

Package: eigenmodel (via r-universe)

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Title Semiparametric Factor and Regression Models for Symmetric Relational Data

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Description Estimation of the parameters in a model for symmetric relational data (e.g., the above-diagonal part of a square matrix), using a model-based eigenvalue decomposition and regression. Missing data is accommodated, and a posterior mean for missing data is calculated under the assumption that the data are missing at random. The marginal distribution of the relational data can be arbitrary, and is fit with an ordered probit specification. See Hoff (2007) [doi:10.48550/arXiv.0711.1146](https://doi.org/10.48550/arXiv.0711.1146), for details on the model.

License GPL-2

URL <https://pdhoff.github.io/>

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Contents

eigenmodel-package	2
addlines	3
eigenmodel_mcmc	4
eigenmodel_setup	5

plot.eigenmodel_post	6
rb_fc	6
rmvnorm	7
rUL_fc	7
rZ_fc	8
ULU	9
XB	9
Y_Gen	10
Y_impute	10
Y_Pro	11
YX_Friend	11
Index	12

eigenmodel-package	<i>Semiparametric Factor and Regression Models for Symmetric Relational Data</i>
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Description

Estimation of the parameters in a model for symmetric relational data (e.g., the above-diagonal part of a square matrix), using a model-based eigenvalue decomposition and regression. Missing data is accomodated, and a posterior mean for missing data is calculated under the assumption that the data are missing at random. The marginal distribution of the relational data can be arbitrary, and is fit with an ordered probit specification. See Hoff (2007) <arXiv:0711.1146> for details on the model.

Details

Package: eigenmodel
Type: Package
Version: 1.11
Date: 2019-05-28
License: GPL Version 2

Author(s)

Peter Hoff <peter.hoff@duke.edu>

References

Hoff (2007) “Modeling homophily and stochastic equivalence in symmetric relational data”

Examples

```

data(YX_Friend)

fit<-eigenmodel_mcmc(Y=YX_Friend$Y,X=YX_Friend$X,R=2,S=50,burn=50)

# in general you should run the Markov chain longer than 50 scans

plot(fit)

# people familiar with MCMC might want to implement
# their own Markov chains:

Y<-YX_Friend$Y
X<-YX_Friend$X

eigenmodel_setup(R=2)

for(s in 1:50) {  # you should run your chain longer than 50 scans

    Z<-rZ_fc()
    UL<-rUL_fc()
    b<-rb_fc()

}

#fit_Gen<-eigenmodel_mcmc(Y=Y_Gen,R=3,S=10000)

#fit_Pro<-eigenmodel_mcmc(Y=Y_Pro,R=3,S=10000)

```

addlines

Adds lines between nodes to an existing plot of nodes

Description

Adds lines between nodes to an existing plot of nodes

Usage

```
addlines(U, Y, col = "green", lwd = 1, lty = 1)
```

Arguments

U	an n x 2 matrix of node locations
Y	a symmetric matrix
col	color of the lines
lwd	width of the lines
lty	line type

Author(s)

Peter Hoff

eigenmodel_mcmc	<i>Approximate the posterior distribution of parameters in an eigenmodel</i>
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Description

Construct approximate samples from the posterior distribution of the parameters and latent variables in an eigenmodel for symmetric relational data.

Usage

```
eigenmodel_mcmc(Y, X = NULL, R = 2, S = 1000, seed = 1, Nss = min(S,
1000), burn = 100)
```

Arguments

Y	an n x n symmetric matrix with missing diagonal entries. Off-diagonal missing values are allowed.
X	an n x n x p array of regressors
R	the rank of the approximating factor matrix
S	number of samples from the Markov chain
seed	a random seed
Nss	number of samples to be saved
burn	number of initial scans of the Markov chain to be dropped

Value

a list with the following components:

Z_postmean	posterior mean of the latent variable in the probit specification
ULU_postmean	posterior mean of the reduced-rank approximating matrix
Y_postmean	the original data matrix with missing values replaced by posterior means
L_postsamp	samples of the eigenvalues

b_postsamp	samples of the regression coefficients
Y	original data matrix
X	original regressor array
S	number of scans of the Markov chain

Author(s)

Peter Hoff

Examples

```
data(YX_Friend)

fit<-eigenmodel_mcmc(Y=YX_Friend$Y,X=YX_Friend$X,R=2,S=50,burn=50)

# in general you should run the Markov chain longer than 50 scans

plot(fit)

#fit<-eigenmodel_mcmc(Y=Y_Gen,R=3,S=10000)

#fit<-eigenmodel_mcmc(Y=Y_Pro,R=3,S=10000)
```

eigenmodel_setup

Setup constants and starting values for an eigenmodel fit

Description

Setup constants and starting values for an eigenmodel fit

Usage

```
eigenmodel_setup(R = 0, seed = 1, em_env = .GlobalEnv)
```

Arguments

R	non-negative integer rank of the approximating matrix
seed	a random seed
em_env	enviromnemt within which to do the fitting

Author(s)

Peter Hoff

`plot.eigenmodel_post` *Plot the output of an eigenmodel fit*

Description

A graphical display of MCMC output and posterior estimates of model parameters in an eigenmodel fit. Includes 95 percent quantile-based posterior confidence intervals of regression coefficients.

