

Package: MixGHD (via r-universe)

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Type Package

Title Model Based Clustering, Classification and Discriminant Analysis
Using the Mixture of Generalized Hyperbolic Distributions

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Description Carries out model-based clustering, classification and
discriminant analysis using five different models. The models
are all based on the generalized hyperbolic distribution. The
first model 'MGHD' (Browne and McNicholas (2015)
<[doi:10.1002/cjs.11246](https://doi.org/10.1002/cjs.11246)>) is the classical mixture of
generalized hyperbolic distributions. The 'MGHFA' (Tortora et
al. (2016) <[doi:10.1007/s11634-015-0204-z](https://doi.org/10.1007/s11634-015-0204-z)>) is the mixture of
generalized hyperbolic factor analyzers for high dimensional
data sets. The 'MSGHD' is the mixture of multiple scaled
generalized hyperbolic distributions, the 'cMSGHD' is a 'MSGHD'
with convex contour plots and the 'MCGHD', mixture of coalesced
generalized hyperbolic distributions is a new more flexible
model (Tortora et al. (2019) <[doi:10.1007/s00357-019-09319-3](https://doi.org/10.1007/s00357-019-09319-3)>).
The paper related to the software can be found at
<[doi:10.18637/jss.v098.i03](https://doi.org/10.18637/jss.v098.i03)>.

Imports Bessel,stats, mvtnorm, ghyp, numDeriv, mixture, e1071,cluster,
methods

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 ARI

Adjusted Rand Index.

Description

Compares two classifications using the adjusted Rand index (ARI).

Usage

ARI(x=NULL, y=NULL)

Arguments

x A n dimensional vector of class labels.
 y A n dimensional vector of class labels. .

Details

The ARI has expected value 0 in case of random partition, it is equal to one in case of perfect agreement..

Value

The adjusted Rand index value

Author(s)

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References

L. Hubert and P. Arabie (1985) Comparing Partitions, *Journal of the Classification* 2:193-218.

Examples

```
##loading banknote data
data(banknote)

##model estimation
res=MGHD(data=banknote[,2:7], G=2 )

#result
ARI(res@map, banknote[,1])
```

banknote

Swiss Banknote data

Description

The data set contain 6 measures of 100 genuine and 100 counterfeit Swiss franc banknotes.

Usage

```
data(banknote)
```

Format

A data frame with the following variables:

Status the status of the banknote: genuine or counterfeit

Length Length of bill (mm)

Left Width of left edge (mm)

Right Width of right edge (mm)

Bottom Bottom margin width (mm)

Top Top margin width (mm)

Diagonal Length of diagonal (mm)

References

Flury, B. and Riedwyl, H. (1988). *Multivariate Statistics: A practical approach*. London: Chapman & Hall, Tables 1.1 and 1.2, pp. 5-8

bankruptcy	<i>Bankruptcy data</i>
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Description

The data set contain the ratio of retained earnings (RE) to total assets, and the ratio of earnings before interests and taxes (EBIT) to total assets of 66 American firms recorded in the form of ratios. Half of the selected firms had filed for bankruptcy.

Usage

```
data(bankruptcy)
```

Format

A data frame with the following variables:

Y the status of the firm: 0 bankruptcy or 1 financially sound.

RE ratio

EBIT ratio

References

Altman E.I. (1968). Financial ratios, discriminant analysis and the prediction of corporate bankruptcy. *J Finance* **23**(4): 589-609

cMSGHD	<i>Convex mixture of multiple scaled generalized hyperbolic distributions (cMSGHD).</i>
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Description

Carries out model-based clustering using the convex mixture of multiple scaled generalized hyperbolic distributions. The cMSGHD only allows convex level sets.

Usage

```
cMSGHD(data=NULL, gpar0=NULL, G=2, max.iter=100, label=NULL, eps=1e-2,
method="km", scale=TRUE, nr=10, modelSel="AIC")
```

Arguments

data	A n x p matrix or data frame such that rows correspond to observations and columns correspond to variables.
gpar0	(optional) A list containing the initial parameters of the mixture model. See the 'Details' section.
G	The range of values for the number of clusters.
max.iter	(optional) A numerical parameter giving the maximum number of iterations each EM algorithm is allowed to use.
label	(optional) A n dimensional vector, if label[i]=k then observation i belongs to group k, if NULL then the data has no known groups.
eps	(optional) A number specifying the epsilon value for the convergence criteria used in the EM algorithms. For each algorithm, the criterion is based on the difference between the log-likelihood at an iteration and an asymptotic estimate of the log-likelihood at that iteration. This asymptotic estimate is based on the Aitken acceleration.
method	(optional) A string indicating the initialization criterion, if not specified kmeans clustering is used. Alternative methods are: hierarchical "hierarchical", random "random", kmedoids "kmedoids", and model based "modelBased"
scale	(optional) A logical value indicating whether or not the data should be scaled, true by default.
nr	(optional) A number indicating the number of starting value when random is used, 10 by default.
modelSel	(optional) A string indicating the model selection criterion, if not specified AIC is used. Alternative methods are: BIC,ICL, and AIC3

Details

The arguments gpar0, if specified, is a list structure containing at least one p dimensional vector mu, alpha and phi, a p x p matrix gamma, and a p x 2 matrix cpl containing the vector omega and the vector lambda.

