

# Package ‘hierGWAS’

February 2, 2026

**Title** Assessing statistical significance in predictive GWA studies

**Version** 1.41.0

**Author** Laura Buzdugan

**Maintainer** Laura Buzdugan <buzdugan@stat.math.ethz.ch>

**Description** Testing individual SNPs, as well as arbitrarily large groups of SNPs in GWA studies, using a joint model of all SNPs. The method controls the FWER, and provides an automatic, data-driven refinement of the SNP clusters to smaller groups or single markers.

**Depends** R (>= 3.2.0)

**License** GPL-3

**LazyData** true

**Imports** fastcluster,glmnet, fmsb

**Suggests** BiocGenerics, RUnit, MASS

**biocViews** SNP, LinkageDisequilibrium, Clustering

**Collate** 'cluster.snp.R' 'lasso.select.R' 'multisplit.R' 'MEL.R'

'test.snp.R' 'adj.pval.R' 'comp.cluster.pval.R'

'iterative.DFS.R' 'test.hierarchy.R' 'return.r2.R'

'compute.r2.R'

**git\_url** <https://git.bioconductor.org/packages/hierGWAS>

**git\_branch** devel

**git\_last\_commit** 76af363

**git\_last\_commit\_date** 2025-10-29

**Repository** Bioconductor 3.23

**Date/Publication** 2026-02-01

## Contents

cluster.snp	2
compute.r2	3
hierGWAS	4
multisplit	4
simGWAS	5
test.hierarchy	6

<b>Index</b>	<b>8</b>
--------------	----------

---

cluster.snp	<i>Hierarchical Clustering of SNP Data</i>
-------------	--

---

## Description

Clusters SNPs hierachically.

## Usage

```
cluster.snp(x = NULL, d = NULL, method = "average", SNP_index = NULL)
```

## Arguments

x	The SNP data matrix of size nobs x nvar. Default value is NULL
d	NULL or a dissimilarity matrix. See the 'Details' section.
method	The agglomeration method to be used. This should be (an unambiguous abbreviation of) one of "ward.D", "ward.D2", "single", "complete", "average" (= UPGMA), "mcquitty" (= WPGMA), "median" (= WPGMC) or "centroid" (= UPGMC). See <a href="#">hclust</a> for details.
SNP_index	NULL or the index vector of SNPs to be clustered. See the 'Details' section.

## Details

The SNPs are clustered using [hclust](#), which performs a hierarchical cluster analysis using a set of dissimilarities for the nvar objects being clustered. There are 3 possible scenarios.

If d = NULL, x is used to compute the dissimilarity matrix. The dissimilarity measure between two SNPs is  $1 - LD$  (Linkage Disequilibrium), where LD is defined as the square of the Pearson correlation coefficient. If SNP\_index = NULL, all nvar SNPs will be clustered; otherwise only the SNPs with indices specified by SNP\_index will be considered.

If the user wishes to use a different dissimilarity measure, d needs to be provided. d must be either a square matrix of size nvar x nvar, or an object of class [dist](#). If d is provided, x and SNP\_index will be ignored.

## Value

An object of class [dendrogram](#) which describes the tree produced by the clustering algorithm [hclust](#).

## Examples

```
library(MASS)
x <- mvrnorm(60, mu = rep(0, 60), Sigma = diag(60))
clust.1 <- cluster.snp(x = x, method = "average")
SNP_index <- seq(1, 10)
clust.2 <- cluster.snp(x = x, method = "average", SNP_index = SNP_index)
d <- dist(x)
clust.3 <- cluster.snp(d = d, method = "single")
```

compute.r2

*R2 computation***Description**

Calculates the R2 of a cluster of SNPs.

**Usage**

```
compute.r2(x, y, res.multisplit, covar = NULL, SNP_index = NULL)
```

**Arguments**

**x** The input matrix, of dimension nobs x nvar. Each row represents a subject, each column a SNP.

**y** The response vector. It can be continuous or discrete.

**res.multisplit** The output of `multisplit`.

**covar** NULL or the matrix of covariates one wishes to control for, of size nobs x ncovar.

**SNP\_index** NULL or the index vector of the cluster of SNPs whose R2 will be computed. See the 'Details' section.

**Details**

The R2 of a cluster of SNPs is computed on the second half-samples. The cluster members, are intersected with the SNPs selected by the lasso, and the R2 of this model is calculated. Thus, we obtain B R2 values. Finally, the mean of these values is taken. If the value of `SNP_index` is NULL, the R2 of the full model with all the SNPs will be computed.

**Value**

The R2 value of the SNP cluster

**References**

Buzdugan, L. et al. (2015), Assessing statistical significance in predictive genome-wide association studies. (unpublished)

**Examples**

```
library(MASS)
x <- mvrnorm(60, mu = rep(0, 60), Sigma = diag(60))
beta <- rep(0, 60)
beta[c(5, 9, 3)] <- 1
y <- x %*% beta + rnorm(60)
SNP_index <- c(5, 9, 3)
res.multisplit <- multisplit(x, y)
r2 <- compute.r2(x, y, res.multisplit, SNP_index = SNP_index)
```

---

hierGWASAssessing statistical significance in predictive GWA studies

---

**Description**

Testing individual SNPs, as well as arbitrarily large groups of SNPs in GWA studies, using a joint model of all SNPs. The method controls the FWER, and provides an automatic, data-driven refinement of the SNP clusters to smaller groups or single markers.

**Details**

hierGWAS is a package designed to assess statistical significance in GWA studies, using a hierarchical approach.

