

# Package ‘LassoHiDFastGibbs’

January 18, 2026

**Type** Package

**Title** Fast High-Dimensional Gibbs Samplers for Bayesian Lasso Regression

**Version** 0.1.0

**Description** Provides fast and scalable Gibbs sampling algorithms for Bayesian Lasso regression model in high-dimensional settings. The package implements efficient partially collapsed and nested Gibbs samplers for Bayesian Lasso, with a focus on computational efficiency when the number of predictors is large relative to the sample size.

**License** GPL-3

**Encoding** UTF-8

**SystemRequirements** C++17

**Imports** Rcpp

**LinkingTo** Rcpp, RcppArmadillo, RcppEigen, RcppNumerical, RcppClock

**RoxygenNote** 7.3.2

**Suggests** posterior

**URL** <https://github.com/MJDavoudabadi/LassoHiDFastGibbs>

**BugReports** <https://github.com/MJDavoudabadi/LassoHiDFastGibbs/issues>

**NeedsCompilation** yes

**Author** John Ormerod [aut] (<<https://orcid.org/0000-0002-4650-7507>>),  
Mohammad Javad Davoudabadi [aut, cre, cph],  
Garth Tarr [aut] (<<https://orcid.org/0000-0002-6605-7478>>),  
Samuel Mueller [aut] (<<https://orcid.org/0000-0002-3087-8127>>)

**Maintainer** Mohammad Javad Davoudabadi <mohammad.davoudabadi@qut.edu.au>

**Archs** x64

## Contents

blasso_gibbs_2block_bl . . . . .	2
blasso_gibbs_2block_bs . . . . .	3
blasso_pcg_lambda2_va . . . . .	4
blasso_pcg_sigma2_va . . . . .	6
LassoHiDFastGibbs . . . . .	7

normalize . . . . .	8
penalized_nested_Gibbs . . . . .	9
penalized_pcg_beta_sigma2 . . . . .	10
penalized_pcg_lambda2_sigma2 . . . . .	12
penalized_pcg_sigma2_beta . . . . .	13
penalized_pcg_sigma2_lambda2 . . . . .	14

<b>Index</b>	<b>17</b>
--------------	-----------

---

blasso\_gibbs\_2block\_bl

*Bayesian lasso Gibbs sampler: 2-block (beta-lambda2) variant*

---

## Description

Implements a two-block Gibbs sampler for the Bayesian lasso regression model in which the regression coefficients are updated jointly with the global shrinkage parameter  $\lambda^2$  in one block, while the noise variance and local shrinkage parameters are updated conditionally in separate steps.

## Usage

```
blasso_gibbs_2block_bl(
  vy,
  mX,
  a,
  b,
  u,
  v,
  nsamples,
  lambda_init = 1,
  sigma2_init = 1,
  va_init = NULL,
  verbose = max(1L, floor(nsamples/5)),
  lower = 1e-12,
  upper = 5000
)
```

## Arguments

vy	Numeric response vector of length n.
mX	Numeric design matrix of dimension n x p.
a, b	Hyperparameters for the inverse-gamma prior on $\sigma^2$ .
u, v	Hyperparameters for the prior on $\lambda^2$ .
nsamples	Integer number of MCMC iterations.
lambda_init	Initial value for lambda.
sigma2_init	Initial value for $\sigma^2$ .
va_init	Optional initial values for local shrinkage parameters (length p).
verbose	Print progress every verbose iterations (0 = silent).
lower, upper	Bounds used by the slice sampler for $\lambda^2$ .

**Value**

A list with components:

**mBeta** Matrix of beta draws (nsamples x p).

**vsigma2** Vector of  $\sigma^2$  draws (length nsamples).

**vlambda2** Vector of  $\lambda^2$  draws (length nsamples).

**Examples**

```
set.seed(1)
n <- 30; p <- 6
X <- matrix(rnorm(n * p), n, p)
y <- rnorm(n)
out <- blasso_gibbs_2block_bl(
  vy = y, mX = X,
  a = 1, b = 1, u = 1, v = 1,
  nsamples = 200,
  lambda_init = 1, sigma2_init = 1,
  verbose = 0
)
str(out)
```

---

blasso\_gibbs\_2block\_bs

*Bayesian lasso Gibbs sampler: 2-block (beta-sigma2) variant*

---

**Description**

Implements a two-block Gibbs sampler for the Bayesian lasso regression model in which the regression coefficients are updated jointly with the noise variance  $\sigma^2$  in one block, while the global shrinkage parameter and local shrinkage parameters are updated conditionally in separate steps.

