

Package ‘HLSM’

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Title Hierarchical Latent Space Network Model

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Description Implements Hierarchical Latent Space Network Model (HLSM) for ensemble of networks as described in Sweet, Thomas & Junker (2013). <[DOI:10.3102/1076998612458702](https://doi.org/10.3102/1076998612458702)>.

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HLSMcovplots

*Plotting functions for HLSM objects***Description**

Functions for plotting HLSM/LSM model fits of class 'HLSM'. HLSMcovplots is the most recent function to plot posterior distribution summaries. HLSMplotLikelihood() plots the likelihood, HLSMcovplots() summarizes posterior draws of the parameters from MCMC sample, and HLSMplot.fit.LS() is for plotting the mean latent position estimates.

Usage

```
HLSMplotLikelihood(object, burnin = 0, thin = 1)
HLSMcovplots(fitted.model, burnin = 0, thin = 1)
```

Arguments

object	object of class 'HLSM' obtained as an output from LSM, HLSMrandomEF() or HLSMfixedEF()
fitted.model	model fit from LSM(), HLSMrandomEF() or HLSMfixedEF()
burnin	numeric value to burn the chain for plotting the results from the 'HLSM' object
thin	a numeric thinning value

Value

returns plot objects.

Author(s)

Sam Adhikari & Tracy Sweet

Examples

```
#using advice seeking network of teachers in 15 schools
#to fit the data

#Random effect model#
priors = NULL
tune = NULL
initialVals = NULL
niter = 10

random.fit = HLSMrandomEF(Y = ps.advice.mat, FullX = ps.edge.vars.mat,
initialVals = initialVals, priors = priors,
tune = tune, tuneIn = FALSE, dd = 2, niter = niter)

HLSMcovplots(random.fit)
```

HLSMdiag	<i>Function to conduct diagnostics the MCMC chain from a random effect HLSM (and HLSMfixedEF for fixed effects model)</i>
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Description

Function to compute and report diagnostic plots and statistics for a single or multiple HLSM objects.

Usage

```
HLSMdiag(object, burnin = 0,
          diags = c('psrf', 'raftery', 'traceplot', 'autocorr'),
          col = 1:6, lty = 1)
```

Arguments

object	object or list of objects of class 'HLSM' returned by HLSMrandomEf() or HLSMfixedEF()
burnin	numeric value to burn the chain while extracting results from the 'HLSM' object. Default is burnin = 0.
diags	a character vector that is a subset of c('psrf', 'raftery', 'traceplot', 'autocorr'). Default returns all diagnostics. If only a single chain is supplied in object, 'psrf' throws a warning if explicitly requested by user.
col	a character or integer vector specifying the colors for the traceplot and autocorr plot
lty	a character or integer vector specifying the linetype for the traceplot and autocorr plot

Value

Returns an object of class "HLSMdiag". It is a list that contains variable-level diagnostic tables from either or both of the raftery diagnostic and psrf diagnostic. When returned to the console, a summary table of the diagnostics will be printed instead of the list representation of the object.

call	the matched call.
raftery	list of matrices of suggested niters, burnin, and thinning for each chain.
psrf	list containing psrf, a matrix of psrf estimates and upper limits for variable, and mpsrf the multivariate psrf estimate.

Author(s)

Christian Meyer

HLSMrandomEF	<i>Function to run the MCMC sampler in random effects latent space model, HLSMfixedEF for fixed effects model, or LSM for single network latent space model)</i>
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Description

Function to run the MCMC sampler to draw from the posterior distribution of intercept, slopes, and latent positions. `HLSMrandomEF()` fits random effects model; `HLSMfixedEF()` fits fixed effects model; `LSM()` fits single network model.

Usage

```
HLSMrandomEF(Y,edgeCov=NULL, receiverCov = NULL, senderCov = NULL,
             FullX = NULL,initialVals = NULL, priors = NULL, tune = NULL,
             tuneIn = TRUE,dd=2, niter)
```

```
HLSMfixedEF(Y,edgeCov=NULL, receiverCov = NULL, senderCov = NULL,
            FullX = NULL, initialVals = NULL, priors = NULL, tune = NULL,
            tuneIn = TRUE,dd=2, niter)
```

```
LSM(Y,edgeCov=NULL, receiverCov = NULL, senderCov = NULL,
    FullX = NULL,initialVals = NULL, priors = NULL, tune = NULL,
    tuneIn = TRUE,dd=2, estimate.intercept=FALSE, niter)
```

```
getBeta(object, burnin = 0, thin = 1)
getIntercept(object, burnin = 0, thin = 1)
getLS(object, burnin = 0, thin = 1)
getLikelihood(object, burnin = 0, thin = 1)
```

Arguments

Y	input outcome for different networks. Y can either be <ol style="list-style-type: none"> (i). list of sociomatrices for K different networks (<code>Y[[i]]</code> must be a matrix with named rows and columns) (ii). list of data frame with columns <code>Sender</code>, <code>Receiver</code> and <code>Outcome</code> for K different networks (iii). a dataframe with columns named as follows: <code>id</code> to identify network, <code>Receiver</code> for receiver nodes, <code>Sender</code> for sender nodes and finally, <code>Outcome</code> for the edge outcome.
edgeCov	data frame to specify edge level covariates with <ol style="list-style-type: none"> (i). a column for network id named <code>id</code>, (ii). a column for sender node named <code>Sender</code>, (iii). a column for receiver nodes named <code>Receiver</code>, and (iv). columns for values of each edge level covariates.