There are 4 functions provided: `cluster.snp`, `multisplit`, `test.hierarchy` and `compute.r2`. `cluster.snp` performs the hierarchical clustering of the SNPs, while `multisplit` runs multiple penalized regressions on repeated random subsamples. These 2 functions need to be executed before `test.hierarchy`, which does the hierarchical testing, though the order in which the 2 functions are executed does not matter. `test.hierarchy` provides the final output of the method: a list of SNP groups or individual SNPs, along with their corresponding p-values. Finally, `compute.r2` computes the explained variance of an arbitrary group of SNPs, of any size. This group can encompass all SNPs, SNPs belonging to a certain chromosome, or an individual SNP.

**Author(s)**

Laura Buzdugan laura.buzdugan@stat.math.ethz.ch

**References**

Buzdugan, L. et al. (2015), Assessing statistical significance in predictive genome-wide association studies (unpublished)

---

multisplitVariable Selection on Random Sample Splits.

---

**Description**

Performs repeated variable selection via the lasso on random sample splits.

**Usage**

```
multisplit(x, y, covar = NULL, B = 50)
```

## Arguments

- x The SNP data matrix, of size nobs x nvar. Each row represents a subject, each column a SNP.
- y The response vector. It can be continuous or discrete.
- covar NULL or the matrix of covariates one wishes to control for, of size nobs x ncovar.
- B The number of random splits. Default value is 50.

## Details

The samples are divided into two random splits of approximately equal size. The first subsample is used for variable selection, which is implemented using **glmnet**. The first [nobs/6] variables which enter the lasso path are selected. The procedure is repeated B times.

If one or more covariates are specified, these will be added unpenalized to the regression.

## Value

A data frame with 2 components. A matrix of size B x [nobs/2] containing the second subsample of each split, and a matrix of size B x [nobs/6] containing the selected variables in each split.

## References

Meinshausen, N., Meier, L. and Buhlmann, P. (2009), P-values for high-dimensional regression, Journal of the American Statistical Association 104, 1671-1681.

## Examples

```
library(MASS)
x <- mvrnorm(60, mu = rep(0, 200), Sigma = diag(200))
beta <- rep(1, 200)
beta[c(5, 9, 3)] <- 3
y <- x %*% beta + rnorm(60)
res.multisplit <- multisplit(x, y)
```

---

simGWAS

*Simulated GWAS data*

---

## Description

This data set was simulated using PLINK. Please refer to the vignette for more details.

## Usage

simGWAS

## Format

The dataset contains the following components:

**SNP**.1 The first SNP, of dimension 500 x 1. Each row represents a subject.

...

**SNP**.1000 The last SNP, of dimension 500 x 1. Each row represents a subject.

**y** The response vector. It can be continuous or discrete.

**sex** The first covariate, representing the sex of the subjects: 0 for men and 1 for women.

**age** The second covariate, representing the age of the subjects.

## Value

data.frame

## Examples

```
data(simGWAS)
```

---

test.hierarchy	<i>Hierarchical Testing of SNPs</i>
----------------	-------------------------------------

---

## Description

Performs hierarchical testing of SNPs.

## Usage

```
test.hierarchy(x, y, dendr, res.multisplit, covar = NULL, SNP_index = NULL,
alpha = 0.05, global.test = TRUE, verbose = TRUE)
```

## Arguments

<b>x</b>	The input matrix, of dimension nobs x nvar. Each row represents a subject, each column a SNP.
<b>y</b>	The response vector. It can be continuous or discrete.
<b>dendr</b>	The cluster tree obtained by hierarchically clustering the SNPs using <code>cluster.snp</code> .
<b>res.multisplit</b>	The output of <code>multisplit</code> .
<b>covar</b>	NULL or the matrix of covariates one wishes to control for, of size nobs x ncovar.
<b>SNP_index</b>	NULL or the index vector of SNP to be tested. See the 'Details' section.
<b>alpha</b>	The significance level at which the FWER is controlled. Default value is 0.05.
<b>global.test</b>	Specifies whether the global null hypothesis should be tested. Default value is TRUE. See the 'Details' section.
<b>verbose</b>	Report information on progress. Default value is TRUE

## Details

The testing is performed on the cluster tree given by `dendr`. If the SNP data matrix was divided (e.g. by chromosome), and clustered separately, the user must provide the argument `SNP_index`, to specify which part of the data is being tested.

Testing starts at the highest level, which includes all variables specified by `SNP_index`, and moves down the cluster tree. It stops when a cluster's null hypothesis cannot be rejected anymore. The smallest, still significant clusters will be returned.

By default the parameter `global.test = TRUE`, which means that first the global null hypothesis is tested. If the data is divided (e.g. by chromosome), and clustered separately, this parameter can be set to `FALSE` once the global null has been rejected. This helps save time.

## Value

A list of significant SNP groups with the following components:

<code>SNP_index</code>	The indeces of the SNPs in the group
<code>pval</code>	The p-value of the SNP group

## References

Buzdugan, L. et al. (2015), Assessing statistical significance in predictive genome-wide association studies

## Examples

```
library(MASS)
x <- mvtnorm(60, mu = rep(0, 60), Sigma = diag(60))
beta <- rep(0, 60)
beta[c(5, 9, 3)] <- 1
y <- x %*% beta + rnorm(60)
dendr <- cluster.snp(x = x, method = "average")
res.multisplit <- multisplit(x, y)
sign.clusters <- test.hierarchy(x, y, dendr, res.multisplit)
```

# Index

\* **datasets**  
  simGWAS, 5

  cluster.snp, 2, 4  
  compute.r2, 3, 4

  dendrogram, 2  
  dist, 2

  glmnet, 5

  hclust, 2  
  hierGWAS, 4  
  hierGWAS-package (hierGWAS), 4

  multisplit, 4, 4

  simGWAS, 5

  test.hierarchy, 4, 6