**Usage**

```
blasso_gibbs_2block_bs(
  vy,
  mX,
  a,
  b,
  u,
  v,
  nsamples,
  lambda_init = 1,
  sigma2_init = 1,
  verbose = max(1L, floor(nsamples/5))
)
```

**Arguments**

<code>vy</code>	Numeric response vector of length <code>n</code> .
<code>mX</code>	Numeric design matrix of dimension <code>n</code> x <code>p</code> .
<code>a, b</code>	Hyperparameters for the inverse-gamma prior on $\sigma^2$ .
<code>u, v</code>	Hyperparameters for the prior on $\lambda^2$ .
<code>nsamples</code>	Integer number of MCMC iterations.
<code>lambda_init</code>	Initial value for $\lambda$ .
<code>sigma2_init</code>	Initial value for $\sigma^2$ .
<code>verbose</code>	Print progress every <code>verbose</code> iterations (0 = silent).

**Value**

A list with components:

- mBeta** Matrix of beta draws (`nsamples` x `p`).
- vsigma2** Vector of  $\sigma^2$  draws (length `nsamples`).
- vlambda2** Vector of  $\lambda^2$  draws (length `nsamples`).

**Examples**

```
set.seed(1)
n <- 30; p <- 6
X <- matrix(rnorm(n * p), n, p)
y <- rnorm(n)
out <- blasso_gibbs_2block_bs(
  vy = y, mX = X,
  a = 1, b = 1, u = 1, v = 1,
  nsamples = 200,
  lambda_init = 1, sigma2_init = 1,
  verbose = 0
)
str(out)
```

---

`blasso_pcg_lambda2_va` *Bayesian lasso PCG sampler: lambda2 collapsed over local scales*

---

**Description**

Lasso-specific Partially-collapsed Gibbs (PCG) variant with the local scales (va) collapsed in the  $\lambda^2$  update.

**Usage**

```
blasso_pcg_lambda2_va(
  vy,
  mX,
  a,
  b,
```

```

    u,
    v,
    nsamples,
    lambda_init = 1,
    sigma2_init = 1,
    verbose = max(1L, floor(as.integer(nsamples)/10))
  )

```

## Arguments

<code>vy</code>	Numeric response vector of length <code>n</code> .
<code>mX</code>	Numeric design matrix of dimension <code>n</code> x <code>p</code> .
<code>a, b</code>	Hyperparameters for the inverse-gamma prior on $\sigma^2$ .
<code>u, v</code>	Hyperparameters for the prior on $\lambda^2$ .
<code>nsamples</code>	Number of MCMC iterations.
<code>lambda_init</code>	Initial value for $\lambda$ .
<code>sigma2_init</code>	Initial value for $\sigma^2$ .
<code>verbose</code>	Print progress every verbose iterations (0 = silent).

## Value

A list with components:

**mBeta** Matrix of beta draws (nsamples x p).  
**vsigma2** Vector of sigma<sup>2</sup> draws (length nsamples).  
**vlambda2** Vector of lambda<sup>2</sup> draws (length nsamples).

## Examples

```

set.seed(1)
n <- 40; p <- 6
X <- matrix(rnorm(n * p), n, p)
beta <- c(1.2, 2, -1, 0.5, 0.75, 2.5)
y <- X %*% beta + rnorm(n)

out <- blasso_pcg_lambda2_va(
  vy = y, mX = X,
  a = 1, b = 1, u = 1, v = 1,
  nsamples = 200,
  lambda_init = 1, sigma2_init = 1,
  verbose = 0
)

summary(out$vlambda2)

```

---

blasso\_pcg\_sigma2\_va    *Bayesian lasso PCG sampler: sigma2 collapsed over local scales*

---

## Description

Lasso-specific Partially-collapsed Gibbs (PCG) variant with the local scales (va) collapsed in the  $\sigma^2$  update.