Value

A S4 object of class [MixGHD](#) with slots:

index	Value of the index used for model selection (AIC or ICL or BIC or AIC3) for each G, the index used is specified by the user, if not specified AIC is used.
BIC	Bayesian information criterion.
ICL	Integrated completed likelihood.
AIC	Akaike information criterion.
AIC3	Akaike information criterion 3.
gpar	A list of the model parameters
loglik	The log-likelihood values.
map	A vector of integers indicating the maximum a posteriori classifications for the best model.
z	A matrix giving the raw values upon which map is based.

Author(s)

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References

C. Tortora, B.C. Franczak, R.P. Browne, and P.D. McNicholas (2019). A Mixture of Coalesced Generalized Hyperbolic Distributions. *Journal of Classification* 36(1) 26-57.
 C. Tortora, R. P. Browne, A. ElSherbiny, B. C. Franczak, and P. D. McNicholas (2021). Model-Based Clustering, Classification, and Discriminant Analysis using the Generalized Hyperbolic Distribution: MixGHD R package, *Journal of Statistical Software* 98(3) 1–24, <doi:10.18637/jss.v098.i03>.

See Also

[MGHD MSGHD](#)

Examples

```
##Generate random data
set.seed(3)

mu1 <- mu2 <- c(0,0)
Sigma1 <- matrix(c(1,0.85,0.85,1),2,2)
Sigma2 <- matrix(c(1,-0.85,-0.85,1),2,2)

X1 <- mvrnorm(n=150,mu=mu1,Sigma=Sigma1)
X2 <- mvrnorm(n=150,mu=mu2,Sigma=Sigma2)

X <- rbind(X1,X2)

##model estimation
em=cMSGHD(X,G=2,max.iter=30,method="random",nr=2)

#result
plot(em)
```

 coef

Coefficients for objects of class MixGHD

Description

Coefficients of the estimated model.

Usage

```
## S4 method for signature 'MixGHD'
coef(object)
```

Arguments

object An S4 object of class [MixGHD](#).

Value

The coefficients of the estimated model

Author(s)

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Examples

```
##loading bankruptcy data
data(bankruptcy)

##model estimation
res=MCGHD(data=bankruptcy[,2:3],G=2,method="kmedoids",max.iter=30)
#rcoefficients of the model
coef(res)
```

contourpl *Contour plot*

Description

Contour plot for a given set of parameters.

Usage

```
contourpl(input)
```

Arguments

input An S4 object of class [MixGHD](#).

Value

The contour plot

Author(s)

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Examples

```
##loading bankruptcy data
data(bankruptcy)

##model estimation
res=MCGHD(data=bankruptcy[,2:3],G=2,method="kmedoids",max.iter=30)
#result
contourpl(res)
```

DA

Discriminant analysis using the mixture of generalized hyperbolic distributions.

Description

Carries out model-based discriminant analysis using 5 different models: the mixture of multiple scaled generalized hyperbolic distributions (MGHD), the mixture of generalized hyperbolic factor analyzers (MGHFA), the mixture of multiple scaled generalized hyperbolic distributions (MSGHD), the mixture of convex multiple scaled generalized hyperbolic distributions (cMSGHD) and the mixture of coalesced generalized hyperbolic distributions (MCGHD).

Usage

```
DA(train,trainL,test,testL,method="MGHD",starting="km",max.iter=100,
eps=1e-2,q=2,scale=TRUE)
```

Arguments

<code>train</code>	A $n_1 \times p$ matrix or data frame such that rows correspond to observations and columns correspond to variables of the training data set.
<code>trainL</code>	A n_1 dimensional vector of membership for the units of the training set. If <code>trainL[i]=k</code> then observation belongs to group k .
<code>test</code>	A $n_2 \times p$ matrix or data frame such that rows correspond to observations and columns correspond to variables of the test data set.
<code>testL</code>	A n_2 dimensional vector of membership for the units of the test set. If <code>testL[i]=k</code> then observation belongs to group k .
<code>method</code>	(optional) A string indicating the method to be used form discriminant analysis , if not specified MGHD is used. Alternative methods are: MGHFA, MSGHD, cMSGHD, MCGHD.
<code>starting</code>	(optional) A string indicating the initialization criterion, if not specified kmeans clustering is used. Alternative methods are: hierarchical "hierarchical",random "random", kmedoids "kmedoids", and model based "modelBased"
<code>max.iter</code>	(optional) A numerical parameter giving the maximum number of iterations each EM algorithm is allowed to use.

eps	(optional) A number specifying the epsilon value for the convergence criteria used in the EM algorithms. For each algorithm, the criterion is based on the difference between the log-likelihood at an iteration and an asymptotic estimate of the log-likelihood at that iteration. This asymptotic estimate is based on the Aitken acceleration.
q	(optional) used only if MGHFA method is selected. A numerical parameter giving the number of factors.
scale	(optional) A logical value indicating whether or not the data should be scaled, true by default.

Value

A list with components

model	An S4 object of class <code>MixGHD</code> with the model parameters.
testMembership	A vector of integers indicating the membership of the units in the test set
ARITest	A value indicating the adjusted rand index for the test set.
ARITrain	A value indicating the adjusted rand index for the train set.

Author(s)

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References

R.P. Browne, and P.D. McNicholas (2015). A Mixture of Generalized Hyperbolic Distributions. *Canadian Journal of Statistics*, 43.2 176-198.
 C. Tortora, B.C. Franczak, R.P. Browne, and P.D. McNicholas (2019). A Mixture of Coalesced Generalized Hyperbolic Distributions. *Journal of Classification* (to appear).
 C.Tortora, P.D. McNicholas, and R.P. Browne (2016). Mixtures of Generalized Hyperbolic Factor Analyzers. *Advanced in data analysis and classification* 10(4) p.423-440.

