

Package: heteromixgm (via r-universe)

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Type Package

Title Copula Graphical Models for Heterogeneous Mixed Data

Imports Matrix, igraph, parallel, tmvtnorm, glasso, BDgraph, methods, stats, utils, MASS

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Description A multi-core R package that allows for the statistical modeling of multi-group multivariate mixed data using Gaussian graphical models. Combining the Gaussian copula framework with the fused graphical lasso penalty, the 'heteromixgm' package can handle a wide variety of datasets found in various sciences. The package also includes an option to perform model selection using the AIC, BIC and EBIC information criteria, a function that plots partial correlation graphs based on the selected precision matrices, as well as simulate mixed heterogeneous data for exploratory or simulation purposes and one multi-group multivariate mixed agricultural dataset pertaining to maize yields. The package implements the methodological developments found in Hermes et al. (2024) <[doi:10.1080/10618600.2023.2289545](https://doi.org/10.1080/10618600.2023.2289545)>.

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data_sim	<i>data_sim</i>
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Description

Simulate mixed multi-group data.

Usage

```
data_sim(network, n, p, K, ncat, rho, gamma_g = NULL, gamma_o, gamma_b = NULL,
gamma_p = NULL, prob = NULL, nclass = NULL)
```

Arguments

network	Type of network, either "circle", "Random", "Cluster", "Scale-free", "AR1" or "AR2".
n	Number of observations.
p	Number of variables.
K	Number of groups.
ncat	Number of categories for ordinal variables.
rho	Dissimilarity parameter inducing dissimilarity between the K datasets.
gamma_g	Proportion of Gaussian variables in the data.
gamma_o	Proportion of ordinal variables in the data.
gamma_b	Proportion of binomial variables in the data.
gamma_p	Proportion of Poisson variables in the data..
prob	Edge occurence probability in random graph.
nclass	Number of clusters in cluster graph.

Value

z	A list of K n by p matrices representing the latent Gaussian transformed (observed) data.
theta	A list of K n by p matrices representing the precision matrices corresponding to the latent Gaussian (unobserved) data.

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References

1. Hermes, S., van Heerwaarden, J., & Behrouzi, P. (2024). Copula graphical models for heterogeneous mixed data. *Journal of Computational and Graphical Statistics*, 1-15.

Examples

```
data_sim(network = "Random", n = 10, p = 50, K = 3, ncat = 6, rho = 0.25,
gamma_o = 0.5, gamma_b = 0.1, gamma_p = 0.2, prob = 0.05)
```

heteromixgm

heteromixgm

Description

This function implements either the Gibbs or approximation method within the Gaussian copula graphical model to estimate the conditional expectation for the data that not follow Gaussianity assumption (e.g. ordinal, discrete, continuous non-Gaussian, or mixed dataset).

Usage

```
heteromixgm(X, method, lambda1, lambda2, ncores)
```

Arguments

X	A list containing K $n_k \times p$ matrices (K is the number of groups, n_k is the sample size for group k and p is the number of variables)
method	Choice between "Gibbs" and "Approximate" indicating which method to use.
lambda1	Vector containing values (in [0,1]) for the sparsity penalization of each Θ^k .
lambda2	Vector containing values (in [0,1]) for the similarity penalization between the Θ^k .
ncores	Number of cores to be used during parallel computing.

Value

Z	New transformation of the data based on given or default Sigma.
ES	Expectation of covariance matrix(diagonal scaled to 1) of the Gaussian copula graphical model.
Sigma	The covariance matrix of the latent variable given the data.
Theta	The inverse covariance matrix of the latent variable given the data.
loglik	Value of the Log likelihood under the estimated parameters.

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References

1. Hermes, S., van Heerwaarden, J., & Behrouzi, P. (2024). Copula graphical models for heterogeneous mixed data. *Journal of Computational and Graphical Statistics*, 1-15.

Examples

```
data(maize)
l1 <- c(0.4)
l2 <- c(0,0.1)
ncores <- 1
est <- heteromixgm(maize, "Approximate", l1, l2, ncores)
```

 initialize

initialize

Description

Initializes parameters to be used in the approximate method algorithm.

Usage

```
initialize(y, ncores)
```

Arguments

y	Data.
ncores	Number of cores to be used during parallel computing.

Value

ES	Expectation of covariance matrices (diagonal scaled to 1) of the Gaussian copula graphical model.
Z	New transformation of the data based on given or default Sigma.
lower_upper	Lower and upper truncation points for the truncated normal distribution.

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References

1. Hermes, S., van Heerwaarden, J., & Behrouzi, P. (2024). Copula graphical models for heterogeneous mixed data. *Journal of Computational and Graphical Statistics*, 1-15.

Examples

```
y <- list(matrix(runif(25), 5, 5),matrix(runif(25), 5, 5),matrix(runif(25),
5, 5))
ncores <- 1
initialize(y, ncores)
```

lower .upper

lower.upper

Description

Calculates lower and upper bands for each data point, using a set of cut-points which is obtained from the Gaussian copula.

