

# Package ‘NetMix’

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**Type** Package

**Title** Dynamic Mixed-Membership Network Regression Model

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**Description** Stochastic collapsed variational inference on mixed-membership stochastic blockmodel for networks, incorporating node-level predictors of mixed-membership vectors, as well as dyad-level predictors. For networks observed over time, the model defines a hidden Markov process that allows the effects of node-level predictors to evolve in discrete, historical periods. In addition, the package offers a variety of utilities for exploring results of estimation, including tools for conducting posterior predictive checks of goodness-of-fit and several plotting functions. The package implements methods described in Olivella, Pratt and Imai (2019) 'Dynamic Stochastic Blockmodel Regression for Social Networks: Application to International Conflicts', available at <<https://www.santiagoolivella.info/pdfs/socnet.pdf>>.

**BugReports** <https://github.com/solivella/NetMix/issues>

**NeedsCompilation** yes

**License** GPL (>= 2)

**Depends** R (>= 3.5.0)

**SystemRequirements** C++11

**Suggests** ergm (>= 3.9.4), ggplot2 (>= 3.1.1), network (>= 1.13), scales (>= 1.0.0)

**Imports** clue (>= 0.3-58), graphics (>= 3.5.2), grDevices (>= 3.5.2), gtools (>= 3.8.1), igraph (>= 1.2.4.1), lda (>= 1.4.2), Matrix (>= 1.2-15), MASS (>= 7.3-51.4), methods (>= 3.5.2), poisbinom (>= 1.0.1), Rcpp (>= 1.0.2), stats (>= 3.5.2), utils (>= 3.5.2)

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approxB	<i>Internal functions and generics for mmsbm package</i>
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### Description

These are various utilities and generic methods used by the main package function.

### Usage

```
approxB(y, d_id, pi_mat, directed = TRUE)

getZ(pi_mat)

alphaLBound(par, tot_nodes, c_t, x_t, s_mat, t_id, var_beta, mu_beta)

alphaGrad(par, tot_nodes, c_t, x_t, s_mat, t_id, var_beta, mu_beta)

.cbind.fill(...)

.scaleVars(x, keep_const = TRUE)

.transf_muvar(orig, is_var, is_array, des.mat, nblock = NULL, nstate = NULL)
```

```

.bar.legend(colPalette, range)

.mpower(mat, p)

.findPerm(block_list, target_mat = NULL, use_perms = TRUE)

.transf(mat)

.compute.alpha(X, beta)

.vcovBeta(
  all_phi,
  beta_coef,
  n.sim,
  n.blk,
  n.hmm,
  n.nodes,
  n.periods,
  mu.beta,
  var.beta,
  est_kappa,
  t_id_n,
  X
)

.e.pi(alpha_list, kappa, C_mat = NULL)

.initPi(
  soc_mats,
  dyads,
  edges,
  nodes_pp,
  dyads_pp,
  n.blocks,
  periods,
  directed,
  ctrl
)

```

### Arguments

<code>y, d_id, pi_mat, directed</code>	Internal arguments for blockmodel approximation.
<code>par</code>	Vector of parameter values.
<code>tot_nodes</code>	Integer vector; total number of nodes each node interacts with.
<code>c_t</code>	Integer matrix; samples from Poisson-Binomial counts of a node instantiating a group.

<code>x_t</code>	Numeric matrix; transposed monadic design matrices.
<code>s_mat</code>	Integer matrix; Samples of HMM states by time period.
<code>t_id</code>	Integer vector; for each node, what time-period is it observed in? zero-indexed.
<code>mu_beta, var_beta</code>	Numeric arrays; prior mean and variances of monadic coefficients.
<code>...</code>	Numeric vectors; vectors of potentially different length to be cbind-ed.
<code>x, keep_const</code>	Internal arguments for matrix scaling.
<code>orig</code>	Object to be transformed.
<code>is_var</code>	Boolean. Is the object to be transformed a variance term?
<code>is_array</code>	Boolean. Is the object to be transformed an array?
<code>des.mat</code>	Numeric matrix. Design matrix corresponding to transformed object.
<code>nblock</code>	Number of groups in model, defaults to NULL.
<code>nstate</code>	Number of hidden Markov states in model, defaults to NULL.
<code>colPalette</code>	A function produced by <code>colorRamp</code> .
<code>range</code>	The range of values to label the legend.
<code>mat</code>	Numeric matrix.
<code>p</code>	Numeric scalar; power to raise matrix to.
<code>block_list</code>	List of matrices; each element is a square, numeric matrix that defines a block-model,
<code>target_mat</code>	Numeric matrix; reference blockmodel that those in <code>block_list</code> should be aligned to. Optional, defaults to NULL.
<code>use_perms</code>	Boolean; should all row/column permutations be explored when realigning matrices? defaults to TRUE.
<code>X</code>	Numeric matrix; design matrix of monadic predictors.
<code>beta</code>	Numeric array; array of coefficients associated with monadic predictors. It of dimensions Nr. Predictors by Nr. of Blocks by Nr. of HMM states.
<code>all_phi, beta_coef, n.sim, n.blk, n.hmm, n.nodes, n.periods, mu.beta, var.beta, est_kappa, t_id_n,</code>	Additional internal arguments for covariance estimation.
<code>alpha_list</code>	List of mixed-membership parameter matrices.
<code>kappa</code>	Numeric matrix; matrix of marginal HMM state probabilities.
<code>C_mat</code>	Numeric matrix; matrix of posterior counts of block instantiations per node.
<code>soc_mats, dyads, edges, nodes_pp, dyads_pp, n.blocks, periods, ctrl</code>	Internal arguments for MM computation.