See Also

"`MixGHD`" `MGHD` `MGHFA` `MSGHD` `cMSGHD` `MCGHD` `ARI` `MixGHD-class` `MixGHD`

Examples

```
##loading banknote data
data(banknote)
banknote[,1]=as.numeric(factor(banknote[,1]))

##divide the data in training set and test set
train=banknote[c(1:74,126:200),]
test=banknote[75:125,]

##model estimation
```

```

model=DA(train[,2:7],train[,1],test[,2:7],test[,1],method="MGHD",max.iter=20)

#result
model$ARItest

```

dCGHD *Density of a coalesced generalized hyperbolic distribution (MSGHD).*

Description

Compute the density of a p dimensional coalesced generalized hyperbolic distribution.

Usage

```

dCGHD(data,p,mu=rep(0,p),alpha=rep(0,p),sigma=diag(p),lambda=1,omega=1,
       omegav=rep(1,p),lambdav=rep(1,p),wg=0.5,gam=NULL,phi=NULL)

```

Arguments

<code>data</code>	$n \times p$ data set
<code>p</code>	number of variables.
<code>mu</code>	(optional) the p dimensional mean
<code>alpha</code>	(optional) the p dimensional skewness parameter α
<code>sigma</code>	(optional) the $p \times p$ dimensional scale matrix
<code>lambda</code>	(optional) the 1 dimensional index parameter λ
<code>omega</code>	(optional) the 1 dimensional concentration parameter ω
<code>omegav</code>	(optional) the p dimensional concentration parameter ω_{av}
<code>lambdav</code>	(optional) the p dimensional index parameter λ_{av}
<code>wg</code>	(optional) weight
<code>gam</code>	(optional) the $p \times p$ gamma matrix
<code>phi</code>	(optional) the p dimensional vector ϕ

Details

The default values are: 0 for the mean and the skewness parameter α , $\text{diag}(p)$ for σ , 1 for ω , and 0.5 for λ .

Value

A n dimensional vector with the density from a coalesced generalized hyperbolic distribution

Author(s)

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References

C. Tortora, B.C. Franczak, R.P. Browne, and P.D. McNicholas (2019). A Mixture of Coalesced Generalized Hyperbolic Distributions. *Journal of Classification* (to appear).

Examples

```
x = seq(-3,3,length.out=30)
y = seq(-3,3,length.out=30)
xyS1 = matrix(0,nrow=length(x),ncol=length(y))
for(i in 1:length(x)){
  for(j in 1:length(y)){
    xy <- matrix(cbind(x[i],y[j]),1,2)
    xyS1[i,j] = dCGHD(xy,2)
  }
}
contour(x=x,y=y,z=xyS1, levels=c(.005,.01,.025,.05,.1,.25), main="CGHD",ylim=c(-3,3), xlim=c(-3,3))
```

dGHD

Density of a generalized hyperbolic distribution (GHD).

Description

Compute the density of a p dimensional generalized hyperbolic distribution.

Usage

```
dGHD(data,p, mu=rep(0,p),alpha=rep(0,p),sigma=diag(p),omega=1,lambda=0.5, log=FALSE)
```

Arguments

data	n x p data set
p	number of variables.
mu	(optional) the p dimensional mean
alpha	(optional) the p dimensional skewness parameter alpha
sigma	(optional) the p x p dimensional scale matrix
omega	(optional) the unidimensional concentration parameter omega
lambda	(optional) the unidimensional index parameter lambda
log	(optional) if TRUE returns the log of the density

Details

The default values are: 0 for the mean and the skewness parameter α , $\text{diag}(p)$ for σ , 1 for ω , and 0.5 for λ .

Value

A n dimensional vector with the density from a generalized hyperbolic distribution

Author(s)

Cristina Tortora, Aisha ElSherbiny, Ryan P. Browne, Brian C. Franczak, and Paul D. McNicholas.
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References

R.P. Browne, and P.D. McNicholas (2015). A Mixture of Generalized Hyperbolic Distributions. *Canadian Journal of Statistics*, 43.2 176-198

Examples

```
x = seq(-3,3,length.out=50)
y = seq(-3,3,length.out=50)
xyS1 = matrix(0,nrow=length(x),ncol=length(y))
for(i in 1:length(x)){
  for(j in 1:length(y)){
    xy <- matrix(cbind(x[i],y[j]),1,2)
    xyS1[i,j] = dGHD(xy,2)
  }
}
contour(x=x,y=y,z=xyS1, levels=c(.005,.01,.025,.05,.1,.25), main="MGHD",ylim=c(-3,3), xlim=c(-3,3))
```

dMSGHD

Density of a multiple-scaled generalized hyperbolic distribution (MS-GHD).

Description

Compute the density of a p dimensional multiple-scaled generalized hyperbolic distribution.

Usage

```
dMSGHD(data,p,mu=rep(0,p),alpha=rep(0,p),sigma=diag(p),omegav=rep(1,p),
  lambdav=rep(0.5,p),gam=NULL,phi=NULL,log=FALSE)
```

Arguments

data	n x p data set
p	number of variables.
mu	(optional) the p dimensional mean
alpha	(optional) the p dimensional skewness parameter alpha
sigma	(optional) the p x p dimensional scale matrix
omegav	(optional) the p dimensional concentration parameter omega
lambda	(optional) the p dimensional index parameter lambda
gam	(optional) the p x p gamma matrix
phi	(optional) the p dimensional vector phi
log	(optional) if TRUE returns the log of the density

Details

The default values are: 0 for the mean and the skewness parameter alpha, $\text{diag}(p)$ for sigma, 1 for omega, and 0.5 for lambda.