## Usage

```
blasso_pcg_sigma2_va(
  vy,
  mX,
  a,
  b,
  u,
  v,
  nsamples,
  lambda_init = 1,
  sigma2_init = 1,
  va_init = NULL,
  verbose = max(1L, floor(as.integer(nsamples)/10)),
  lower = 1e-12,
  upper = 5000
)
```

## Arguments

vy	Numeric response vector of length n.
mX	Numeric design matrix of dimension n x p.
a, b	Hyperparameters for the inverse-gamma prior on $\sigma^2$ .
u, v	Hyperparameters for the prior on $\lambda^2$ .
nsamples	Number of MCMC iterations.
lambda_init	Initial value for $\lambda$ .
sigma2_init	Initial value for $\sigma^2$ .
va_init	Optional initial local-scale vector (length p). If NULL, defaults to rep(1, ncol(mX)).
verbose	Print progress every verbose iterations (0 = silent).
lower, upper	Bounds used by the slice sampler.

## Value

A list with components:

**mBeta** Matrix of beta draws (nsamples x p).  
**vsigma2** Vector of  $\sigma^2$  draws (length nsamples).  
**vlambda2** Vector of  $\lambda^2$  draws (length nsamples).

## Examples

```
set.seed(1)
n <- 40; p <- 6
X <- matrix(rnorm(n * p), n, p)
beta <- c(1.2, 2, -1, 0.5, 0.75, 2.5)
y <- X %*% beta + rnorm(n)

out <- blasso_pcg_sigma2_va(
  vy = y, mX = X,
  a = 1, b = 1, u = 1, v = 1,
  nsamples = 200,
  lambda_init = 1, sigma2_init = 1,
  va_init = rep(1, p),
  verbose = 0
)

summary(out$vsigma2)
```

---

LassoHiDFastGibbs

*Fast High-Dimensional Gibbs Samplers for Bayesian Lasso Regression*

---

## Description

Provides fast and scalable Gibbs sampling algorithms for Bayesian Lasso regression model in high-dimensional settings. The package implements efficient partially collapsed and nested Gibbs samplers for Bayesian Lasso, with a focus on computational efficiency when the number of predictors is large relative to the sample size.

## Author(s)

**Maintainer:** Mohammad Javad Davoudabadi <mohammad.davoudabadi@qut.edu.au> [copyright holder]

Authors:

- John Ormerod <john.ormerod@sydney.edu.au> ([ORCID](#))
- Garth Tarr <garth.tarr@gmail.com> ([ORCID](#))
- Samuel Mueller <samuel.mueller@sydney.edu.au> ([ORCID](#))

## See Also

Useful links:

- <https://github.com/MJDavoudabadi/LassoHiDFastGibbs>
- Report bugs at <https://github.com/MJDavoudabadi/LassoHiDFastGibbs/issues>

normalize

*Normalize Response and Covariates*

---

**Description**

This function centers and (optionally) scales the response vector and each column of the design matrix using the population variance. It is used to prepare data for Bayesian Lasso regression.

**Usage**

```
normalize(y, X, scale = TRUE)
```

**Arguments**

y	A numeric response vector.
X	A numeric matrix or data frame of covariates (design matrix).
scale	Logical; if TRUE, variables are scaled to have unit population variance (default is TRUE).

**Value**

A list with the following elements:

- vy: Normalized response vector.
- mX: Normalized design matrix.
- mu.y: Mean of the response vector.
- sigma2.y: Population variance of the response vector.
- mu.x: Vector of column means of X.
- sigma2.x: Vector of population variances for columns of X.