## Details

These functions are meant for internal use only.

**Value**

See individual return section for each function:

**.cbind.fill** Matrix of cbind'ed elements in . . . , with missing values in each vector filled with NA.

**.mpower** Matrix; the result of raising mat to the p power.

**.findPerm** List of permuted blockmodel matrices.

**.transf** Matrix with transformed mixed-membership vectors along its rows, s.t. no element is equal to 0.0 or 1.0.

**.compute.alpha** List of predicted alpha matrices, one element per HMM state.

**.e.pi** Matrix of predicted mixed-membership vectors along its rows, with expectation computed over marginal distribution over HMM states for each time period.

**.missing** Transformed data.frame with missing values list-wise deleted, or expanded with missing indicator variables.

**.createSocioB** List of sociomatrices.

**.vertboot2** List of bootstrapped sociomatrices.

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 coef.mmsbm

---

*Extract Regression Coefficients for a Fitted mmsbm Object*


---

**Description**

Extract Regression Coefficients for a Fitted mmsbm Object

**Usage**

```
## S3 method for class 'mmsbm'
coef(object, param = "All", ...)
```

**Arguments**

object	An object of class mmsbm, a result of a call to mmsbm
param	Character string, which set of parameters should the vcov be extracted for? One of "MonadCoef", "DyadCoef" or "All" (the default).
...	Currently ignored

**Value**

For param="DyadCoef", a numeric vector. For param="MonadCoef", an array with HMM states along the third dimension. For param="All", named list of individual return components.

**Author(s)**

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**Examples**

```
library(NetMix)
## Load datasets
data("lazega_dyadic")
data("lazega_monadic")
## Estimate model with 2 groups
lazega_mmsbm <- mmsbm(SocializeWith ~ Coworkers,
  ~ School,
  senderID = "Lawyer1",
  receiverID = "Lawyer2",
  nodeID = "Lawyer",
  data.dyad = lazega_dyadic,
  data.monad = lazega_monadic,
  n.blocks = 2,
  mmsbm.control = list(seed = 123,
    conv_tol = 1e-2,
    hessian = FALSE))

coef(lazega_mmsbm, "MonadCoef")
```

---

covFX	<i>Generate estimated monadic covariate effects for estimated mmsbm model</i>
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**Description**

The function estimates the effect of a shift in monadic covariate values on the probability of edge formation in the network.

**Usage**

```
covFX(fm, cov, shift, max.val = FALSE)
```

**Arguments**

fm	An object of class mmsbm, a result of a call to mmsbm.
cov	Character string identifying the monadic covariate to be shifted.
shift	Numeric value specifying the desired increase or decrease in the monadic covariate. The monadic predictor will be shifted by this value for all nodes and time periods.
max.val	An optional numeric value specifying the maximum possible value for the monadic covariate.

**Value**

List with named components:

**Overall Avg. Effect** Overall average effect of the covariate shift on the predicted probability of edge formation.

**Avg. Effect by Time** Vector of average effects of the covariate shift on the predicted probability of edge formation for each time period.

**Avg. Effect by Node** Vector of average effects of the covariate shift on the predicted probability of edge formation for each node.

**Avg. Effect by Dyad** Vector of average effects of the covariate shift on the predicted probability of edge formation for each node dyad.

**Avg. Effect Dyad-Time** Vector of estimated effects of the covariate shift on the predicted probability of edge formation for each node dyad-time unit.

**Author(s)**

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**Examples**

```
library(NetMix)
## Load datasets
data("lazega_dyadic")
data("lazega_monadic")
## Estimate model with 2 groups
lazega_mmsbm <- mmsbm(SocializeWith ~ Coworkers,
  ~ Age,
  senderID = "Lawyer1",
  receiverID = "Lawyer2",
  nodeID = "Lawyer",
  data.dyad = lazega_dyadic,
  data.monad = lazega_monadic,
  n.blocks = 2,
  mmsbm.control = list(seed = 123,
    conv_tol = 1e-2,
    hessian = FALSE))

## Compute effect of decreasing every lawyers' age by 10 years
fx_list <- covFX(lazega_mmsbm, cov = "Age", shift = -10)
fx_list[["Overall Avg. Effect of Age"]]
```

gof

*Posterior predictive checks using structural network characteristics***Description**

The function generates a variety of plots that serve as posterior predictive checks on the goodness of fit of a fitted `mmsbm` object.

