

# Package ‘bastah’

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**Type** Package

**Title** Big Data Statistical Analysis for High-Dimensional Models

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**Repository** CRAN

**Description** Big data statistical analysis for high-dimensional models is made possible by modifying `lasso.proj()` in 'hdi' package by replacing its nodewise-regression with sparse precision matrix computation using 'BigQUIC'.

**License** GPL (== 2)

**LazyData** TRUE

**Enhances** doMC, rPython

**Imports** BigQuic, foreach, glmnet, lars, MASS, stats, Matrix, scalreg

**Depends** R (>= 3.2.0)

**NeedsCompilation** no

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**Description**

Big data statistical analysis for high-dimensional models is made possible by modifying `lasso.proj()` in 'hdi' package by replacing its nodewise-regression with sparse precision matrix computation using 'BigQUIC'.

**Usage**

```
bastah (X, y, categorical = FALSE, family = "gaussian", mcorr = "holm",
        N = 10000, ncores = 4, verbose = FALSE)
```

**Arguments**

X	An n by p numeric design matrix with p columns for p predictor variables and n rows corresponding to n observations.
y	A numeric response variable of length n.
categorical	Type of data in the design matrix. (default = FALSE)
family	Family of the response variable. It should be either "gaussian" or "binomial". (default = "gaussian")
mcorr	Multiple correction method. It can be either "WY" or any of <code>p.adjust.methods</code> . (default = "holm")
N	It is the number of samples to take for the empirical distribution which is used to correct the p-values if multiple correction method is "WY" (Westfall-Young). (default = 10000)
ncores	Maximum number of cores to be used for parallel execution. (default = 4)
verbose	Prints more information if this is set to TRUE. (default = FALSE)

**Details**

In this package `lasso.proj` function of `hdi` package is updated for application on big data. The original `lasso.proj` is updated by replacing node-wise regression with scaled lasso. BigQUIC is used for sparse precision matrix calculation. Data is always normalized before processing. Normalization technique used by Vlaming and Groenen (2014) is used. The method has been successfully used on large SNP (Single Nucleotide Polymorphism) datasets for GWAS (Genomewide Association Study).

The package can use `scikit-learn` (<http://scikit-learn.org>) for a better performance. It is advised to install `doMC`, `rPython`, `python`, `numpy` and `scikit-learn`. The package uses `scikit-learn` at runtime, therefore, `python`, `numpy` and `scikit-learn` are not required for package installation and can be installed after installation of the package.

NOTE: We have noticed that `lars` package in R crashes, so it is recommended to use `scikit-learn`.

NOTE: In preprocessing step, variables having a constant value are not considered. The list of variables used is returned in selection variable of the result.

**Value**

An object with Class "bastah"

pval	Calculated p-values
pval.corr	Corrected p-values
sigmahat	Estimated standard deviation
bhat	Estimated coefficients
selection	Indices of variables selected for analysis

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**References**

C. Hsieh, M. Sustik, I. Dhillon, P. Ravikumar, R. Poldrack. In Neural Information Processing Systems (NIPS), December 2013. (Oral)

S. van de Geer, P. Buhlmann, Y. Ritov and R. Dezeure (2014) On asymptotically optimal confidence regions and tests for high-dimensional models. *Annals of Statistics* 42, 1166-1202.

C. Zhang, S. Zhang(2014) Confidence intervals for low dimensional parameters in high dimensional linear models. *Journal of the Royal Statistical Society: Series B* 76, 217-242.

P. Buhlmann and S. van de Geer(2015) High-dimensional inference in misspecified linear models. *Electronic Journal of Statistics* 9, 1449-1473.

R. de Vlaming and P. J. F. Groenen (2014) The current and future use of ridge regression for prediction in quantitative genetics. *BioMed Research International*, 2015, 143712.

**Examples**

```
# The package is accompanied with a simulated genome-wide association
# study dataset "snps" containing n=100 observations of p=500 predictors
data(snps)
# The association of SNPs to the phenotype can be identified using bastah
# NOTE: We have noticed that lars package in R crashes,
# so it is recommended to use scikit-learn (see package details).
## Not run:
result = bastah(X = snps$X, y = snps$y, family = "binomial", verbose = TRUE)

## End(Not run)
```

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`snps`*SNP data set*

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**Description**

Simulated Single Polymorphism Nucleotide (SNP) dataset containing  $n = 100$  observations of  $p = 500$  predictors (SNPs, 1=Homozygote1, 0=Heterozygote, -1=Homozygote2) and a one-dimensional response (1=case, 0=control). The dataset is generated using GWAsimulator.

**Usage**

```
data(snps)
```

**Format**

**y** Phenotype (1=case, 0=control) of 100 individuals.

**x** SNP genotype data (SNPs, 1=Homozygote1, 0=Heterozygote, -1=Homozygote2) of 500 simulated SNPs.

**References**

C. Li and M. Li (2008) GWAsimulator: a rapid whole-genome simulation program. *Bioinformatics* 24 (1): 140-142.

**Examples**

```
data(snps)
```

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