=c("spherical")))
res = mixmodPredict(X,out@bestResult)
Xdenoised = denoisePatches(X,out,P = res@proba,sigma = 0.1)
ImRec = reconstructImage(Xdenoised,16,16)
oldpar <- par(no.readonly = TRUE)
par(mfrow=c(1,3))</pre>
```

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```
imshow(Im); imshow(ImNoise); imshow(ImRec)
par(oldpar)
```

Friend

The Friend data set from Lazega (2001)

## **Description**

Lazega (2001) <doi:10.2307/3556688> collected a network data set detailing interactions between a set of 71 lawyers in a corporate law firm in the USA. The data include measurements of the advice network, friendship network and co-worker network between the lawyers within the firm. Further covariates associated with each lawyer in the firm are also available including age, seniority, college education and office location.

# Usage

```
data("Friend")
```

#### **Format**

A large network object, which can be managed with the network library, with 71 nodes.

#### References

Lazega, E., The Collegial Phenomenon: The Social Mechanisms of Cooperation Among Peers in a Corporate Law Partnership, Oxford University Press, 2001 <doi:10.2307/3556688>.

# **Examples**

```
data(Friend)
```

imageToPatch

Transform an image into a collection of patches

# Description

Transform an image into a collection of small images (patches) that cover the original image.

# Usage

```
imageToPatch(Im,f)
```

# Arguments

Im the image for which one wants to extract local patches.

f the size of the desired patches (fxf).

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## Value

A data frame of all extracted patches is returned.

## Author(s)

C. Bouveyron & J. Delon

# **Examples**

```
Im = diag(16)
ImNoise = Im + rnorm(256,0,0.1)
X = imageToPatch(ImNoise,4)
out = mixmodCluster(X,10,model=mixmodGaussianModel(family=c("spherical")))
res = mixmodPredict(X,out@bestResult)
Xdenoised = denoisePatches(X,out,P = res@proba,sigma = 0.1)
ImRec = reconstructImage(Xdenoised,16,16)
oldpar <- par(no.readonly = TRUE)
par(mfrow=c(1,3))
imshow(Im); imshow(ImNoise); imshow(ImRec)
par(oldpar)</pre>
```

imshow

Display an image

## **Description**

A simple way of displaying an image, using the image function.

# Usage

```
imshow(x,col=palette(gray(0:255/255)),useRaster = TRUE,...)
```

## **Arguments**

x the image to display as a matrix.

col the color palette to use when displaying the image.

useRaster logical; if TRUE a bitmap raster is used to plot the image instead of polygons.

The grid must be regular in that case, otherwise an error is raised. For the behaviour when this is not specified, see the 'Details' section of the image func-

tion.

... additionial arguments to provide to subfunctions.

#### Value

This function returns nothing.

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## See Also

image

# **Examples**

```
Im = diag(16)
imshow(Im)
```

NIR

The chemometrics near-infrared (NIR) data set

# Description

The chemometrics near-infrared (NIR) data set has 202 observations and 2801 variables: 2800 near-infrared wavelength measures and 1 class variable. The data were obtained from the analysis of three types of textiles. The data set was first introduce in Devos et al. (2009) <doi:10.1016/j.chemolab.2008.11.005>.

## Usage

```
data("velibCount")
```

#### **Format**

A data frame with 202 observations and 2801 variables. The first variable indicates the class-memberships of the observations.

# References

Devos, O., Ruckebusch, C., Durand, A., Duponchel, L., and Huvenne, J.-P., Support vector machines (SVM) in near infrared (NIR) spectroscopy: Focus on parameters optimization and model interpretation, Chemometrics and Intelligent Laboratory Systems, 96, 27–33, 2009 <doi:10.1016/j.chemolab.2008.11.005>.

# **Examples**

```
data(NIR)
matplot(t(NIR[,-1]),type='l',col=NIR[,1])
```

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PoliticalBlogs

The political blog data set

# Description

The political blog data set shows the linking structure in online blogs which commentate on French political issues; the data were collected by Observatoire Presidentielle in October 2006. The data were first used by Latouche et al. (2011) <doi:10.1214/10-AOAS382>.

### Usage

```
data("PoliticalBlogs")
```

#### **Format**

A large network object, which can be managed with the network library, with 196 nodes.

#### References

P. Latouche, E. Birmelé, and C. Ambroise. "Overlapping stochastic block models with application to the French political blogosphere". In: Annals of Applied Statistics 5.1, p. 309-336, 2011 <doi:10.1214/10-AOAS382>.

## **Examples**

```
data(PoliticalBlogs)

# Visualization with the network library
library(network)
plot(PoliticalBlogs)
```

puffin

The puffin data set

# **Description**

The puffin data set contains 69 individuals (birds) described by 5 categorical variables, in addition to class labels.