**Examples**

```
set.seed(1)
X <- matrix(rnorm(100 * 10), 100, 10)
beta <- c(2, -3, rep(0, 8))
y <- as.vector(X %*% beta + rnorm(100))
norm_result <- normalize(y, X)
```



---

penalized\_nested\_Gibbs

*Penalized nested Gibbs sampler for Bayesian linear regression*


---

## Description

Runs the nested Gibbs sampler for a Gaussian linear model  $y = X\beta + \epsilon$  with either a lasso or horseshoe penalty (shrinkage prior) on  $\beta$ . The algorithm supports both  $n \geq p$  and  $p > n$  regimes.

## Usage

```
penalized_nested_Gibbs(
  vy,
  mX,
  penalty_type = c("lasso", "horseshoe"),
  a,
  b,
  u,
  v,
  nsamples,
  lambda_init = 1,
  va_init = NULL,
  verbose = max(1L, floor(as.integer(nsamples)/10)),
  lower = 1e-12,
  upper = 5000,
  s_beta = 1L,
  s_siglam = 1L
)
```

## Arguments

vy	Numeric response vector of length $n$ .
mX	Numeric design matrix of dimension $n \times p$ .
penalty_type	Character string: either "lasso" or "horseshoe".
a, b	Hyperparameters for the inverse-gamma prior on $\sigma^2$ .
u, v	Hyperparameters for the prior on $\lambda^2$ .
nsamples	Integer number of outer MCMC iterations.
lambda_init	Initial value for $\lambda$ .
va_init	Optional initial values for the local shrinkage parameters (length $p$ ). If NULL, a vector of ones is used.
verbose	Print progress every verbose iterations (0 = silent).
lower, upper	Bounds for the slice sampler used for $\lambda^2$ .
s_beta	Integer: number of inner updates of $\beta$ per outer iteration.
s_siglam	Integer: number of inner updates of $(\sigma^2, \lambda^2)$ per $\beta$ update.

**Value**

A list with components:

**mBeta** Matrix of sampled  $\beta$  draws (rows correspond to stored draws).

**vsigma2** Vector of sampled  $\sigma^2$  draws.

**vlambda2** Vector of sampled  $\lambda^2$  draws.

lm\_penalized\_nested\_gibbs().

**Examples**

```
set.seed(1)
n <- 50; p <- 10
X <- matrix(rnorm(n * p), n, p)
y <- rnorm(n)

out <- penalized_nested_Gibbs(
  vy = y, mX = X,
  penalty_type = "lasso",
  a = 1, b = 1, u = 1, v = 1,
  nsamples = 200,
  lambda_init = 1,
  va_init = NULL,
  verbose = 0,
  lower = 1e-12,
  upper = 5000,
  s_beta = 1,
  s_siglam = 1
)
str(out)
```

---

penalized\_pcg\_beta\_sigma2

*Penalized PCG sampler: beta block, lambda2 collapsed over sigma2*

---

**Description**

Partially-collapsed Gibbs (PCG) sampler variant that updates  $\beta$  in a dedicated block and samples  $\lambda^2$  using a collapsed step over  $\sigma^2$ .

**Usage**

```
penalized_pcg_beta_sigma2(
  vy,
  mX,
  penalty_type = c("lasso", "horseshoe"),
  a,
  b,
  u,
  v,
  nsamples,
```

```

lambda_init = 1,
sigma2_init = 1,
verbose = max(1L, floor(as.integer(nsamples)/10))
)

```

### Arguments

<code>vy</code>	Numeric response vector of length <code>n</code> .
<code>mX</code>	Numeric design matrix of dimension <code>n</code> x <code>p</code> .
<code>penalty_type</code>	Character string: "lasso" or "horseshoe".
<code>a, b</code>	Hyperparameters for the inverse-gamma prior on $\sigma^2$ .
<code>u, v</code>	Hyperparameters for the prior on $\lambda^2$ .
<code>nsamples</code>	Number of MCMC iterations.
<code>lambda_init</code>	Initial value for $\lambda$ .
<code>sigma2_init</code>	Initial value for $\sigma^2$ .
<code>verbose</code>	Print progress every verbose iterations (0 = silent).

