# Package 'hierGWAS'

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Title Assessing statistical significance in predictive GWA studies
Version 1.4.0
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<b>Description</b> 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.
<b>Depends</b> R (>= $3.2.0$ )
License GPL-3
LazyData true
Imports fastcluster,glmnet, fmsb
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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'
NeedsCompilation no
R topics documented:  cluster.snp
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Hierarchical Clustering of SNP Data

### **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 hclust for details.

#### **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")</pre>
```

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compute.r2	R2 computation
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### **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.
у	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)</pre>
```

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hierGWAS	Asessing statistical significance in predictive GWA studies
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### **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.

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### References

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 (2015), Assessing statistical significance in predictive genome-wide association studies (unpublished)

multisplit	Variable Selection on Random Sample Splits.	
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### **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.
У	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.
В	The number of random splits. Default value is 50.

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#### **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  $\times$  [nobs/2] containing the second subsample of each split, and a matrix of size B  $\times$  [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)</pre>
```

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 \,\mathrm{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, represeting the sex of the subjects: 0 for men and 1 for women.

age The second covariate, represeting the age of the subjects.

### Value

data.frame

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### **Examples**

data(simGWAS)

test.hierarchy Hier

Hierarchical Testing of SNPs

#### **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

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.

dendr The cluster tree obtained by hierchically clustering the SNPs using cluster.snp.

res.multisplit The output of multisplit.

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

alpha The significance level at which the FWER is controlled. Default value is 0.05. global.test Specifies wether the global null hypothesis should be tested. Default value is

TRUE. See the 'Details' section.

verbose 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:

SNP\_index The indeces of the SNPs in the group

pval The p-value of the SNP group

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### References

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

### **Examples**

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

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