Value

A n dimensional vector with the density from a multiple-scaled generalized hyperbolic distribution

Author(s)

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References

C. Tortora, B.C. Franczak, R.P. Browne, and P.D. McNicholas (2019). A Mixture of Coalesced Generalized Hyperbolic Distributions. *Journal of Classification* (to appear).

Examples

```
x = seq(-3,3,length.out=50)
y = seq(-3,3,length.out=50)
xyS1 = matrix(0,nrow=length(x),ncol=length(y))
for(i in 1:length(x)){
  for(j in 1:length(y)){
    xy <- matrix(cbind(x[i],y[j]),1,2)
    xyS1[i,j] = dMSGHD(xy,2)
  }
}
contour(x=x,y=y,z=xyS1, levels=seq(.005,.25,by=.005), main="MSGHD")
```

MCGHD

*Mixture of coalesced generalized hyperbolic distributions (MCGHD).***Description**

Carries out model-based clustering using the mixture of coalesced generalized hyperbolic distributions.

Usage

```
MCGHD(data=NULL, gpar0=NULL, G=2, max.iter=100, eps=1e-2, label=NULL,
       method="km", scale=TRUE, nr=10, modelSel="AIC")
```

Arguments

data	A $n \times p$ matrix or data frame such that rows correspond to observations and columns correspond to variables.
gpar0	(optional) A list containing the initial parameters of the mixture model. See the 'Details' section.
G	The range of values for the number of clusters.
max.iter	(optional) A numerical parameter giving the maximum number of iterations each EM algorithm is allowed to use.
eps	(optional) A number specifying the epsilon value for the convergence criteria used in the EM algorithms. For each algorithm, the criterion is based on the difference between the log-likelihood at an iteration and an asymptotic estimate of the log-likelihood at that iteration. This asymptotic estimate is based on the Aitken acceleration.
label	(optional) A n dimensional vector, if $label[i]=k$ then observation i belongs to group k , If $label[i]=0$ then observation i has no known group, if NULL then the data has no known groups.
method	(optional) A string indicating the initialization criterion, if not specified kmeans clustering is used. Alternative methods are: hierarchical "hierarchical", random "random", and model based "modelBased"
scale	(optional) A logical value indicating whether or not the data should be scaled, true by default.
nr	(optional) A number indicating the number of starting value when random is used, 10 by default.
modelSel	(optional) A string indicating the model selection criterion, if not specified AIC is used. Alternative methods are: BIC, ICL, and AIC3

Details

The arguments gpar0, if specified, has to be a list structure containing as much element as the number of components G . Each element must include the following parameters: one p dimensional vector μ , α and ϕ , a $p \times p$ matrix γ , a $p \times 2$ vector cpl containing the vectors ω and λ , and a 2-dimensional vector containing the ω_0 and λ_0 .

Value

A S4 object of class [MixGHD](#) with slots:

index	Value of the index used for model selection (AIC or ICL or BIC or AIC3) for each G, the index used is specified by the user, if not specified AIC is used.
BIC	Bayesian information criterion.
ICL	Integrated completed likelihood..
AIC	Akaike information criterion.
AIC3	Akaike information criterion 3.
gpar	A list of the model parameters in the rotated space.
loglik	The log-likelihood values.
map	A vector of integers indicating the maximum a posteriori classifications for the best model.
par	A list of the model parameters.
z	A matrix giving the raw values upon which map is based.

Author(s)

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References

C. Tortora, B.C. Franczak, R.P. Browne, and P.D. McNicholas (2019). A Mixture of Coalesced Generalized Hyperbolic Distributions. *Journal of Classification* 36(1) 26-57.
C. Tortora, R. P. Browne, A. ElSherbiny, B. C. Franczak, and P. D. McNicholas (2021). Model-Based Clustering, Classification, and Discriminant Analysis using the Generalized Hyperbolic Distribution: MixGHD R package, *Journal of Statistical Software* 98(3) 1–24, <doi:10.18637/jss.v098.i03>.

See Also

[MGHD](#), [MSGHD](#)

Examples

```
##loading banknote data
data(banknote)

##model estimation
model=MCGHD(banknote[,2:7],G=2,max.iter=20)

#result
#summary(model)
#plot(model)
table(banknote[,1],model@map)
```

MGHD

*Mixture of generalized hyperbolic distributions (MGHD).***Description**

Carries out model-based clustering and classification using the mixture of generalized hyperbolic distributions.