receiverCov	a data frame to specify nodal covariates as edge receivers with (i.) a column for network id named id, (ii.) a column Node for node names, and (iii). the rest for respective node level covariates.
senderCov	a data frame to specify nodal covariates as edge senders with (i). a column for network id named id, (ii). a column Node for node names, and (iii). the rest for respective node level covariates.
FullX	list of numeric arrays of dimension n by n by p of covariates for K different networks. When FullX is provided to the function, edgeCov, receiverCov and senderCov must be specified as NULL.
initialVals	an optional list of values to initialize the chain. If NULL default initialization is used, else initialVals = list(ZZ, beta, intercept, alpha). For fixed effect model beta is a vector of length p and intercept is a vector of length 1. For random effect model beta is an array of dimension K by p, and intercept is a vector of length K, where p is the number of covariates and K is the number of network. ZZ is an array of dimension NN by dd, where NN is the sum of nodes in all K networks.
priors	an optional list to specify the hyper-parameters for the prior distribution of the paramters. If priors = NULL, default value is used. Else, priors= list(MuBeta, VarBeta, MuZ, VarZ, PriorA, PriorB) MuBeta is a numeric vector of length PP + 1 specifying the mean of prior distribution for coefficients and intercept VarBeta is a numeric vector for the variance of the prior distribution of coefficients and intercept. Its length is same as that of MuBeta. MuZ is a numeric vector of length same as the dimension of the latent space, specifying the prior mean of the latent positions. VarZ is a numeric vector of length same as the dimension of the latent space, specifying diagonal of the variance covariance matrix of the prior of latent positions. PriorA, PriorB is a numeric variable to indicate the rate and scale parameters for the inverse gamma prior distribution of the hyper parameter of variance of slope and intercept
tune	an optional list of tuning parameters for tuning the chain. If tune = NULL, default tuning is done. Else, tune = list(tuneBeta, tuneInt, tuneZ). tuneBeta and tuneInt have the same structure as beta and intercept in initialVals. ZZ is a vector of length NN.
tuneIn	a logical to indicate whether tuning is needed in the MCMC sampling. Default is FALSE.

<code>dd</code>	dimension of latent space.
<code>estimate.intercept</code>	When TRUE, the intercept will be estimated. If the variance of the latent positions are of interest, <code>intercept=FALSE</code> will allow users to obtain a unique variance. The intercept can also be inputed by the user.
<code>niter</code>	number of iterations for the MCMC chain.
<code>object</code>	object of class 'HLSM' returned by <code>HLSM()</code> or <code>HLSMfixedEF()</code>
<code>burnin</code>	numeric value to burn the chain while extracting results from the 'HLSM' object. Default is <code>burnin = 0</code> .
<code>thin</code>	numeric value by which the chain is to be thinned while extracting results from the 'HLSM' object. Default is <code>thin = 1</code> .

Details

The `HLSMfixedEF` and `HLSMrandomEF` functions will not automatically assess thinning and burn-in. To ensure appropriate inference, see `HLSMdiag`. See also `LSM` for fitting network data from a single network.

Value

Returns an object of class "HLSM". It is a list with following components:

<code>draws</code>	list of posterior draws for each parameters.
<code>acc</code>	list of acceptance rates of the parameters.
<code>call</code>	the matched call.
<code>tune</code>	final tuning values

Author(s)

Sam Adhikari & Tracy Sweet

References

Tracy M. Sweet, Andrew C. Thomas and Brian W. Junker (2013), "Hierarchical Network Models for Education Research: Hierarchical Latent Space Models", *Journal of Educational and Behavioral Statistics*.

Examples

```
library(HLSM)

#Set values for the inputs of the function
priors = NULL
tune = NULL
initialVals = NULL
niter = 10
```

```
#Fixed effect HLSM on Pitt and Spillane data

fixed.fit = HLSMfixedEF(Y = ps.advice.mat, senderCov=ps.node.df,
  initialVals = initialVals,priors = priors,
  tune = tune,tuneIn = FALSE,dd = 2,niter = niter)
summary(fixed.fit)

lsm.fit = LSM(Y=School9Network,edgeCov=School9EdgeCov,
  senderCov=School9NodeCov, receiverCov=School9NodeCov, niter = niter)
names(lsm.fit)
```

schoolsAdviceData *HLSM: Included Data Sets*

Description

Data set included with the HLSM package: network variables from Pitts and Spillane (2009).

Usage

```
ps.advice.mat
ps.advice.df
ps.all.vars.mat
ps.edge.vars.mat
ps.edge.df
ps.school.vars.mat
ps.teacher.vars.mat
ps.node.df
School9Network
School9NodeCov
School9EdgeCov
```

Format

ps.advice.mat: a list of 15 sociomatrices of advice seeking network, one for each school.

ps.advice.df: a data frame of all ties.

ps.all.vars.mat: a list of 15 arrays of all the covariates, one for each school. edge.vars.mat: a list of edge level covariates for 15 different school.

ps.edge.df: a dataframe of all edge covariates.

ps.school.vars.mat: a list of school level covariates for all 15 schools.

ps.teacher.vars.mat: a list of node level covariates for all 15 schools.

ps.node.df: a dataframe of all node covariates.

ps.all.vars.mat: a single list of length 15 containing the covariates mentioned above.

School9Network: a single adjacency matrix from School 9.

School9NodeCov: a dataframe with node covariates

School9EdgeCov: a dataframe with dyad-level covariates.

Author(s)

Sam Adhikari

References

Pitts, V., & Spillane, J. (2009). "Using social network methods to study school leadership". *International Journal of Research & Method in Education*, 32, 185-207

Sweet, T.M., Thomas, A.C., and Junker, B.W. (2012). "Hierarchical Network Models for Education Research: Hierarchical Latent Space Models". *Journal of Educational and Behavioral Statistics*.

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