

# Package ‘BayesianGLasso’

January 20, 2025

**Title** Bayesian Graphical Lasso

**Version** 0.2.0

**Description** Implements a data-augmented block Gibbs sampler for simulating the posterior distribution of concentration matrices for specifying the topology and parameterization of a Gaussian Graphical Model (GGM). This sampler was originally proposed in Wang (2012) <doi:10.1214/12-BA729>.

**Depends** R (>= 3.0.0)

**License** GPL-3

**Encoding** UTF-8

**LazyData** true

**Imports** statmod, MASS

**RoxygenNote** 6.0.1

**NeedsCompilation** no

**Author** Patrick Trainor [aut, cre],  
Hao Wang [aut]

**Maintainer** Patrick Trainor <patrick.trainor@louisville.edu>

**Repository** CRAN

**Date/Publication** 2017-07-19 10:52:36 UTC

## Contents

blockGLasso . . . . .	2
<b>Index</b>	<b>4</b>

---

 blockGLasso

*Block Gibbs sampler function*


---

**Description**

Blockwise sampling from the conditional distribution of a permuted column/row for simulating the posterior distribution for the concentration matrix specifying a Gaussian Graphical Model

**Usage**

```
blockGLasso(X, iterations = 2000, burnIn = 1000, lambdaPriora = 1,
  lambdaPriorb = 1/10, verbose = TRUE)
```

**Arguments**

X	Data matrix
iterations	Length of Markov chain after burn-in
burnIn	Number of burn-in iterations
lambdaPriora	Shrinkage hyperparameter (lambda) gamma distribution shape
lambdaPriorb	Shrinkage hyperparameter (lambda) gamma distribution scale
verbose	logical; if TRUE return MCMC progress

**Details**

Implements the block Gibbs sampler for the Bayesian graphical lasso introduced in Wang (2012). Samples from the conditional distribution of a permuted column/row for simulating the posterior distribution for the concentration matrix specifying a Gaussian Graphical Model

**Value**

Sigma	List of covariance matrices from the Markov chain
Omega	List of concentration matrices from the Markov chains
Lambda	Vector of simulated lambda parameters

**Author(s)**

Patrick Trainor (University of Louisville)  
Hao Wang

**References**

Wang, H. (2012). Bayesian graphical lasso models and efficient posterior computation. *Bayesian Analysis*, 7(4). <doi:10.1214/12-BA729> .

**Examples**

```
# Generate true covariance matrix:
s<-.9**toeplitz(0:9)
# Generate multivariate normal distribution:
set.seed(5)
x<-MASS::mvrnorm(n=100,mu=rep(0,10),Sigma=s)
blockGLasso(X=x)
```

```
# Same example with short MCMC chain:
s<-.9**toeplitz(0:9)
set.seed(6)
x<-MASS::mvrnorm(n=100,mu=rep(0,10),Sigma=s)
blockGLasso(X=x,iterations=100,burnIn=100)
```

# Index

blockGLasso, [2](#)