Usage

```
MGHD(data=NULL, gpar0=NULL, G=2, max.iter=100, label=NULL, eps=1e-2,
      method="kmeans", scale=TRUE, nr=10, modelSel="AIC")
```

Arguments

<code>data</code>	A $n \times p$ matrix or data frame such that rows correspond to observations and columns correspond to variables.
<code>gpar0</code>	(optional) A list containing the initial parameters of the mixture model. See the 'Details' section.
<code>G</code>	The range of values for the number of clusters.
<code>max.iter</code>	(optional) A numerical parameter giving the maximum number of iterations each EM algorithm is allowed to use.
<code>label</code>	(optional) A n dimensional vector, if <code>label[i]=k</code> then observation i belongs to group k , If <code>label[i]=0</code> then observation i has no known group, if <code>NULL</code> then the data has no known groups.
<code>eps</code>	(optional) A number specifying the epsilon value for the convergence criteria used in the EM algorithms. For each algorithm, the criterion is based on the difference between the log-likelihood at an iteration and an asymptotic estimate of the log-likelihood at that iteration. This asymptotic estimate is based on the Aitken acceleration.
<code>method</code>	(optional) A string indicating the initialization criterion, if not specified <code>kmeans</code> clustering is used. Alternative methods are: hierarchical " <code>hierarchical</code> ", random " <code>random</code> ", and model based " <code>modelBased</code> " clustering
<code>scale</code>	(optional) A logical value indicating whether or not the data should be scaled, true by default.
<code>nr</code>	(optional) A number indicating the number of starting value when random is used, 10 by default.
<code>modelSel</code>	(optional) A string indicating the model selection criterion, if not specified <code>AIC</code> is used. Alternative methods are: <code>BIC</code> , <code>ICL</code> , and <code>AIC3</code>

Details

The arguments `gpar0`, if specified, is a list structure containing at least one p dimensional vector μ , and α , a $p \times p$ matrix σ , and a 2 dimensional vector containing ω and λ .

Value

A S4 object of class `MixGHD` with slots:

<code>index</code>	Value of the index used for model selection (AIC or ICL or BIC or AIC3) for each G, the index used is specified by the user, if not specified AIC is used.
<code>BIC</code>	Bayesian information criterion.
<code>ICL</code>	Integrated completed likelihood..
<code>AIC</code>	Akaike information criterion.
<code>AIC3</code>	Akaike information criterion 3.
<code>gpar</code>	A list of the model parameters.
<code>loglik</code>	The log-likelihood values.
<code>map</code>	A vector of integers indicating the maximum a posteriori classifications for the best model.
<code>z</code>	A matrix giving the raw values upon which map is based.

Author(s)

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References

R.P. Browne, and P.D. McNicholas (2015). A Mixture of Generalized Hyperbolic Distributions. *Canadian Journal of Statistics*, 43.2 176-198\ C. Tortora, R. P. Browne, A. ElSherbiny, B. C. Franczak, and P. D. McNicholas (2021). Model-Based Clustering, Classification, and Discriminant Analysis using the Generalized Hyperbolic Distribution: MixGHD R package, *Journal of Statistical Software* 98(3) 1–24, <doi:10.18637/jss.v098.i03>.

Examples

```
##loading crabs data
data(crabs)

##model estimation
model=MGHD(data=crabs[,4:8], G=2 )

#result
plot(model)
table(model@map, crabs[,2])

## Classification
##loading bankruptcy data
data(bankruptcy)
#70% belong to the training set
label=bankruptcy[,1]
#for a Classification porpuse the label cannot be 0
label[1:33]=2
a=round(runif(20)*65+1)
```

```

label[a]=0

##model estimation
model=MGHD(data=bankruptcy[,2:3], G=2, label=label )

#result
table(model@map,bankruptcy[,1])
plot(model)

```

MGHFA

Mixture of generalized hyperbolic factor analyzers (MGHFA).

Description

Carries out model-based clustering and classification using the mixture of generalized hyperbolic factor analyzers.

Usage

```

MGHFA(data=NULL, gpar0=NULL, G=2, max.iter=100,
label =NULL ,q=2,eps=1e-2 , method="kmeans", scale=TRUE ,nr=10)

```

Arguments

data	A matrix or data frame such that rows correspond to observations and columns correspond to variables.
gpar0	(optional) A list containing the initial parameters of the mixture model. See the 'Details' section.
G	The range of values for the number of clusters.
max.iter	(optional) A numerical parameter giving the maximum number of iterations each EM algorithm is allowed to use.
label	(optional) A n dimensional vector, if label[i]=k then observation i belongs to group k, If label[i]=0 then observation i has no known group, if NULL then the data has no known groups.
q	The range of values for the number of factors.
eps	(optional) A number specifying the epsilon value for the convergence criteria used in the EM algorithms. For each algorithm, the criterion is based on the difference between the log-likelihood at an iteration and an asymptotic estimate of the log-likelihood at that iteration. This asymptotic estimate is based on the Aitken acceleration.
method	(optional) A string indicating the initialization criterion, if not specified kmeans clustering is used. Alternative methods are: hierarchical "hierarchical" and model based "modelBased" clustering
scale	(optional) A logical value indicating whether or not the data should be scaled, true by default.
nr	(optional) A number indicating the number of starting value when random is used, 10 by default.

Details

The arguments `gpar0`, if specified, is a list structure containing at least one p dimensional vector `mu`, `alpha` and `phi`, a $p \times p$ matrix `gamma`, a 2 dimensional vector `cpl` containing `omega` and `lambda`.

Value

A S4 object of class `MixGHD` with slots:

<code>Index</code>	Bayesian information criterion value for each combination of G and q .
<code>BIC</code>	Bayesian information criterion.
<code>gpar</code>	A list of the model parameters.
<code>loglik</code>	The log-likelihood values.
<code>map</code>	A vector of integers indicating the maximum a posteriori classifications for the best model.
<code>z</code>	A matrix giving the raw values upon which <code>map</code> is based.

