

# Package ‘ScaleSpikeSlab’

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**Type** Package

**Title** Scalable Spike-and-Slab

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**Description** A scalable Gibbs sampling implementation for high dimensional Bayesian regression with the continuous spike-and-slab prior. Niloy Biswas, Lester Mackey and Xiao-Li Meng, “Scalable Spike-and-Slab” (2022) <[arXiv:2204.01668](https://arxiv.org/abs/2204.01668)>.

**License** GPL (>= 2)

**Imports** Rcpp, stats, TruncatedNormal

**LinkingTo** Rcpp, RcppEigen

**RoxygenNote** 7.1.2

**NeedsCompilation** yes

**Author** Niloy Biswas [aut, cre] (<<https://orcid.org/0000-0001-9081-5702>>),  
Lester Mackey [aut],  
Xiao-Li Meng [aut]

**Maintainer** Niloy Biswas <[niloy\\_biswas@harvard.edu](mailto:niloy_biswas@harvard.edu)>

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riboflavin

*Riboflavin GWAS dataset*

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### Description

Dataset of riboflavin production by *Bacillus subtilis* containing  $n = 71$  observations of a one-dimensional response (riboflavin production) and  $p = 4088$  predictors (gene expressions). The one-dimensional response corresponds to riboflavin production.

### Usage

```
data(riboflavin)
```

### Format

A data frame containing a vector  $y$  of length 71 (responses) and a matrix  $X$  of dimension 71 by 4088 (gene expressions)

### Details

The processed dataset is the same as in the R packages `qut` and `hdi`.

### References

Buhlmann, P., Kalisch, M. and Meier, L. (2014) *High-dimensional statistics with a view towards applications in biology*. Annual Review of Statistics and its Applications **1**, 255–278

### Examples

```
data(riboflavin)
y <- as.vector(riboflavin$y)
X <- as.matrix(riboflavin$x)
```

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spike\_slab\_linear

*spike\_slab\_linear*

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### Description

Generates Markov chain targeting the posterior corresponding to Bayesian linear regression with spike and slab priors

**Usage**

```

spike_slab_linear(
  chain_length,
  X,
  y,
  tau0,
  tau1,
  q,
  a0 = 1,
  b0 = 1,
  rinit = NULL,
  verbose = FALSE,
  burnin = 0,
  store = TRUE,
  Xt = NULL,
  XXt = NULL,
  tau0_inverse = NULL,
  tau1_inverse = NULL
)

```

**Arguments**

chain_length	Markov chain length
X	matrix of length n by p
y	Response
tau0	prior hyperparameter (non-negative real)
tau1	prior hyperparameter (non-negative real)
q	prior hyperparameter (strictly between 0 and 1)
a0	prior hyperparameter (non-negative real)
b0	prior hyperparameter (non-negative real)
rinit	initial distribution of Markov chain (default samples from the prior)
verbose	print iteration of the Markov chain (boolean)
burnin	chain burnin (non-negative integer)
store	store chain trajectory (boolean)
Xt	Pre-calculated transpose of X
XXt	Pre-calculated matrix $X \cdot \text{transpose}(X)$ (n by n matrix)
tau0_inverse	Pre-calculated matrix $\text{inverse}(I + \tau_0^2 \cdot XXt)$ (n by n matrix)
tau1_inverse	Pre-calculated matrix $\text{inverse}(I + \tau_1^2 \cdot XXt)$ (n by n matrix)

**Value**

Output from Markov chain targeting the posterior corresponding to Bayesian linear regression with spike and slab priors

**Examples**

```

# Synthetic dataset
syn_data <- synthetic_data(n=100,p=200,s0=5,error_std=2,type='linear')
X <- syn_data$X
y <- syn_data$y

# Hyperparamters
params <- spike_slab_params(n=nrow(X),p=ncol(X))

# Run S^3
sss_chain <- spike_slab_linear(chain_length=4e3,burnin=1e3,X=X,y=y,
tau0=params$tau0,tau1=params$tau1,q=params$q,a0=params$a0,b0=params$b0,
verbose=FALSE,store=FALSE)

# Use posterior probabilities for variable selection
sss_chain$z_ergodic_avg[1:10]

```

---

spike\_slab\_logistic    *spike\_slab\_logistic*

---

**Description**

Generates Markov chain targeting the posterior corresponding to Bayesian logistic regression with spike and slab priors

**Usage**

```

spike_slab_logistic(
  chain_length,
  X,
  y,
  tau0,
  tau1,
  q,
  rinit = NULL,
  verbose = FALSE,
  burnin = 0,
  store = TRUE,
  Xt = NULL,
  XXt = NULL
)