**Usage**

```
gof(x, ...)

## S3 method for class 'mmsbm'
gof(
  x,
  gof_stat = c("Geodesics", "Degree"),
  level = 0.95,
  samples = 50,
  new.data.dyad = NULL,
  new.data.monad = NULL,
  seed = NULL,
  ...
)
```

**Arguments**

<code>x</code>	An object of class <code>mmsbm</code> , a result of a call to <code>mmsbm</code> .
<code>...</code>	Currently ignored.
<code>gof_stat</code>	Character vector. Accepts any subset from "Geodesics", "Degree", "Indegree", "Outdegree", "3-Motifs", "Dyad Shared Partners", "Edge Shared Partners", and "Incoming K-stars". See details.
<code>level</code>	Double. Level of credible interval for posterior predictive distribution around structural quantities of interest.
<code>samples</code>	Integer. Number of sampled networks from model's posterior predictive using <a href="#">simulate.mmsbm</a> .
<code>new.data.dyad</code>	See <a href="#">simulate.mmsbm</a> . Enables out-of-sample checking.
<code>new.data.monad</code>	See <a href="#">simulate.mmsbm</a> . Enables out-of-sample checking.
<code>seed</code>	See <a href="#">simulate.mmsbm</a> .

**Details**

Goodness of fit of network models has typically been established by evaluating how the structural characteristics of predicted networks compare to those of the observed network. When estimated in a Bayesian framework, this approach is equivalent to conducting posterior predictive checks on these structural quantities of interest. When `new.data.dyad` and/or `new.data.monad` are passed that are



different from those used in estimation, this is equivalent to conducting posterior predictive checks out-of-sample.

The set of structural features used to determine goodness of fit is somewhat arbitrary, and chosen mostly to incorporate various first order, second order, and (to the extent possible) third-order characteristics of the network. "Geodesics" focuses on the distribution over observed and predicted geodesic distances between nodes; "Indegree" and "Outdegree" focuses on the distribution over incoming and outgoing connections per node; "3-motifs" focus on a distribution over possible connectivity patterns between triads (i.e. the triadic census); "Dyad Shared Partners" focuses on the distribution over the number of shared partners between any two days; "Edge Shared Partners" is similarly defined, but w.r.t. edges, rather than dyads; and finally "Incoming K-stars" focuses on a frequency distribution over stars with  $k=1, \dots$  spokes.

Obtaining samples of the last three structural features can be very computationally expensive, and is discouraged on networks with more than 50 nodes.

### Value

A ggplot object.

### Author(s)

Santiago Olivella (olivella@unc.edu), Adeline Lo (aylo@wisc.edu), Tyler Pratt (tyler.pratt@yale.edu), Kosuke Imai (imai@harvard.edu)

### Examples

```
library(NetMix)
## Load datasets
data("lazega_dyadic")
data("lazega_monadic")

## Estimate model with 2 groups
lazega_mmsbm <- mmsbm(SocializeWith ~ Coworkers,
  senderID = "Lawyer1",
  receiverID = "Lawyer2",
  nodeID = "Lawyer",
  data.dyad = lazega_dyadic,
  data.monad = lazega_monadic,
  n.blocks = 2,
  mmsbm.control = list(seed = 123,
    conv_tol = 1e-2,
    hessian = FALSE))

## Plot observed (red) and simulated (gray) distributions over
## indegrees
## (typically a larger number of samples would be taken)
## (strictly requires ggplot2)

gof(lazega_mmsbm, gof_stat = "Indegree", samples = 2)
```

---

 head.mmsbm

*Identify nodes with most frequent membership in latent groups*


---

### Description

The function lists the nodes (optionally, node-time periods) that most frequently instantiate membership in each latent group.

### Usage

```
## S3 method for class 'mmsbm'
head(x, n = 6, t = NULL, node = TRUE, t.correct = FALSE, ...)
```

### Arguments

x	An object of class mmsbm, a result of a call to mmsbm.
n	Numeric or integer; specifies how many units will be identified for each group.
t	Optional vector of time periods to be used for assessing latent group membership.
node	Logical; indicates whether latent group memberships should be averaged at the node level. If FALSE, the function returns the node-time period units with highest estimated membership in each latent group.
t.correct	Logical; indicates whether latent group memberships should be corrected for temporal trends. If TRUE, the function returns the node-time period units with highest estimated membership in each latent group.
...	Currently ignored

### Value

List of length n.blocks. Each entry contains a sorted vector of average latent membership probabilities of length n.