Author(s)

Cristina Tortora, Aisha ElSherbiny, Ryan P. Browne, Brian C. Franczak, and Paul D. McNicholas.
Maintainer: Cristina Tortora <crisrina.tortora@sjsu.edu>

References

C.Tortora, P.D. McNicholas, and R.P. Browne (2016). Mixtures of Generalized Hyperbolic Factor Analyzers. *Advanced in data analysis and classification* 10(4) p.423-440.\ C. Tortora, R. P. Browne, A. ElSherbiny, B. C. Franczak, and P. D. McNicholas (2021). Model-Based Clustering, Classification, and Discriminant Analysis using the Generalized Hyperbolic Distribution: `MixGHD` R package, *Journal of Statistical Software* 98(3) 1–24, <doi:10.18637/jss.v098.i03>.

Examples

```
## Classification
#70% belong to the training set
data(sonar)
label=sonar[,61]
set.seed(4)
a=round(runif(62)*207+1)
label[a]=0

##model estimation
model=MGHFA(data=sonar[,1:60], G=2, max.iter=25 ,q=2,label=label)

#result
table(model@map,sonar[,61])
summary(model)
```

 MixGHD class

 Class "MixGHD"

Description

This class pertains to results of the application of function [MGHD](#), [MSGHD](#), [cMSGHD](#), [MCGHD](#), and [MGHFA](#).

Objects from the Class

Objects can be created as a result to a call to [MGHD](#), [MSGHD](#), [cMSGHD](#), [MCGHD](#), and [MGHFA](#).

Slots

index Value of the index used for model selection (AIC or ICL or BIC or AIC3) for each G, the index used is specified by the user, if not specified AIC is used.

BIC Bayesian information criterion value.

ICL ICL index.

AIC AIC index.

AIC3 AIC3 index.

gpar A list of the model parameters (in the rotated space for MCGHD).

loglik The log-likelihood values.

map A vector of integers indicating the maximum a posteriori classifications for the best model.

par Only for MCGHD. A list of the model parameters.

z A matrix giving the raw values upon which map is based.

Methods

plot `signature(x = "MixGHD")` Provides plots of [MixGHD-class](#) by plotting the following elements:

- the value of the log likelihood for each iteration.
- Scatterplot of the data of all the possible couples of coordinates coloured according to the cluster. Only for less than 10 variables.
- If the number of variables is two: scatterplot and contour plot of the data coloured according to the cluster

summary `summary(x = "MixGHD")`.

Provides a summary of [MixGHD-class](#) objects by printing the following elements:

- The number components used for the model
- BIC;
- AIC;
- AIC3;
- ICL;
- A table with the number of element in each cluster.

Author(s)

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See Also

[MixGHD-class](#)

Examples

```
##loading bankruptcy data
data(bankruptcy)

##model estimation
#res=MCGHD(data=bankruptcy[,2:3],G=2,method="kmedoids",max.iter=30)
#result
#plot(res)
#summary(res)
```

 MixGHD-class

Class MixGHD.

Description

This class pertains to results of the application of function [MGHD](#), [MCGHD](#), [MSGHD](#), [cMSGHD](#).

Details

Plot the loglikelihood value for each iteration of the EM algorithm. If $p=2$ it shows a contour plot. If $2 < p < 10$ shows a splom of the data colored according to the cluster membership.

Slots

Index Bayesian information criterion value for each combination of G and q .

BIC Bayesian information criterion value.

gpar A list of the model parameters.

loglik The log-likelihood values.

map A vector of integers indicating the maximum a posteriori classifications for the best model.

z A matrix giving the raw values upon which map is based.

method A string indicating the used method: MGHD, MGHFA, MSGHD, cMSGHD, MCGHD.

data A matrix or data frame such that rows correspond to observations and columns correspond to variables.

par (only for MCGHD) A list of the model parameters in the rotated space.

Methods

signature(x = "MixGHD", y = "missing") S4 method for plotting objects of [MixGHD-class](#).

Author(s)

Cristina Tortora, Aisha ElSherbiny, Ryan P. Browne, Brian C. Franczak, and Paul D. McNicholas.
Maintainer: Cristina Tortora <crisrina.tortora@sjsu.edu>

See Also

[MixGHD-class](#), [MGHD](#), [MCGHD](#), [MSGHD](#), [cMSGHD](#), [MGHFA](#)

Examples

```
##loading banknote data
data(bankruptcy)

##model estimation
model=MSGHD(bankruptcy[,2:3],G=2,max.iter=30)

#result
summary(model)
plot(model)
```

MSGHD	<i>Mixture of multiple scaled generalized hyperbolic distributions (MSGHD).</i>
-------	---

Description

Carries out model-based clustering using the mixture of multiple scaled generalized hyperbolic distributions.

Usage

```
MSGHD(data=NULL,gpar0=NULL,G=2,max.iter=100,label=NULL,eps=1e-2,
method="km",scale=TRUE,nr=10, modelSel="AIC")
```

Arguments

data	A n x p matrix or data frame such that rows correspond to observations and columns correspond to variables.
gpar0	(optional) A list containing the initial parameters of the mixture model. See the 'Details' section.
G	The range of values for the number of clusters.