Usage

```
## S3 method for class 'eigenmodel_post'
plot(x, ...)
```

Arguments

`x` an object of class `eigenmodel_post`
`...` additional plotting options

Author(s)

Peter Hoff

`rb_fc` *Sample from the full conditional distribution of the regression coefficients*

Description

Sample from the full conditional distribution of the regression coefficients in an eigenmodel

Usage

```
rb_fc(E = Z - ULU(UL))
```

Arguments

`E` a symmetric matrix

Value

a $p \times 1$ vector

Author(s)

Peter Hoff

rmvnorm

*Sample from the multivariate normal distribution***Description**

Sample from the multivariate normal distribution

Usage

```
rmvnorm(mu, Sig2)
```

Arguments

mu a p x 1 vector
Sig2 a p x p positive definite matrix

Value

a p x 1 vector

Author(s)

Peter Hoff

Examples

```
rmvnorm( c(0,0,0),diag(rep(3,1)) )
```

rUL_fc

*Sample UL from its full conditional distribution***Description**

Samples the components of a reduced rank approximating matrix from their full conditional distributions

Usage

```
rUL_fc(E = Z - XB(X, b))
```

Arguments

E an n x n symmetric matrix to be modeled with a reduced rank matrix

Value

A list with the following components:

- U an n x r matrix of eigenvectors
- L an r x r diagonal matrix of eigenvalues

Author(s)

Peter Hoff

rZ_fc	<i>Sample from the full conditional distribution of the probit latent variables</i>
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Description

Sample from the full conditional distribution of the latent variables in the ordered probit model

Usage

rZ_fc(EZ = XB(X, b) + ULU(UL), MH = TRUE)

Arguments

- EZ a symmetric matrix with elements equal to the expected values of the latent variables
- MH whether or not to do a Metropolis update in addition to the Gibbs sampling

Value

a symmetric matrix

Author(s)

Peter Hoff

ULU

Computes a matrix from its eigenvalue decomposition

Description

Computes a matrix from its eigenvalue decomposition

Usage

ULU(UL)

Arguments

UL a list with first component “U”, an $n \times r$ matrix and the second component “L”
 an $r \times r$ diagonal matrix

Value

an $n \times n$ matrix

Author(s)

Peter Hoff

XB

Computes a sociomatrix of regression effects

Description

Computes a sociomatrix of regression effects

Usage

XB(X, b)

Arguments

X an $n \times n \times p$ array
 b a $p \times 1$ vector

Value

an $n \times n$ matrix

Author(s)

Peter Hoff

Y_Gen

Relations between words in the 1st chapter of Genesis

Description

The i,j th entry of this matrix is the numerical count of the number of times word i was next to word j in the first chapter of Genesis.

Examples

```
data(Y_Gen)
```

Y_impute

Impute missing values of a sociomatrix

Description

Impute missing values of a sociomatrix

Usage

```
Y_impute()
```

Details

Imputes missing values of a sociomatrix from a matrix of latent variables and an ordered-probit specification.

Value

symmetric matrix

Author(s)

Peter Hoff

Y_Pro*Butland's protein-protein interaction data*

Description

Butland's protein-protein interaction data

References

Butland et al (2005) "Interaction network containing conserved and essential protein complexes in Escherichia coli"

Examples

```
data(Y_Pro)
```

YX_Friend*Sex, race and friendship data from a 12th grade classroom*

Description

A list in which Y encodes the presence of a friendship tie between 90 12th graders. The array X indicates pairs of the same sex and of the same race.

Source

<http://www.cpc.unc.edu/projects/addhealth/design>

Examples

```
data(YX_Friend)
```

Index

- * **datasets**
 - Y_Gen, [10](#)
 - Y_Pro, [11](#)
 - YX_Friend, [11](#)
- * **distribution**
 - rmvnorm, [7](#)
- * **models**
 - eigenmodel_mcmc, [4](#)
 - eigenmodel_setup, [5](#)
 - rb_fc, [6](#)
 - rmvnorm, [7](#)
 - rUL_fc, [7](#)
 - rZ_fc, [8](#)
 - ULU, [9](#)
 - XB, [9](#)
 - Y_impute, [10](#)
- * **multivariate**
 - addlines, [3](#)
 - eigenmodel_mcmc, [4](#)
 - eigenmodel_setup, [5](#)
 - plot.eigenmodel_post, [6](#)
 - rb_fc, [6](#)
 - rmvnorm, [7](#)
 - rUL_fc, [7](#)
 - rZ_fc, [8](#)
 - ULU, [9](#)
 - XB, [9](#)
 - Y_impute, [10](#)
- * **package**
 - eigenmodel-package, [2](#)

addlines, [3](#)

eigenmodel (eigenmodel-package), [2](#)

eigenmodel-package, [2](#)

eigenmodel_mcmc, [4](#)

eigenmodel_setup, [5](#)

plot.eigenmodel_post, [6](#)

rb_fc, [6](#)

rmvnorm, [7](#)

rUL_fc, [7](#)

rZ_fc, [8](#)

ULU, [9](#)

XB, [9](#)

Y_Gen, [10](#)

Y_impute, [10](#)

Y_Pro, [11](#)

YX_Friend, [11](#)