### Author(s)

Santiago Olivella (olivella@unc.edu), Adeline Lo (aylo@wisc.edu), Tyler Pratt (tyler.pratt@yale.edu), Kosuke Imai (imai@harvard.edu)

### Examples

```
library(NetMix)
## Load datasets
data("lazega_dyadic")
data("lazega_monadic")
## Estimate model with 2 groups
set.seed(123)
lazega_mmsbm <- mmsbm(SocializeWith ~ Coworkers,
  ~ School + Practice + Status,
```

```

senderID = "Lawyer1",
receiverID = "Lawyer2",
nodeID = "Lawyer",
data.dyad = lazega_dyadic,
data.monad = lazega_monadic,
n.blocks = 2,
mmsbm.control = list(seed = 123,
                     conv_tol = 1e-2,
                     hessian = FALSE))

## Show top 6 lawyers in each estimated latent block
head(lazega_mmsbm)

```

---

lazega_dyadic	<i>Dyadic predictors in the Lazega friendship network (Lazega 2001).</i>
---------------	--

---

## Description

A dataset containing edges and dyad-level predictors in the network of friendships among lawyers in a New England law firm. More details are available in Lazega (2001).

## Usage

```
data(lazega_dyadic)
```

## Format

A data frame with 5041 rows and 4 variables:

**Lawyer1**, **Lawyer2** lawyer ID, corresponding to identifiers common to those in lazega\_monadic;  
numeric

**SocializeWith** value of edge in network; binary

**Coworkers** are the corresponding lawyers in the same office? boolean

## Source

<https://github.com/Z-co/networkdata/blob/master/networkdata/data/lazega.rda>

## References

Emmanuel Lazega, *The Collegial Phenomenon: The Social Mechanisms of Cooperation Among Peers in a Corporate Law Partnership*, Oxford University Press (2001).

---

lazega\_monadic      *Monadic predictors in the Lazega friendship network (Lazega 2001).*

---

### Description

A dataset containing vertex-level predictors in the network of sought-after advise among lawyers in a New England law firm. More details are available in Lazega (2001).

### Usage

```
data(lazega_monadic)
```

### Format

A data frame with 71 rows and 7 variables:

**Lawyer** lawyer ID, corresponding to identifiers common to those in lazega\_dyadic; numeric

**Age** age, in years; numeric

**Gender** 1=man; 2=woman; factor

**School** 1=harvard, yale; 2=ucon; 3= other; factor

**Practice** 1=litigation; 2=corporate; factor

**Seniority** time in the firm, in years; numeric

**Status** 1=partner; 2=associate; factor

### Source

Emmanuel Lazega, *The Collegial Phenomenon: The Social Mechanisms of Cooperation Among Peers in a Corporate Law Partnership*, Oxford University Press (2001).

<https://github.com/Z-co/networkdata/blob/master/networkdata/data/lazega.rda>

---

mmsbm      *Dynamic mixed-membership stochastic blockmodel with covariates*

---

### Description

The function estimates a dynamic mixed-membership stochastic blockmodel that incorporates covariates.

**Usage**

```

mmsbm(
  formula.dyad,
  formula.monad = ~1,
  senderID,
  receiverID,
  nodeID = NULL,
  timeID = NULL,
  data.dyad,
  data.monad = NULL,
  n.blocks,
  n.hmmstates = 1,
  directed = TRUE,
  mmsbm.control = list()
)

```

**Arguments**

<code>formula.dyad</code>	A formula object. The variable in <code>data.dyad</code> that contains binary edges should be used as a LHS, and any dyadic predictors can be included on the RHS (when no dyadic covariates are available, use $y \sim 1$ ). Same syntax as a <code>glm</code> formula.
<code>formula.monad</code>	An optional formula object. LHS is ignored. RHS contains names of nodal attributes found in <code>data.monad</code> .
<code>senderID</code>	Character string. Quoted name of the variable in <code>data.dyad</code> identifying the sender node. For undirected networks, the variable simply contains name of first node in dyad. Cannot contain special character "@".
<code>receiverID</code>	Character string. Quoted name of the variable in <code>data.dyad</code> identifying the receiver node. For undirected networks, the variable simply contains name of second node in dyad. Cannot contain special character "@".
<code>nodeID</code>	Character string. Quoted name of the variable in <code>data.monad</code> identifying a node in either <code>data.dyad[, senderID]</code> or <code>data.dyad[, senderID]</code> . If not NULL, every node <code>data.dyad[, senderID]</code> or <code>data.dyad[, senderID]</code> must be present in <code>data.monad[, nodeID]</code> . Cannot contain special character "@".
<code>timeID</code>	Character string. Quoted name of the variable in both <code>data.dyad</code> and <code>data.monad</code> indicating the time in which network (and corresponding nodal attributes) were observed. The variable itself must be composed of integers. Cannot contain special character "@".
<code>data.dyad</code>	Data frame. Sociomatrix in "long" (i.e. dyadic) format. Must contain at least three variables: the sender identifier (or identifier of the first node in an undirected networks dyad), the receiver identifier (or identifier of the second node in an undirected network dyad), and the value of the edge between them. Currently, only edges between zero and one (inclusive) are supported.
<code>data.monad</code>	Data frame. Nodal attributes. Must contain a node identifier matching the names of nodes used in the <code>data.dyad</code> data frame.
<code>n.blocks</code>	Integer value. How many latent groups should be used to estimate the model?