<code>max.iter</code>	(optional) A numerical parameter giving the maximum number of iterations each EM algorithm is allowed to use.
<code>label</code>	(optional) A n dimensional vector, if <code>label[i]=k</code> then observation <code>i</code> belongs to group <code>k</code> , If <code>label[i]=0</code> then observation <code>i</code> has no known group, if <code>NULL</code> then the data has no known groups.
<code>eps</code>	(optional) A number specifying the epsilon value for the convergence criteria used in the EM algorithms. For each algorithm, the criterion is based on the difference between the log-likelihood at an iteration and an asymptotic estimate of the log-likelihood at that iteration. This asymptotic estimate is based on the Aitken acceleration.
<code>method</code>	(optional) A string indicating the initialization criterion, if not specified <code>kmeans</code> clustering is used. Alternative methods are: hierarchical " <code>hierarchical</code> ", random " <code>random</code> ", and model based " <code>modelBased</code> " clustering
<code>scale</code>	(optional) A logical value indicating whether or not the data should be scaled, true by default.
<code>nr</code>	(optional) A number indicating the number of starting value when random is used, 10 by default.
<code>modelSel</code>	(optional) A string indicating the model selection criterion, if not specified AIC is used. Alternative methods are: BIC,ICL, and AIC3

Details

The arguments `gpar0`, if specified, is a list structure containing at least one `p` dimensional vector `mu`, `alpha` and `phi`, a `pxp` matrix `gamma`, and a `px2` matrix `cpl` containing the vector `omega` and the vector `lambda`.

Value

A S4 object of class `MixGHD` with slots:

<code>index</code>	Value of the index used for model selection (AIC or ICL or BIC or AIC3) for each <code>G</code> ,the index used is specified by the user, if not specified AIC is used.
<code>BIC</code>	Bayesian information criterion.
<code>ICL</code>	Integrated completed likelihood.
<code>AIC</code>	Akaike information criterion.
<code>AIC3</code>	Akaike information criterion 3.
<code>gpar</code>	A list of the model parameters
<code>loglik</code>	The log-likelihood values.
<code>map</code>	A vector of integers indicating the maximum a posteriori classifications for the best model.
<code>z</code>	A matrix giving the raw values upon which <code>map</code> is based.

Author(s)

Cristina Tortora, Aisha ElSherbiny, Ryan P. Browne, Brian C. Franczak, and Paul D. McNicholas.
 Maintainer: Cristina Tortora <crisrina.tortora@sjsu.edu>

References

C. Tortora, B.C. Franczak, R.P. Browne, and P.D. McNicholas (2019). A Mixture of Coalesced Generalized Hyperbolic Distributions. *Journal of Classification* 36(1) 26-57. \ C. Tortora, R. P. Browne, A. ElSherbiny, B. C. Franczak, and P. D. McNicholas (2021). Model-Based Clustering, Classification, and Discriminant Analysis using the Generalized Hyperbolic Distribution: MixGHD R package, *Journal of Statistical Software* 98(3) 1–24, <doi:10.18637/jss.v098.i03>.

See Also

[MGHD](#)

Examples

```
##loading banknote data
data(banknote)

##model estimation
model=MSGHD(banknote[,2:7],G=2,max.iter=30)

#result
table(banknote[,1],model@map)
summary(model)
plot(model)
```

plot

Plot objects of class MixGHD.

Description

Plots the loglikelihood function and for $p < 10$ shows the splom of the data.

Usage

```
## S4 method for signature 'MixGHD'
plot(x,y)
```

Arguments

x	A object of MixGHD-class
;	
y	Not used; for compatibility with generic plot.

Details

Plot the loglikelihood vale for each iteration of the EM algorithm. If $p=2$ it shows a contour plot. If $2 < p < 10$ shows a splom of the data colored according to the cluster membership.

Methods

signature(x = "MixGHD", y = "missing") S4 method for plotting objects of [MixGHD-class](#).

Author(s)

Cristina Tortora. Maintainer: Cristina Tortora <crisrina.tortora@sjsu.edu>

See Also

[MixGHD-class](#), [MGHD](#), [MCGHD](#), [MSGHD](#), [cMSGHD](#), [MGHFA](#)

Examples

```
##loading banknote data
data(bankruptcy)

##model estimation
model=MCGHD(bankruptcy[,2:3],G=2,max.iter=30)

#result

plot(model)
```

predict

Membership prediction for objects of class MixGHD

Description

Cluster membership

Usage

```
## S4 method for signature 'MixGHD'
predict(object)
```

Arguments

object An S4 object of class [MixGHD](#).

Value

The cluster membership

Author(s)

Cristina Tortora Maintainer: Cristina Tortora <crisrina.tortora@sjsu.edu>

Examples

```
##loading bankruptcy data
data(bankruptcy)

##model estimation
res=MCGHD(data=bankruptcy[,2:3],G=2,method="kmedoids",max.iter=30)
#rcoefficients of the model
predict(res)
```

rCGHD	<i>Pseudo random number generation from a coalesced generalized hyperbolic distribution (MSGHD).</i>
-------	--

Description

Generate n pseudo random numbers from a p dimensional coalesced generalized hyperbolic distribution.

Usage

```
rCGHD(n,p,mu=rep(0,p),alpha=rep(0,p),sigma=diag(p),omega=1,lambda=0.5,
omegav=rep(1,p),lambdav=rep(0.5,p),wg=0.5)
```

Arguments

<code>n</code>	number of observations.
<code>p</code>	number of variables.
<code>mu</code>	(optional) the p dimensional mean
<code>alpha</code>	(optional) the p dimensional skewness parameter α
<code>sigma</code>	(optional) the $p \times p$ dimensional scale matrix
<code>lambda</code>	(optional) the 1 dimensional index parameter λ
<code>omega</code>	(optional) the 1 dimensional concentration parameter ω
<code>omegav</code>	(optional) the p dimensional concentration parameter ω
<code>lambdav</code>	(optional) the p dimensional index parameter λ
<code>wg</code>	(optional) the weight

Details

The default values are: 0 for the mean and the skewness parameter α , $\text{diag}(p)$ for σ , 1 for ω , and 0.5 for λ .

