

Package ‘MNARclust’

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Type Package

Title Clustering Data with Non-Ignorable Missingness using
Semi-Parametric Mixture Models

Version 1.1.0

Description Clustering of data under a non-ignorable missingness mechanism. Clustering is achieved by a semi-parametric mixture model and missingness is managed by using the pattern-mixture approach. More details of the approach are available in Du Roy de Chaumaray et al. (2020) <[doi:10.48550/arXiv.2009.07662](https://doi.org/10.48550/arXiv.2009.07662)>.

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License GPL (>= 2)

Imports Rcpp, parallel, sn, rmutl

LinkingTo Rcpp, RcppArmadillo

ByteCompile true

URL <https://arxiv.org/abs/2009.07662>

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LazyData true

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MNARclust-package	<i>MNARclust.</i>
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Description

Clustering method to analyze continuous or mixed-type data with missingness. The missingness mechanism can be non ignorable. The approach considers a semi-parametric mixture model.

Details

Package:	MNARclust
Type:	Package
Version:	1.1.0
Date:	2021-12-01
License:	GPL-3
LazyLoad:	yes

echo	<i>Echocardiogram data set</i>
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Description

All the patients suffered heart attacks at some point in the past. Some are still alive and some are not. The survival and still-alive variables, when taken together, indicate whether a patient survived for at least one year following the heart attack.

Format

A data frame with 132 observations on 13 variables (more details on this data set are presented in <http://archive.ics.uci.edu/ml/datasets/Echocardiogram>).

Details

This data set arise from the UCI machine learning repository (more details on this data set are presented <http://archive.ics.uci.edu/ml/datasets/Echocardiogram>)

References

Salzberg, S. (1988). Exemplar-based learning: Theory and implementation (Technical Report TR-10-88). Harvard University, Center for Research in Computing Technology, Aiken Computation Laboratory (33 Oxford Street; Cambridge, MA 02138).

Examples

```
data(echo)
```

MNARcluster	<i>Clustering function</i>
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Description

Clustering method to analyze continuous or mixed-type data with missingness. The missingness mechanism can be non ignorable. The approach considers a semi-parametric mixture model.

Usage

```
MNARcluster(  
  x,  
  K,  
  nbinit = 20,  
  nbCPU = 1,  
  tol = 0.01,  
  band = band.default(x),  
  seedvalue = 123  
)
```

Arguments

x	matrix used for clustering
K	number of components
nbinit	number of random starting points
nbCPU	number of CPU used for parallel computing (only Unix and Linux systems are allowed)
tol	stopping rule
band	bandwidth (numeric vector).
seedvalue	value of the seed (used to set the initializations of the MM algorithm)

Value

Returns a list containing the proportions (proportions), matrix of probabilities of missngness (rho), the posterior probabilities of classification (classproba), the partition (zhat) and the logarithme of the smoothed-likelihood (logSmoothlike)

References

Clustering Data with Non-Ignorable Missingness using Semi-Parametric Mixture Models, Marie Du Roy de Chaumaray and Matthieu Marbac <arXiv:2009.07662>.

Examples

```
set.seed(123)
# Data generation
ech <- rMNAR(n=100, K=2, d=4, delta=2, gamma=2)
# Clustering
res <- MNARcluster(ech$x, K=2)
# Confusion matrix between the estimated and the true partiion
table(res$zhat, ech$z)
```

rMNAR

Function used to simulate data from mixture model with specific missingness mechanism

Description

Generation of data set to perform the simulation presented in Section 4.1 of Du Roy de Chaumaray (2020)

Usage

```
rMNAR(
  n,
  K,
  d = 3,
  delta = 3,
  gamma = 1,
  law = "gauss",
  linkmissing = "logit-X"
)
```

Arguments

n	sample size (numeric of length 1)
K	number of clusters (numeric of length 1)
d	number of variables (numeric of length 1)
delta	tuning parameter to define the rate of misclassification (numeric of length 1)
gamma	tuning parameter to define the rate of missingness (numeric of length 1)
law	specifies the distribution of the variables within components (character that must be equal to gauss, student, laplace or skewgauss)
linkmissing	specify the missingness mechanism (character that must be equal to MCAR, logit-Z, logit-X or censoring)

Value

rMNAR returns a list containing the observed data (x), the true cluster membership (z), the complete data (x_{full}), the cluster membership given by the Baye's rule (z_{hat}), the empirical rates of misclassification (`meanerrorclass`) and missngness (`meanmiss`).

References

Clustering Data with Non-Ignorable Missingness using Semi-Parametric Mixture Models, Marie Du Roy de Chaumaray and Matthieu Marbac <[arXiv:2009.07662](https://arxiv.org/abs/2009.07662)>.

Examples

```
set.seed(123)
# Data generation
ech <- rMNAR(n=100, K=3, d=3, delta=2, gamma=1)
# Head of the observed data
head(ech$x)
# Table of the cluster memberships
table(ech$z)
# Empirical rate of misclassification
ech$meanerrorclass
# Empirical rate of missingness
ech$meanmiss
```

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