n.hmmstates	Integer value. How many hidden Markov state should be used in the HMM? Defaults to 1 (i.e. no HMM).
directed	Boolean. Is the network directed? Defaults to TRUE.
mmsbm.control	A named list of optional algorithm control parameters. <ul style="list-style-type: none"> <li><b>seed</b> Integer. Seed the RNG. By default, a random seed is generated and returned for reproducibility purposes.</li> <li><b>nstart</b> Integer. Number of random initialization trials. Defaults to 5.</li> <li><b>spectral</b> Boolean. Type of initialization algorithm for mixed-membership vectors in static case. If TRUE (default), use spectral clustering with degree correction; otherwise, use kmeans algorithm.</li> <li><b>init_gibbs</b> Boolean. Should a collapsed Gibbs sampler of non-regression mmsbm be used to initialize mixed-membership vectors, instead of a spectral or simple kmeans initialization? Setting to TRUE will result in slower initialization and faster model estimation. When TRUE, results are typically very sensitive to choice of alpha (see below).</li> <li><b>alpha</b> Numeric positive value. Concentration parameter for collapsed Gibbs sampler to find initial mixed-membership values when <code>init_gibbs=TRUE</code>. Defaults to 1.0.</li> <li><b>missing</b> Means of handling missing data. One of "indicator method" (default) or "listwise deletion".</li> <li><b>svi</b> Boolean; should stochastic variational inference be used? Defaults to TRUE.</li> <li><b>vi_iter</b> Number of maximum iterations in stochastic variational updates. Defaults to 5e2.</li> <li><b>batch_size</b> When <code>svi=TRUE</code>, proportion of nodes sampled in each local. Defaults to 0.05 when <code>svi=TRUE</code>, and to 1.0 otherwise.</li> <li><b>forget_rate</b> When <code>svi=TRUE</code>, value between (0.5,1], controlling speed of decay of weight of prior parameter values in global steps. Defaults to 0.75 when <code>svi=TRUE</code>, and to 0.0 otherwise.</li> <li><b>delay</b> When <code>svi=TRUE</code>, non-negative value controlling weight of past iterations in global steps. Defaults to 1.0 when <code>svi=TRUE</code>, and ignored otherwise.</li> <li><b>opt_iter</b> Number of maximum iterations of BFGS in global step. Defaults to 10e3.</li> <li><b>hessian</b> Boolean indicating whether the Hessian matrix of regression coefficients should be returned. Defaults to TRUE.</li> <li><b>assortative</b> Boolean indicating whether blockmodel should be assortative (i.e. stronger connections within groups) or disassortative (i.e. stronger connections between groups). Defaults to TRUE.</li> <li><b>mu_block</b> Numeric vector with two elements: prior mean of blockmodel's main diagonal elements, and prior mean of blockmodel's offdiagonal elements. Defaults to <code>c(5.0, -5.0)</code> if <code>assortative=TRUE</code> (default) and to <code>c(-5.0, 5.0)</code> otherwise.</li> <li><b>var_block</b> Numeric vector with two positive elements: prior variance of blockmodel's main diagonal elements, and prior variance of blockmodel's offdiagonal elements. Defaults to <code>c(5.0, 5.0)</code>.</li> </ul>