# Usage

```
data("puffin")
```

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## **Format**

A data frame with 69 observations and 6 variables.

```
class the class of the observations gender gender of the bird eyebrow gender of the bird collar gender of the bird sub.caudal gender of the bird border gender of the bird
```

#### **Source**

The data were provided by Bretagnolle, V., Museum d'Histoire Naturelle, Paris.

# **Examples**

```
data(puffin)
```

reconstructImage

Reconstructing an image from a patch decomposition

# Description

A simple way of reconstructing an image from a patch decomposition.

# Usage

```
reconstructImage(X,nl,nc)
```

# **Arguments**

X the matrix of patches to be used for reconstructing the image.

nl the number of rows of the image.

nc the number of columns of the image.

# Value

an image is returned as a matrix object, that can be display with the imshow function.

# Author(s)

C. Bouveyron & J. Delon

14 rqda

## **Examples**

```
Im = diag(16)
ImNoise = Im + rnorm(256,0,0.1)
X = imageToPatch(ImNoise,4)
out = mixmodCluster(X,10,model=mixmodGaussianModel(family=c("spherical")))
res = mixmodPredict(X,out@bestResult)
Xdenoised = denoisePatches(X,out,P = res@proba,sigma = 0.1)
ImRec = reconstructImage(Xdenoised,16,16)
oldpar <- par(no.readonly = TRUE)
par(mfrow=c(1,3))
imshow(Im); imshow(ImNoise); imshow(ImRec)
par(oldpar)</pre>
```

rgda

Robust (quadratic) discriminant analysis

#### **Description**

Robust (quadratic) discriminant analysis implements a discriminant analysis method which is robust to label noise. This function implements the method described in Lawrence and Scholkopf (2003, ISBN:1-55860-778-1).

### Usage

```
rqda(X,lbl,Y,maxit=50,disp=FALSE,...)
```

# Arguments

X a data frame containing the learning observations.

1b1 the class labels of the learning observations.

Y a data frame containing the new observations to classify.

maxit the maximum number of iterations.

disp logical, if TRUE, several plots are displayed.

... additional arguments to provide to subfunctions.

## Value

A list is returned with the following elements:

the estimated class proportions. nu the estimated class means. mu the estimated covariance matrices. S the estimated purity level of the labels. gamma Τi the posterior probabilties of the labels knowing the observed labels for the learning observations. Ρi the class posterior probabilities of the observations to classify. the class assignments of the observations to classify. cls 11 the log-likelihood value.

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## Author(s)

C. Bouveyron

# References

Lawrence, N., and Scholkopf, B., Estimating a kernel Fisher discriminant in the presence of label noise, Pages 306–313 of: Proceedings of the Eighteenth International Conference on Machine Learning. ICML'01. San Francisco, CA, USA, 2001 (ISBN:1-55860-778-1).

## **Examples**

```
n = 50
m1 = c(0,0); m2 = 1.5*c(1,-1)
S1 = 0.1*diag(2); S2 = 0.25 * diag(2)
X = rbind(mvrnorm(n,m1,S1),mvrnorm(2*n,m2,S2))
cls = rep(1:2,c(n,2*n))

# Label perturbation
ind = rbinom(3*n,1,0.4); lb = cls
lb[ind==1 & cls==1] = 2
lb[ind==1 & cls==2] = 1

# Classification with RQDA
res = rqda(X,lb,X)
table(cls,res$cls)
```

**UScongress** 

The US congress vote data set

# **Description**

The US congress vote data set contains the votes (yes, no, abstained or absent) of 434 members of the 98th US Congress on 16 different key issues. This data set involves three-level categorical data.

# Usage

```
data("UScongress")
```

#### **Format**

A data frame with 434 observations on 16 different key issues. The first variables indicates the political party of the congressmen.

#### Source

http://archive.ics.uci.edu/ml/datasets/Congressional+Voting+Records

## **Examples**

```
data(UScongress)
```

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usps358

The handwritten digits usps358 data set

# **Description**

The handwritten digits usps358 data set is a subset of the famous USPS data from UCI, which contains only the 1 756 images of the digits 3, 5 and 8.

### Usage

```
data("usps358")
```

#### **Format**

A data frame with 1756 observations on the following 257 variables: cls is a numeric vector encoding the class of the digits, V1 to V256 are numeric vectors corresponding to the pixels of the 8x8 images.

#### **Source**

The data set is a subset of the famous USPS data from UCI (https://archive.ics.uci.edu/ml/index.php). The usps358 data set contains only the 1 756 images of the digits 3, 5 and 8 which are the most difficult digits to discriminate.

## **Examples**

data(usps358)

varSelEM

A variable selection algorithm for clustering

# Description

A variable selection algorithm for clustering which implements the method described in Law et al. (2004) <doi:10.1109/TPAMI.2004.71>.

## Usage

```
varSelEM(X,G,maxit=100,eps=1e-6)
```

# **Arguments**

Χ	a data frame containing the observations to cluster.