Value

A n times p matrix of numbers pseudo randomly generated from a coalesced generalized hyperbolic distribution

Author(s)

Cristina Tortora, Aisha ElSherbiny, Ryan P. Browne, Brian C. Franczak, and Paul D. McNicholas.
 Maintainer: Cristina Tortora <crisrina.tortora@sjsu.edu>

References

C. Tortora, B.C. Franczak, R.P. Browne, and P.D. McNicholas (2019). A Mixture of Coalesced Generalized Hyperbolic Distributions. *Journal of Classification* (to appear).

Examples

```
data=rCGHD(300,2,alpha=c(2,-2),omegav=c(2,2),omega=3)

plot(data)
```

rGHD	<i>Pseudo random number generation from a generalized hyperbolic distribution (GHD).</i>
------	--

Description

Generate n pseudo random numbers from a p dimensional generalized hyperbolic distribution.

Usage

```
rGHD(n,p, mu=rep(0,p),alpha=rep(0,p),sigma=diag(p),omega=1,lambda=0.5)
```

Arguments

n	number of observations.
p	number of variables.
mu	(optional) the p dimensional mean
alpha	(optional) the p dimensional skewness parameter alpha
sigma	(optional) the p x p dimensional scale matrix
omega	(optional) the unidimensional concentration parameter omega
lambda	(optional) the unidimensional index parameter lambda

Details

The default values are: 0 for the mean and the skewness parameter alpha, diag(p) for sigma, 1 for omega, and 0.5 for lambda.

Value

A n times p matrix of numbers psudo randomly generated from a generalilzed hyperbolic distribution

Author(s)

Cristina Tortora, Aisha ElSherbiny, Ryan P. Browne, Brian C. Franczak, and Paul D. McNicholas.
Maintainer: Cristina Tortora <crisrina.tortora@sjsu.edu>

References

R.P. Browne, and P.D. McNicholas (2015). A Mixture of Generalized Hyperbolic Distributions. *Canadian Journal of Statistics*, 43.2 176-198

Examples

```
data=rGHD(300,2,alpha=c(2,-2))
```

```
plot(data)
```

rMSGHD

Pseudo random number generation from a mulitple-scaled generalized hyperbolic distribution (MSGHD).

Description

Generate n pseudo random numbers from a p dimensional mulitple-scaled generalized hyperbolic distribution.

Usage

```
rMSGHD(n,p, mu=rep(0,p),alpha=rep(0,p),sigma=diag(p),omegav=rep(1,p),lambdav=rep(0.5,p))
```

Arguments

n	number of observations.
p	number of variables.
mu	(optional) the p dimensional mean
alpha	(optional) the p dimensional skewness parameter alpha
sigma	(optional) the p x p dimensional scale matrix
omegav	(optional) the p dimensional concentration parameter omega
lambdav	(optional) the p dimensional index parameter lambda

Details

The default values are: 0 for the mean and the skweness parameter alpha, diag(p) for sigma, 1 for omega, and 0.5 for lambda.

Value

A n times p matrix of numbers pseudo randomly generated from a generalized hyperbolic distribution

Author(s)

Cristina Tortora, Aisha ElSherbiny, Ryan P. Browne, Brian C. Franczak, and Paul D. McNicholas.
Maintainer: Cristina Tortora <crisina.tortora@sjsu.edu>

References

C. Tortora, B.C. Franczak, R.P. Browne, and P.D. McNicholas (2019). A Mixture of Coalesced Generalized Hyperbolic Distributions. *Journal of Classification* (to appear).

Examples

```
data=rMSGHD(300,2,alpha=c(2,-2),omegav=c(2,2))  
  
plot(data)
```

sonar

Sonar data

Description

The data report the patterns obtained by bouncing sonar signals at various angles and under various conditions. There are 208 patterns in all, 111 obtained by bouncing sonar signals off a metal cylinder and 97 obtained by bouncing signals off rocks. Each pattern is a set of 60 numbers (variables) taking values between 0 and 1.

Usage

```
data(sonar)
```

Format

A data frame with 208 observations and 61 columns. The first 60 columns contain the variables. The 61st column gives the material: 1 rock, 2 metal.

Source

UCI machine learning repository

References

R.P. Gorman and T. J. Sejnowski (1988) Analysis of hidden units in a layered network trained to classify sonar targets. *Neural Networks* **1**: 75-89

summary

Plot objects of class MixGHD.

Description

Methods for function summary aimed at summarizing the S4 classes included in the [MixGHD](#)-package

Arguments

object A object of [MixGHD-class](#).

Methods

signature(object = "MixGHD") S4 method for summarizing objects of [MixGHD-class](#).

Author(s)

Cristina Tortora. Maintainer: Cristina Tortora <cristina.tortora@sjsu.edu>

See Also

[MixGHD](#) [MixGHD-class](#), [MGHD](#), [MCGHD](#), [MSGHD](#), [cMSGHD](#), [MGHFA](#)

Examples

```
##loading banknote data
data(bankruptcy)

##model estimation
model=MSGHD(bankruptcy[,2:3],G=2,max.iter=30)

#result

summary(model)
```

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