- mu\_beta** Either single numeric value, in which case the same prior mean is applied to all monadic coefficients, or an array that is `n.predictors` by `n.blocks` by `n.hmmstates`, where `n.predictors` is the number of monadic predictors for which a prior mean is being set (prior means need not be set for all) predictors). The rows in the array should be named to identify which variables a prior mean is being set for. Defaults to a common prior mean of 0.0 for all monadic coefficients.
- var\_beta** See `mu_beta`. Defaults to a single common prior variance of 5.0 for all (standardized) monadic coefficients.
- mu\_gamma** Either a single numeric value, in which case the same prior mean is applied to all dyadic coefficients, or a named vector of numeric values (with names corresponding to the name of the variable for which a prior mean is being set). Defaults to a common prior mean of 0.0 for all dyadic coefficients.
- var\_gamma** See `mu_gamma`. Defaults to a single common prior variance of 5.0 for all (standardized) dyadic coefficients.
- eta** Numeric positive value. Concentration hyper-parameter for HMM. Defaults to 1.0.
- se\_sim** Number of samples from variational posterior of latent variables on which approximation to variance-covariance matrices are based. Defaults to 10.
- dyad\_vcov\_samp** Maximum number of dyads to sample in computation of variance-covariance of dyadic and blockmodel parameters, when compared to ten percent of the observed dyads. Defaults to 1000.
- fixed\_mm** Optional character vector, with "nodeID@timeID" as elements, indicating which mixed-membership vectors should remain constant at their initial values throughout estimation. When only one year is observed, elements should be "nodeID@1". Typically used with `mm_init_t`.
- mm\_init\_t** Matrix, `n.blocks` by nodes across years. Optional initial values for mixed-membership vectors. Although initial values need not be provided for all nodes, column names must have a nodeID@timeID format to avoid ambiguity. When only one year is observed, names should be "nodeID@1".
- kappa\_init\_t** Matrix, `n.hmmstates` by number of years. Optional initial values for variational parameters for state probabilities. Columns must be named according to unique year values.
- b\_init\_t** Matrix, `n.blocks` by `n.blocks`. Optional initial values for blockmodel.
- beta\_init** Array, `n.predictors` by `n.blocks` by `n.hmmstates`. Optional initial values for monadic coefficients. If
- gamma\_init** Vector. Optional initial values for dyadic coefficients.
- permute** Boolean. Should all permutations be tested to realign initial block models in dynamic case? If FALSE, realignment is done via faster graph matching algorithm, but may not be exact. Defaults to TRUE.
- conv\_tol** Numeric value. Absolute tolerance for VI convergence. Defaults to 1e-3.
- verbose** Boolean. Should extra information be printed as model iterates? Defaults to FALSE.

**Value**

Object of class `mmsbm`. List with named components:

**MixedMembership** Matrix of variational posterior of mean of mixed-membership vectors. nodes by `n.blocks`.

**BlockModel** `n.blocks` by `n.blocks` matrix of estimated tie log-odds between members of corresponding latent groups. The `blockmodel`.

**vcov\_blockmodel** If `hessian=TRUE`, variance-covariance matrix of parameters in `blockmodel`, ordered in column-major order.

**MonadCoef** Array of estimated coefficient values for monadic covariates. Has `n.blocks` columns, and `n.hmmstates` slices.

**vcov\_monad** If `hessian=TRUE`, variance-covariance matrix of monadic coefficients.

**DyadCoef** Vector estimated coefficient values for dyadic covariates.

**vcov\_dyad** If `hessian=TRUE`, variance-covariance matrix of dyadic coefficients.

**TransitionKernel** Matrix of estimated HMM transition probabilities.

**Kappa** Matrix of marginal probabilities of being in an HMM state at any given point in time. `n.hmmstates` by years (or whatever time interval networks are observed at).

**LowerBound** Final LB value

**lb** Vector of all LB across iterations, useful to check early convergence issues.

**niter** Final number of VI iterations.

**converged** Convergence indicator; zero indicates failure to converge.

**NodeIndex** Order in which nodes are stored in all return objects.

**monadic.data, dyadic.data** Model frames used during estimation (stripped of attributes).

**forms** Values of selected formal arguments used by other methods.

**seed** The value of RNG seed used during estimation.

**call** Original (unevaluated) function call.

**Author(s)**

Santiago Olivella (olivella@unc.edu), Adeline Lo (aylo@wisc.edu), Tyler Pratt (tyler.pratt@yale.edu), Kosuke Imai (imai@harvard.edu)

**Examples**

```
library(NetMix)
## Load datasets
data("lazega_dyadic")
data("lazega_monadic")
## Estimate model with 2 groups
## Setting to `hessian=TRUE` increases computation time
## but is needed if standard errors are to be computed.
lazega_mmsbm <- mmsbm(SocializeWith ~ Coworkers,
                      ~ School + Practice + Status,
                      senderID = "Lawyer1",
```



```

receiverID = "Lawyer2",
nodeID = "Lawyer",
data.dyad = lazega_dyadic,
data.monad = lazega_monadic,
n.blocks = 2,
mmsbm.control = list(seed = 123,
                     conv_tol = 1e-2,
                     hessian = FALSE))

```

---

mmsbm\_fit

*Fitter Function for dynamic MMSBM Model*


---

### Description

This is the interface to the C++ fitter for the dynamic mixed-membership stochastic blockmodel for network regression.

### Arguments

z_t	Numeric matrix; transpose of monadic design matrix. Should not include intercept row.
x_t	Numeric matrix; transpose of dyadic design matrix.
y	Numeric vector; vector of edge values. Must have same number of elements as <code>ncol(x_t)</code>
time_id_dyad	Integer vector; zero-based time-period identifier for each node.
nodes_per_period	Integer vector; total number of unique nodes observed in each time period.
node_id_dyad	Integer matrix; zero-based sender and receiver identifier per dyad.
mu_b	Numeric matrix; matrix of prior means for elements in blockmodel matrix.
var_b	Numeric matrix; matrix of prior variances for elements in blockmodel matrix.
pi_init	Numeric matrix; matrix of initial mixed-memberships. Nodes along columns.
kappa_init_t	Numeric matrix; matrix of initial marginal HMM state probabilities. Time-periods along columns.
b_init_t	Numeric matrix; square matrix of initial values of blockmodel.
beta_init	Numeric vector; flat array (column-major order) of initial values of monadic coefficients.
gamma_init	Numeric vector; vector of initial values of dyadic coefficients
control	List; see the <code>mmsbm.control</code> argument of <a href="#">mmsbm</a>

### Value

Unclassed list with named components; see Value of [mmsbm](#)

**Warning**

This function is for internal use only. End-users should always resort to `mmsbm`. In particular, that interface post-processes the return value of this internal in important ways.