### Value

A list with components:

**mBeta** Matrix of beta draws (`nsamples` x `p`).

**vsigma2** Vector of  $\sigma^2$  draws (length `nsamples`).

**vlambda2** Vector of  $\lambda^2$  draws (length `nsamples`).

### Examples

```

set.seed(1)
n <- 40; p <- 6
X <- matrix(rnorm(n * p), n, p)
beta <- c(1.2, 2, -1, 0.5, 0.75, 2.5)
y <- X %*% beta + rnorm(n)

out <- penalized_pcg_beta_sigma2(
  vy = y, mX = X, penalty_type = "horseshoe",
  a = 1, b = 1, u = 1, v = 1,
  nsamples = 200,
  lambda_init = 1, sigma2_init = 1,
  verbose = 0
)

summary(out$mBeta)

```

---

penalized\_pcg\_lambda2\_sigma2

*Penalized PCG sampler: lambda2 collapsed over sigma2*


---

## Description

Partially-collapsed Gibbs (PCG) sampler for a Gaussian linear model with a shrinkage prior/penalty on regression coefficients. This variant samples  $\lambda^2$  using a collapsed step over  $\sigma^2$  (see implementation).

## Usage

```
penalized_pcg_lambda2_sigma2(
  vy,
  mX,
  penalty_type = c("lasso", "horseshoe"),
  a,
  b,
  u,
  v,
  nsamples,
  lambda_init = 1,
  sigma2_init = 1,
  verbose = max(1L, floor(as.integer(nsamples)/10))
)
```

## Arguments

vy	Numeric response vector of length n.
mX	Numeric design matrix of dimension n x p.
penalty_type	Character string: "lasso" or "horseshoe".
a, b	Hyperparameters for the inverse-gamma prior on $\sigma^2$ .
u, v	Hyperparameters for the prior on $\lambda^2$ .
nsamples	Number of MCMC iterations.
lambda_init	Initial value for $\lambda$ .
sigma2_init	Initial value for $\sigma^2$ .
verbose	Print progress every verbose iterations (0 = silent).

## Value

A list with components:

**mBeta** Matrix of beta draws (nsamples x p).  
**vsigma2** Vector of sigma<sup>2</sup> draws (length nsamples).  
**vlambda2** Vector of lambda<sup>2</sup> draws (length nsamples).

**Examples**

```

set.seed(1)
n <- 40; p <- 6
X <- matrix(rnorm(n * p), n, p)
beta <- c(1.2, 2, -1, 0.5, 0.75, 2.5)
y <- X %*% beta + rnorm(n)
out <- penalized_pcg_lambda2_sigma2(
  vy = y, mX = X, penalty_type = "lasso",
  a = 1, b = 1, u = 1, v = 1,
  nsamples = 200, lambda_init = 1, sigma2_init = 1,
  verbose = 0
)
str(out)

```

---

penalized\_pcg\_sigma2\_beta

*Penalized PCG sampler: sigma2 collapsed over beta*


---

**Description**

Partially-collapsed Gibbs (PCG) sampler variant that samples  $\sigma^2$  using a collapsed step over  $\beta$ . Requires initial values for local scales `va_init`.

**Usage**

```

penalized_pcg_sigma2_beta(
  vy,
  mX,
  penalty_type = c("lasso", "horseshoe"),
  a,
  b,
  u,
  v,
  nsamples,
  lambda_init = 1,
  sigma2_init = 1,
  va_init = NULL,
  verbose = max(1L, floor(as.integer(nsamples)/10)),
  lower = 1e-12,
  upper = 5000
)

```

**Arguments**

<code>vy</code>	Numeric response vector of length <code>n</code> .
<code>mX</code>	Numeric design matrix of dimension <code>n x p</code> .
<code>penalty_type</code>	Character string: "lasso" or "horseshoe".
<code>a, b</code>	Hyperparameters for the inverse-gamma prior on $\sigma^2$ .
<code>u, v</code>	Hyperparameters for the prior on $\lambda^2$ .

<code>nsamples</code>	Number of MCMC iterations.
<code>lambda_init</code>	Initial value for $\lambda$ .
<code>sigma2_init</code>	Initial value for $\sigma^2$ .
<code>va_init</code>	Optional initial local-scale vector (length <code>p</code> ). If NULL, defaults to <code>rep(1, ncol(mX))</code> .
<code>verbose</code>	Print progress every verbose iterations (0 = silent).
<code>lower, upper</code>	Bounds used by the slice sampler.