```

**Arguments**

chain_length	Markov chain length
X	matrix of length n by p
y	Response

tau0	prior hyperparameter (non-negative real)
tau1	prior hyperparameter (non-negative real)
q	prior hyperparameter (strictly between 0 and 1)
rinit	initial distribution of Markov chain (default samples from the prior)
verbose	print iteration of the Markov chain (boolean)
burnin	chain burnin (non-negative integer)
store	store chain trajectory (boolean)
Xt	Pre-calculated transpose of X
XXt	Pre-calculated matrix $X \times \text{transpose}(X)$ (n by n matrix)

**Value**

Output from Markov chain targeting the posterior corresponding to Bayesian logistic regression with spike and slab priors

**Examples**

```
# Synthetic dataset
syn_data <- synthetic_data(n=100,p=200,s0=5,error_std=2,type='logistic')
X <- syn_data$X
y <- syn_data$y

# Hyperparamters
params <- spike_slab_params(n=nrow(X),p=ncol(X))

# Run S^3
sss_chain <- spike_slab_logistic(chain_length=4e3,burnin=1e3,X=X,y=y,
tau0=params$tau0,tau1=params$tau1,q=params$q,verbose=FALSE,store=FALSE)

# Use posterior probabilities for variable selection
sss_chain$z_ergodic_avg[1:10]
```

---

spike\_slab\_params      *spike\_slab\_params*

---

**Description**

Generates hyperparameters for spike-and-slab

**Usage**

```
spike_slab_params(n, p)
```

**Arguments**

n	number of observations
p	number of covariates

**Value**

spike-and-slab hyperparameters  $q$ ,  $\tau_0$ ,  $\tau_1$ ,  $a_0$ ,  $b_0$

**Examples**

```
hyper_params <- spike_slab_params(n=100,p=200)
print(hyper_params)
```

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spike\_slab\_probit      *spike\_slab\_probit*

---

**Description**

Generates Markov chain targeting the posterior corresponding to Bayesian probit regression with spike and slab priors

**Usage**

```
spike_slab_probit(
  chain_length,
  X,
  y,
  tau0,
  tau1,
  q,
  rinit = NULL,
  verbose = FALSE,
  burnin = 0,
  store = TRUE,
  Xt = NULL,
  XXt = NULL,
  tau0_inverse = NULL,
  tau1_inverse = NULL
)
```

**Arguments**

chain_length	Markov chain length
X	matrix of length $n$ by $p$
y	Response
tau0	prior hyperparameter (non-negative real)
tau1	prior hyperparameter (non-negative real)
q	prior hyperparameter (strictly between 0 and 1)
rinit	initial distribution of Markov chain (default samples from the prior)
verbose	print iteration of the Markov chain (boolean)

burnin	chain burnin (non-negative integer)
store	store chain trajectory (boolean)
Xt	Pre-calculated transpose of X
XXt	Pre-calculated matrix $X \cdot \text{transpose}(X)$ (n by n matrix)
tau0_inverse	Pre-calculated matrix inverse $(I + \text{tau}0^2 \cdot XXt)$ (n by n matrix)
tau1_inverse	Pre-calculated matrix inverse $(I + \text{tau}1^2 \cdot XXt)$ (n by n matrix)

**Value**

Output from Markov chain targeting the posterior corresponding to Bayesian logistic regression with spike and slab priors

**Examples**

```
# Synthetic dataset
syn_data <- synthetic_data(n=100,p=200,s0=5,error_std=2,type='probit')
X <- syn_data$X
Xt <- t(X)
y <- syn_data$y

# Hyperparameters
params <- spike_slab_params(n=nrow(X),p=ncol(X))

# Run S^3
sss_chain <- spike_slab_probit(chain_length=4e3,burnin=1e3,X=X,y=y,
tau0=params$tau0,tau1=params$tau1,q=params$q,verbose=FALSE,store=FALSE)

# Use posterior probabilities for variable selection
sss_chain$z_ergodic_avg[1:10]
```

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synthetic_data	<i>synthetic_data</i>
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**Description**

Generates synthetic linear and logistic regression data

**Usage**

```
synthetic_data(
  n,
  p,
  s0,
  error_std,
  type = "linear",
  scale = TRUE,
  signal = "constant"
)
```

**Arguments**

n	number of observations
p	number of covariates
s0	sparsity (number of non-zero components of the true signal)
error_std	Standard deviation of the Gaussian noise (linear regression only)
type	dataset type ('linear' or 'logistic')
scale	design matrix X has columns mean zero and standard deviation 1 (TRUE or FALSE)
signal	non-zero components of the true signal ('constant' or 'deacy')

**Value**

Design matrix, response and true signal vector for linear and logistic regression

**Examples**

```
syn_data <- synthetic_data(n=100,p=200,s0=5,error_std=2)

# syn_data$X is an n by p design matrix
dim(syn_data$X)

# syn_data$y is a length n response vector
length(syn_data$y)

# syn_data$true_beta is a length n response vector with only the first s0 entries non-zero
all(syn_data$true_beta[1:5]!=0)
all(syn_data$true_beta[-c(1:5)]==0)
```



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