**Author(s)**

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---

plot.mmsbm

*Various visualization tools for 'mmsbm' objects*

---

**Description**

The function provides a variety of plotting options for a fitted `mmsbm` object.

**Usage**

```
## S3 method for class 'mmsbm'
plot(x, type = "groups", FX = NULL, node = NULL, ...)
```

**Arguments**

<code>x</code>	An object of class <code>mmsbm</code> , a result of a call to <code>mmsbm</code> .
<code>type</code>	character string denoting the type of plot. The default, "groups," plots the estimated matrix of group by group edge formation probabilities as a network plot, with nodes representing groups (sized proportional to relative membership) and edge colors encoding probability of between-group ties. "blockmodel" plots the same information, but using a tile plot instead of a network plot. "membership" plots average membership in each latent group by time period. "effect" provides a series of plots showing the estimated effect of a shift in monadic covariate values.
<code>FX</code>	with <code>type == "effect"</code> ; a list resulting from a call to <code>covFX</code> .
<code>node</code>	with <code>type == "membership"</code> ; a character string specifying the node for which group membership should be plotted.
<code>...</code>	Currently ignored

**Value**

The requested plot object.

**Author(s)**

Santiago Olivella (olivella@unc.edu), Adeline Lo (aaylo@wisc.edu), Tyler Pratt (tyler.pratt@yale.edu), Kosuke Imai (imai@harvard.edu)

**Examples**

```

library(NetMix)
## Load datasets
data("lazega_dyadic")
data("lazega_monadic")
## Estimate model with 2 groups
lazega_mmsbm <- mmsbm(SocializeWith ~ Coworkers,
  ~ School + Practice + Status,
  senderID = "Lawyer1",
  receiverID = "Lawyer2",
  nodeID = "Lawyer",
  data.dyad = lazega_dyadic,
  data.monad = lazega_monadic,
  n.blocks = 2,
  mmsbm.control = list(seed = 123,
    conv_tol = 1e-2,
    hessian = FALSE))

## Plot blockmodel as network
plot(lazega_mmsbm)

```

---

predict.mmsbm

*Predict edges based on estimated mmsbm model*


---

**Description**

The function produces expected posterior edges based on estimated parameters and (optionally new) predictor data

**Usage**

```

## S3 method for class 'mmsbm'
predict(
  object,
  new.data.dyad = NULL,
  new.data.monad = NULL,
  forecast = FALSE,
  type = c("link", "response", "mm"),
  ...
)

```

**Arguments**

object            Object of class mmsbm.  
new.data.dyad    An optional data.frame object.  
new.data.monad   An optional data.frame object.

forecast	Boolean. Should prediction forecast one step into the future? Defaults to FALSE.
type	Character string. The default is to use the linear predictor of edges. The alternative "response" returns predicted probabilities. The alternative "mm" returns predicted mixed-membership vectors.
...	Currently ignored

### Value

If `new.data.dyad = NULL`, vector of length `nrow(object$dyadic.data)`. Else, vector of length `nrow(new.data.dyad)`.

### Author(s)

Santiago Olivella (olivella@unc.edu), Adeline Lo (aylo@wisc.edu), Tyler Pratt (tyler.pratt@yale.edu), Kosuke Imai (imai@harvard.edu)

### Examples

```
library(NetMix)
## Load datasets
data("lazega_dyadic")
data("lazega_monadic")
## Estimate model with 2 groups
lazega_mmsbm <- mmsbm(SocializeWith ~ Coworkers,
  ~ School + Practice + Status,
  senderID = "Lawyer1",
  receiverID = "Lawyer2",
  nodeID = "Lawyer",
  data.dyad = lazega_dyadic,
  data.monad = lazega_monadic,
  n.blocks = 2,
  mmsbm.control = list(seed = 123,
    conv_tol = 1e-2,
    hessian = FALSE))

## Get in-sample predicted edge probabilities
lazega_preds <- predict(lazega_mmsbm, type = "response")
```

---

simulate.mmsbm

*Simulate a complete sociomatrix from an mmsbm object*

---

### Description

The function generates one sample network from the posterior predictive of the model represented by a fitted mmsbm object.