## Value

A list with components:

**mBeta** Matrix of beta draws (`nsamples` x `p`).

**vsigma2** Vector of  $\sigma^2$  draws (length `nsamples`).

**vlambda2** Vector of  $\lambda^2$  draws (length `nsamples`).

## Examples

```
set.seed(1)
n <- 40; p <- 6
X <- matrix(rnorm(n * p), n, p)
beta <- c(1.2, 2, -1, 0.5, 0.75, 2.5)
y <- X %*% beta + rnorm(n)

out <- penalized_pcg_sigma2_beta(
  vy = y, mX = X, penalty_type = "horseshoe",
  a = 1, b = 1, u = 1, v = 1,
  nsamples = 200,
  lambda_init = 1, sigma2_init = 1,
  va_init = rep(1, p),
  verbose = 0
)

summary(out$vsigma2)
```

---

penalized\_pcg\_sigma2\_lambda2

*Penalized PCG sampler: sigma2 collapsed over lambda2*

---

## Description

Partially-collapsed Gibbs (PCG) sampler variant that samples  $\sigma^2$  using a collapsed step over  $\lambda^2$  (see implementation). Requires initial values for local scales `va_init`; if omitted, it is set to a vector of ones.

**Usage**

```
penalized_pcg_sigma2_lambda2(
  vy,
  mX,
  penalty_type = c("lasso", "horseshoe"),
  a,
  b,
  u,
  v,
  nsamples,
  lambda_init = 1,
  sigma2_init = 1,
  va_init = NULL,
  verbose = max(1L, floor(as.integer(nsamples)/10)),
  lower = 1e-12,
  upper = 5000
)
```

**Arguments**

vy	Numeric response vector of length n.
mX	Numeric design matrix of dimension n x p.
penalty_type	Character string: "lasso" or "horseshoe".
a, b	Hyperparameters for the inverse-gamma prior on $\sigma^2$ .
u, v	Hyperparameters for the prior on $\lambda^2$ .
nsamples	Number of MCMC iterations.
lambda_init	Initial value for $\lambda$ .
sigma2_init	Initial value for $\sigma^2$ .
va_init	Optional initial local-scale vector (length p). If NULL, defaults to rep(1, ncol(mX)).
verbose	Print progress every verbose iterations (0 = silent).
lower, upper	Bounds used by the slice sampler.

**Value**

A list with components:

**mBeta** Matrix of beta draws (nsamples x p).

**vsigma2** Vector of sigma<sup>2</sup> draws (length nsamples).

**vlambda2** Vector of lambda<sup>2</sup> draws (length nsamples).

**Examples**

```
set.seed(1)
n <- 40; p <- 6
X <- matrix(rnorm(n * p), n, p)
beta <- c(1.2, 2, -1, 0.5, 0.75, 2.5)
y <- X %*% beta + rnorm(n)

out <- penalized_pcg_sigma2_lambda2(
  vy = y, mX = X, penalty_type = "lasso",
```

```
a = 1, b = 1, u = 1, v = 1,  
nsamples = 200,  
lambda_init = 1, sigma2_init = 1,  
va_init = rep(1, p),  
verbose = 0  
)  
  
str(out)
```



# Index

`blasso_gibbs_2block_bl`, [2](#)  
`blasso_gibbs_2block_bs`, [3](#)  
`blasso_pcg_lambda2_va`, [4](#)  
`blasso_pcg_sigma2_va`, [6](#)  
  
`LassoHiDFastGibbs`, [7](#)  
`LassoHiDFastGibbs-package`  
    (`LassoHiDFastGibbs`), [7](#)  
  
`normalize`, [8](#)  
  
`penalized_nested_Gibbs`, [9](#)  
`penalized_pcg_beta_sigma2`, [10](#)  
`penalized_pcg_lambda2_sigma2`, [12](#)  
`penalized_pcg_sigma2_beta`, [13](#)  
`penalized_pcg_sigma2_lambda2`, [14](#)