Usage

```
lower .upper(y)
```

Arguments

y An $(n_k \times p)$ matrix corresponding to the data matrix (n_k is the sample size for group k and p is the number of variables).

Value

lower A n_k by p matrix representing the lower band for each data point.
upper A n_k by p matrix representing the upper band for each data point.

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References

1. Hermes, S., van Heerwaarden, J., & Behrouzi, P. (2024). Copula graphical models for heterogeneous mixed data. *Journal of Computational and Graphical Statistics*, 1-15.

Examples

```
y <- list(matrix(runif(25), 5, 5),matrix(runif(25), 5, 5),matrix(runif(25),
5, 5))
lower.upper(y[[1]])
```

maize

Maize data

Description

This is a dataset consisting of maize yields, environmental and management variables measured across 2 groups. The groups pertain to different seasons (2010 and 2013) for farms in Pawe Ethiopia.

Usage

```
data("maize")
```

Format

The format is: List of 2

Details

Contains a subset of data used in the Hermes et al. (2024) paper, which is a subset of data used in the Vasco Silva et al. (forthcoming) paper.

Source

1. Hermes, S., van Heerwaarden, J., & Behrouzi, P. (2024). Copula graphical models for heterogeneous mixed data. *Journal of Computational and Graphical Statistics*, 1-15.

References

1. Hermes, S., van Heerwaarden, J., & Behrouzi, P. (2024). Copula graphical models for heterogeneous mixed data. *Journal of Computational and Graphical Statistics*, 1-15.
2. Vasco Silva, J., J. van Heerwaarden, R. Pytrik, A. G. Laborte, K. Tesfaye, and M. K. van Ittersum (forthcoming). Big data, small explanatory power? lessons learnt with random forest predictive modeling of crop yield in contrasting farming systems.

Examples

```
data(maize)
```

modselect	<i>modselect</i>
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Description

Model selection using the AIC, BIC and eBIC.

Usage

```
modselect(est, X, l1, l2, gamma)
```

Arguments

<code>est</code>	Estimates of model obtained from <code>cgmmd()</code> function
<code>X</code>	A list of K n_k by p data matrices.
<code>l1</code>	Vector containing l1 penalty values.
<code>l2</code>	Vector containing l2 penalty values.
<code>gamma</code>	EBIC gamma parameter.

Value

<code>selectmat</code>	Matrix containing the "optimal" l1 and l2 values for each information criterion.
<code>theta_aic</code>	Estimated precision matrices using the AIC for model selection.
<code>theta_bic</code>	Estimated precision matrices using the BIC for model selection.
<code>theta_ebic</code>	Estimated precision matrices using the EBIC for model selection.

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References

1. Hermes, S., van Heerwaarden, J., & Behrouzi, P. (2024). Copula graphical models for heterogeneous mixed data. *Journal of Computational and Graphical Statistics*, 1-15.

Examples

```
X <- list(matrix(runif(25), 5, 5),matrix(runif(25), 5, 5),matrix(runif(25),
5, 5))
l1 <- c(0.4)
l2 <- c(0,0.1)
gamma <- 0.5
ncores <- 1
est <- heteromixgm(X, "Approximate", l1, l2, ncores)
modselect(est, X, l1, l2, gamma)
```

plot_pcorgraph *Plot partial correlation graphs*

Description

Plots all K partial correlation graphs based on the Θ selected using one of the information criteria.

Usage

```
plot_pcorgraph(Theta, pos_clr, neg_clr, plot_layout, label_cex)
```

Arguments

Theta	List of K selected Θ
pos_clr	Color, hexadecimal color allowed, representing the positive partial correlations in the plotted graphs.
neg_clr	Color, hexadecimal color allowed, representing the negative partial correlations in the plotted graphs.
plot_layout	Number of rows and columns for the plot layout.
label_cex	Size of the vertex labels in the plotted graphs.

Value

There is no return value. The function only shows plots in the graphics output device.

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References

1. Hermes, S., van Heerwaarden, J., & Behrouzi, P. (2024). Copula graphical models for heterogeneous mixed data. *Journal of Computational and Graphical Statistics*, 1-15.

Examples

```
temp <- data_sim(network = "Random", n = 100, p = 20, K = 4, ncat = 6, rho = 0.25,
  gamma_o = 0.5, gamma_b = 0.1, gamma_p = 0.2, prob = 0.05)
X <- temp$z
l1 <- c(0.1)
l2 <- c(0,0.1)
gamma <- 0.5
ncores <- 1
est <- heteromixgm(X, "Approximate", l1, l2, ncores)
temp = modselect(est, X, l1, l2, gamma)
plot_pcorgraph(temp$theta_aic, "green", "red", c(2,2), 4.5)
```


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