**Usage**

```
## S3 method for class 'mmsbm'
simulate(
  object,
  nsim = 1,
  seed = NULL,
  new.data.dyad = NULL,
  new.data.monad = NULL,
  ...
)
```

**Arguments**

object	An object of class mmsbm, a result of a call to mmsbm
nsim	Number of networks to simulate
seed	RNG seed.
new.data.dyad	An optional data.frame object. If not NULL, use these dyadic predictor values instead of those used to fit the original model.
new.data.monad	An optional data.frame object. See new.data.dyad.
...	Currently ignored

**Value**

List of length nsim of simulated networks. If new.data.dyad = NULL, each element is a vector of length nrow(object\$dyadic.data). Else, vector of length nrow(new.data.dyad). If seed is not NULL, return object includes its value as attribute "seed".

**Author(s)**

Santiago Olivella (olivella@unc.edu), Adeline Lo (aylo@wisc.edu), Tyler Pratt (tyler.pratt@yale.edu), Kosuke Imai (imai@harvard.edu)

**Examples**

```
library(NetMix)
## Load datasets
data("lazega_dyadic")
data("lazega_monadic")
## Estimate model with 2 groups
lazega_mmsbm <- mmsbm(SocializeWith ~ Coworkers,
  ~ School + Practice + Status,
  senderID = "Lawyer1",
  receiverID = "Lawyer2",
  nodeID = "Lawyer",
  data.dyad = lazega_dyadic,
  data.monad = lazega_monadic,
  n.blocks = 2,
  mmsbm.control = list(seed = 123,
    conv_tol = 1e-2,
```

```

                                hessian = FALSE))

## Simulate 5 new networks
lazega_sim <- simulate(lazega_mmsbm, nsim = 5, seed = 123)

```

---

summary.mmsbm	<i>Summarize 'mmsbm' object</i>
---------------	---------------------------------

---

### Description

The function summarizes the output of a dynMMSBM model object

### Usage

```

## S3 method for class 'mmsbm'
summary(object, ...)

```

### Arguments

object	An object of class mmsbm, a result of a call to mmsbm.
...	Currently ignored

### Value

List with named components:

**N** Total number of dyad-time period observations.

**Number of Clusters** Number of latent groups included in the dynMMSBM model.

**Percent of Observations in Each Cluster** Average membership in each latent group, across all node-time periods.

**Edge Formation Probabilities** n.groups by n.groups matrix of estimated edge formation probabilities between latent groups.

**Dyadic Coefficients** Vector of estimated coefficient values for dyadic covariates.

**Monadic Coefficients** Array of estimated coefficient values for monadic covariates. Has n.groups columns, and n.hmmstates slices.

**Markov State Probabilities** Average HMM state probabilities across all time periods.

### Author(s)

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**Examples**

```

library(NetMix)
## Load datasets
data("lazega_dyadic")
data("lazega_monadic")
## Estimate model with 2 groups
lazega_mmsbm <- mmsbm(SocializeWith ~ Coworkers,
  ~ School + Practice + Status,
  senderID = "Lawyer1",
  receiverID = "Lawyer2",
  nodeID = "Lawyer",
  data.dyad = lazega_dyadic,
  data.monad = lazega_monadic,
  n.blocks = 2,
  mmsbm.control = list(seed = 123,
    conv_tol = 1e-2,
    hessian = TRUE))

## Summarize estimated model
summary(lazega_mmsbm)

```

vcov.mmsbm

*Extract Variance-Covariance Matrix for a Fitted mmsbm Object***Description**

Extract Variance-Covariance Matrix for a Fitted mmsbm Object

**Usage**

```

## S3 method for class 'mmsbm'
vcov(object, param = "All", ...)

```

**Arguments**

object	An object of class mmsbm, a result of a call to mmsbm
param	Character string, which set of parameters should the vcov be extracted for? One of "MonadCoef", "DyadCoef", "BlockModel" or "All" (the default).
...	Currently ignored

**Value**

For param="All", named list of individual return components. For all other values of param, a numeric covariance matrix.

**Author(s)**

Santiago Olivella (olivella@unc.edu), Adeline Lo (aylo@wisc.edu), Tyler Pratt (tyler.pratt@yale.edu),  
Kosuke Imai (imai@harvard.edu)

**Examples**

```
library(NetMix)
## Load datasets
data("lazega_dyadic")
data("lazega_monadic")
## Estimate model with 2 groups
lazega_mmsbm <- mmsbm(SocializeWith ~ Coworkers,
  ~ School + Practice + Status,
  senderID = "Lawyer1",
  receiverID = "Lawyer2",
  nodeID = "Lawyer",
  data.dyad = lazega_dyadic,
  data.monad = lazega_monadic,
  n.blocks = 2,
  mmsbm.control = list(seed = 123,
    conv_tol = 1e-2,
    se_sim = 2)) # Usually requires more samples.

vcov(lazega_mmsbm, "MonadCoef")
```



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