G the expected number of groups (integer).

maxit the maximum number of iterations (integer). The default value is 100.

eps the convergence threshold. The default value is 1e-6.

velib2D

#### Value

A list is returned with the following elements:

mu the group means for relevant variables.
sigma the group variances for relevant variables.
lambda the group means for irrelevant variables
alpha the group variances for irrelevant variables.

rho the feature saliency.

P the group posterior probabilities.

cls the group memberships.the log-likelihood value.

## Author(s)

C. Bouveyron

#### References

Law, M. H., Figueiredo, M. A. T., and Jain, A. K., Simultaneous feature selection and clustering using mixture models, IEEE Transactions on Pattern Analysis and Machine Intelligence, vol. 26, pp. 1154–1166, 2004 <doi:10.1109/TPAMI.2004.71>.

# **Examples**

```
data(wine27)
X = scale(wine27[,1:27])
cls = wine27$Type

# Clustering and variable selection with VarSelEM
res = varSelEM(X,G=3)

# Clustering table
table(cls,res$cls)
```

velib2D

The bivariate Vélib data set

# **Description**

The bivariate Vélib data set contains data from the bike sharing system of Paris, called Vélib. The data are loading profiles and percentage of broken docks of the bike stations over one week. The data were collected every hour during the period Sunday 1st Sept. - Sunday 7th Sept., 2014. The data were first used in Bouveyron et al. (2015) <doi:10.1214/15-AOAS861>.

# Usage

```
data("velib2D")
```

18 velibCount

#### **Format**

The format is:

- availableBikes: the loading profiles (nb of available bikes / nb of bike docks) of the 1189 stations at 181 time points.

- brokenDockss: the percentage of broken docks of the 1189 stations at 181 time points.
- position: the longitude and latitude of the 1189 bike stations.
- dates: the download dates.
- bonus: indicates if the station is on a hill (bonus = 1).
- names: the names of the stations.

#### **Source**

The real time data are available at https://developer.jcdecaux.com/ (with an api key).

#### References

The data were first used in C. Bouveyron, E. Côme and J. Jacques, The discriminative functional mixture model for the analysis of bike sharing systems, The Annals of Applied Statistics, vol. 9 (4), pp. 1726-1760, 2015 <doi:10.1214/15-AOAS861>.

## **Examples**

data(velib2D)

velibCount

The discrete version (count data) of the Vélib data set

# **Description**

The discrete version (count data) of Vélib data set contains data from the bike sharing system of Paris, called Vélib. The data consist in the number of bikes at stations over one week. The data were collected every hour during the period Sunday 1st Sept. - Sunday 7th Sept., 2014. The data were first used in Bouveyron et al. (2015) <doi:10.1214/15-AOAS861>.

# Usage

```
data("velibCount")
```

## **Format**

The format is:

- data: the nb of available bikes of the 1189 stations at 181 time points.
- position: the longitude and latitude of the 1189 bike stations.
- dates: the download dates.
- bonus: indicates if the station is on a hill (bonus = 1).
- names: the names of the stations.

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## Source

The real time data are available at https://developer.jcdecaux.com/ (with an api key).

#### References

The data were first used in C. Bouveyron, E. Côme and J. Jacques, The discriminative functional mixture model for the analysis of bike sharing systems, The Annals of Applied Statistics, vol. 9 (4), pp. 1726-1760, 2015 <doi:10.1214/15-AOAS861>.

# **Examples**

data(velib2D)

wine27

The (27-dimensional) Italian Wine data set

# **Description**

The (27-dimensional) Italian Wine data set is the result of a chemical analysis of 178 wines grown in the same region in Italy but derived from three different cultivars. The analysis determined the quantities of 27 constituents found in each of the three types of wines.

## Usage

```
data("wine27")
```

## Format

A data frame with 178 observations on the following 29 variables.

Alcohol a numeric vector

Sugar.free\_extract a numeric vector

Fixed\_acidity a numeric vector

Tartaric\_acid a numeric vector

Malic\_acid a numeric vector

Uronic\_acids a numeric vector

pH a numeric vector

Ash a numeric vector

Alcalinity\_of\_ash a numeric vector

Potassium a numeric vector

Calcium a numeric vector

Magnesium a numeric vector

Phosphate a numeric vector

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Chloride a numeric vector

Total\_phenols a numeric vector

Flavanoids a numeric vector

Nonflavanoid\_phenols a numeric vector

Proanthocyanins a numeric vector

Color\_Intensity a numeric vector

Hue a numeric vector

OD280.OD315\_of\_diluted\_wines a numeric vector

OD280.OD315\_of\_flavanoids a numeric vector

Glycerol a numeric vector

X2.3.butanediol a numeric vector

Total\_nitrogen a numeric vector

Proline a numeric vector

Methanol a numeric vector

Type a factor with levels Barbera, Barolo, Grignolino

Year a numeric vector

## **Details**

This data set is an expended version of the popular one from the UCI machine learning repository (http://archive.ics.uci.edu/ml/datasets/Wine).

# **Examples**

data(wine27)

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