

Package: FastStepGraph (via r-universe)

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Type Package

Title A Fast Algorithm for Sparse Precision Matrix Estimation

Version 0.1.2

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Description It implements an improved and computationally faster version of the original Stepwise Gaussian Graphical Algorithm for estimating the Omega precision matrix from high-dimensional data. Zamar, R., Ruiz, M., Lafit, G. and Nogales, J. (2021) <[doi:10.52933/jdssv.v1i2.11](https://doi.org/10.52933/jdssv.v1i2.11)>.

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URL <https://github.com/juancolonna/FastStepGraph>

Depends R (>= 4.3),

Imports doParallel (>= 1.0), foreach (>= 1.5), MASS (>= 7.3)

Suggests knitr, rmarkdown, devtools

VignetteBuilder knitr

Encoding UTF-8

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RoxygenNote 7.2.3

NeedsCompilation no

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Repository <https://juancolonna.r-universe.dev>

RemoteUrl <https://github.com/juancolonna/faststepgraph>

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Contents

cv.FastStepGraph	2
FastStepGraph	3
SigmaAR	5

cv.FastStepGraph	<i>Searches for the optimal combination of alpha_f and alpha_b parameters using Cross-Validation</i>
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Description

cv.FastStepGraph implements the cross-validation for the Fast Step Graph algorithm.

Usage

```
cv.FastStepGraph(
    x,
    alpha_f_min,
    alpha_f_max,
    n_folds = 10,
    b_coef = 0.5,
    n_alpha = 20,
    nei.max = 5,
    data_scale = FALSE,
    data_shuffle = TRUE,
    max.iterations = NULL,
    return_model = FALSE,
    parallel = FALSE,
    n_cores = NULL
)
```

Arguments

x	Data matrix (of size n x p).
alpha_f_min	Minimum alpha_f value for the cross-validation procedure (example 0.1).
alpha_f_max	Maximum alpha_f value for the cross-validation procedure (example 0.9).
n_folds	Number of folds for the cross-validation procedure (default value 10). This parameter also accepts the string 'LOOCV' to perform Leave-One-Out cross-validation.
b_coef	This parameter applies the empirical rule $\alpha_b = b_coef * \alpha_f$ during the initial search for the optimal alpha_f parameter while alpha_b remains fixed, after finding optimal alpha_f, alpha_b is varied to find its optimal value. The default value of b_coef is 0.5.
n_alpha	Number of elements in the grid for the cross-validation (default value 20).
nei.max	Maximum number of variables in every neighborhood (default value 5).
data_scale	Boolean parameter (TRUE or FALSE), when to scale data to zero mean and unit variance (default FALSE).
data_shuffle	Boolean parameter (default TRUE), when samples (rows of X) must be randomly shuffled.

<code>max.iterations</code>	Maximum number of iterations (integer), the defaults values is set to $p*(p-1)$.
<code>return_model</code>	Default FALSE. If set to TRUE, at the end of cross-validation, FastStepGraph is called with the optimal parameters <code>alpha_f</code> and <code>alpha_b</code> , returning <code>vareps</code> , <code>beta</code> , <code>Edges</code> and <code>Omega</code> .
<code>parallel</code>	Boolean parameter (TRUE or FALSE), when to run Cross-Validation in parallel using a multicore architecture (default FALSE).
<code>n_cores</code>	An 'int' value specifying the number of cores do you want to use if 'parallel=TRUE'. If <code>n_cores</code> is not specified, the maximum number of cores on your machine minus one will be set automatically.

Value

A list with the values:

<code>alpha_f_opt</code>	the optimal <code>alpha_f</code> value.
<code>alpha_b_opt</code>	the optimal <code>alpha_b</code> value.
<code>CV.loss</code>	minimum loss.

If `return_model=TRUE`, then also returns:

<code>vareps</code>	Response variables.
<code>beta</code>	Regression coefficients.
<code>Edges</code>	Estimated set of edges.
<code>Omega</code>	Estimated precision matrix.

Author(s)

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Examples

```
data <- FastStepGraph::SigmaAR(30, 50, 0.4) # Simulate Gaussian Data
res <- FastStepGraph::cv.FastStepGraph(data$X, alpha_f_min=0.1, alpha_f_max = 0.9, data_scale=TRUE)
```

FastStepGraph

Fast Stepwise Gaussian Graphical Model

Description

Improved and faster implementation of the Stepwise Gaussian Graphical Algorithm.

Usage

```
FastStepGraph(
  x,
  alpha_f,
  alpha_b = NULL,
  nei.max = 5,
  data_scale = FALSE,
  max.iterations = NULL
)
```

Arguments

<code>x</code>	Data matrix (of size <code>n_samples</code> x <code>p_variables</code>).
<code>alpha_f</code>	Forward threshold (no default value).
<code>alpha_b</code>	Backward threshold. If <code>alpha_b=NULL</code> , then the rule <code>alpha_b <- 0.5*alpha_f</code> is applied.
<code>nei.max</code>	Maximum number of variables in every neighborhood (default value 5).
<code>data_scale</code>	Boolean parameter (TRUE or FALSE), when to scale data to zero mean and unit variance (default FALSE).
<code>max.iterations</code>	Maximum number of iterations (integer), the defaults values is set to $p*(p-1)$.

Value

A list with the values:

<code>vareps</code>	Response variables.
<code>beta</code>	Regression coefficients.
<code>Edges</code>	Estimated set of edges.
<code>Omega</code>	Estimated precision matrix.

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Examples

```
data <- FastStepGraph::SigmaAR(30, 50, 0.4) # Simulate Gaussian Data
G <- FastStepGraph::FastStepGraph(data$X, alpha_f = 0.22, alpha_b = 0.14, data_scale=TRUE)
```

`SigmaAR`*Simulate Covariance Matrix with an Auto-regressive (AR) Model*

Description

Helper function to simulate Simulate Gaussian Data with an Autoregressive (AR) Model

Usage

```
SigmaAR(n_rows, p_columns, phi)
```

Arguments

<code>n_rows</code>	Number of samples (rows of X).
<code>p_columns</code>	Number of variables (columns of X).
<code>phi</code>	Auto-regression coefficient.

Value

A list with the values:

<code>Sigma</code>	A covariance matrix.
<code>Omega</code>	A precision matrix.
<code>X</code>	A normalized data matrix with Gaussian distribution.

Author(s)

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Examples

```
data <- FastStepGraph::SigmaAR(30, 50, 0.4) # Simulate Gaussian Data
```

Index

`cv.FastStepGraph`, [2](#)

`FastStepGraph`, [3](#)

`SigmaAR`, [5](#)