

# Package ‘RVtests’

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**Type** Package

**Title** Rare Variant Tests

**Version** 1.2

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**Depends** R (>= 2.12.1), glmnet, spls, pls

**Description** Use multiple regression methods to test rare variants association with disease traits.

**License** GPL (>= 2)

**LazyLoad** yes

**NeedsCompilation** no

**Repository** CRAN

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RVtests-package

*Rare Variants Tests*


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## Description

Use multiple regression methods to test rare variants association with disease traits.

## Details

Package: RVtests  
 Type: Package  
 Version: 1.2  
 Date: 2013-05-27  
 License: GLP 2.0 or greater  
 LazyLoad: yes

An overview of how to use the package, including the most important functions

## Author(s)

C. Xu, C. M. Greenwood, Maintainer: <changjiang.xu@mail.mcgill.ca>

## References

Xu C, Ladouceur M, Dastani Z, Richards JB, Ciampi A, Greenwood CMT. (2012) Multiple Regression Methods Show Great Potential for Rare Variant Association Tests. PLoS ONE 7(8): e41694. doi:10.1371/journal.pone.0041694

## Examples

```
data(sample.cgeno)
str(sample.cgeno)
x=count2geno(sample.cgeno$cgeno)
dim(x)

set.seed(31018)
y<- rowSums(x[,2:4]*rep(rnorm(3,1,0.1), each=nrow(x))) + 0.4*rnorm(nrow(x))

tmp<- proc.time();RR(x,y,lambda=0:5); proc.time()-tmp
tmp<- proc.time();RR(x,y,weights=c(rep(2,10), rep(1, ncol(x)-10)), lambda=0:5); proc.time()-tmp
tmp<- proc.time();RR(x,y,weights=c(rep(1,10), rep(0, ncol(x)-10)), lambda=0:5); proc.time()-tmp
```

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count2geno	<i>Transforming genotype counts to genotype codes</i>
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**Description**

Transform genotype counts data format to genotype codes format.

**Usage**

```
count2geno(cgeno, indid)
```

**Arguments**

cgeno	A matrix or data frame with 3 columns: indid (individual IDs), snpid (SNP IDs), and count
indid	Individuals ID, including indid in cgeno

**Value**

A matrix of genotypes

**Author(s)**

C. Xu

**See Also**

[geno2count](#)

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geno2count	<i>Transforming genotype codes matrix to genotype counts</i>
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**Description**

Genotype counts

**Usage**

```
geno2count(genotype)
```

**Arguments**

genotype	Genotype matrix or data frame with row and column names, each row as an individual and each column as a snp
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**Value**

Data frame of genotype counts with 3 columns: `indid` (individual IDs), `snpid` (SNP IDs), and `count`

**Author(s)**

C. Xu

**See Also**

[count2geno](#)

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LASSO

*LASSO for Rare Variant Tests*

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**Description**

Use LASSO for selecting significant variants and testing the variants associated with disease traits.

**Usage**

```
LASSO(x, y, family = c("gaussian", "binomial", "poisson", "multinomial", "cox"),
alpha = 1, nlambda = 100, lambda.min.ratio, standardize = TRUE,
size.max, a = 2, npermutation = 0, npermutation.max, min.nonsignificant.counts)
```

**Arguments**

<code>x</code>	Genotype matrix, each row as an individual and each column as a snp
<code>y</code>	Phenotype vector
<code>family</code>	Family: gaussian, binomial, poisson, multinomial, and cox
<code>alpha</code>	alpha = 1 for LASSO, see <code>glmnet</code>
<code>nlambda</code>	see <code>glmnet</code>
<code>lambda.min.ratio</code>	see <code>glmnet</code>
<code>standardize</code>	see <code>glmnet</code>
<code>size.max</code>	Maximum number of variants included
<code>a</code>	Penalty parameter for information criterion, a=2 for AIC.
<code>npermutation</code>	Number of permutation, if less than 1, the permutation will not be run.
<code>npermutation.max</code>	Maximum permutation
<code>min.nonsignificant.counts</code>	Minimum nonsignificant counts

**Details**

Use glmnet package to implement LASSO and an information criterion (AIC, BIC, or GIC) to select a set of variants.

**Value**

nonsignificant.counts	Counts of permuted data that have a higher score than unpermuted data.
pvalue.empirical	Empirical pvalue via permutation
pvalue.nominal	Not available
vs	The selected variants
total.permutation	Total permutation
family	Family

**Author(s)**

C. Xu

**References**

Xu C, Ladouceur M, Dastani Z, Richards JB, Ciampi A, Greenwood CMT. (2012) Multiple Regression Methods Show Great Potential for Rare Variant Association Tests. PLoS ONE 7(8): e41694. doi:10.1371/journal.pone.0041694

**See Also**

[SPLS](#), [glmnet](#)

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 PCR

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*Principal Components Regression for RV tests*


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**Description**

Use principal components for testing rare variants association with disease traits.

**Usage**

```
PCR(x, y, scale = FALSE, ncomp, varpercent,
    npermutation = 100, npermutation.max, min.nonsignificant.counts)
```

**Arguments**

x	Genotype matrix
y	Phenotype vector
scale	If TRUE, scale x and y.
ncomp	Number of components, which could be a vector containing a set of numbers.
varpercent	Explained variance percentage
npermutation	Number of permutation, if less than 1, the permutation will not be run.
npermutation.max	Maximum permutation
min.nonsignificant.counts	Minimum nonsignificant counts

**Value**

score	Correlation between y and y_est
nonsignificant.counts	Counts of permuted data that have a higher score than unpermuted data.
pvalue.empirical	Empirical pvalue via permutation
pvalue.nominal	Theoretical pvalue, not available now.
total.permutation	Total permutation
ncomp.varp	Number of components required for specified variance percentage

**Author(s)**

C. Xu

**References**

Xu C, Ladouceur M, Dastani Z, Richards JB, Ciampi A, Greenwood CMT. (2012) Multiple Regression Methods Show Great Potential for Rare Variant Association Tests. PLoS ONE 7(8): e41694. doi:10.1371/journal.pone.0041694

**See Also**

[PLS](#), [RR](#)

**Description**

Use PLS components for testing rare variants association with disease traits.

**Usage**

```
PLS(x, y, scale = FALSE, ncomp, varpercent,
    npermutation = 100, npermutation.max, min.nonsignificant.counts)
```

**Arguments**

x	Genotype matrix
y	Phenotype vector
scale	If TRUE, scale x and y.
ncomp	Number of components, which could be a vector containing a set of numbers.
varpercent	Explained variance percentage
npermutation	Number of permutation, if less than 1, the permutation will not be run.
npermutation.max	Maximum permutation
min.nonsignificant.counts	Minimum nonsignificant counts

**Value**

score	Correlation between y and y_est
nonsignificant.counts	Counts of permuted data that have a higher score than unpermuted data.
pvalue.empirical	Empirical pvalue via permutation
pvalue.nominal	Theoretical pvalue, not available now.
total.permutation	Total permutation
ncomp.varp	Number of components required for specified variance percentage

**Author(s)**

C. Xu

## References

Xu C, Ladouceur M, Dastani Z, Richards JB, Ciampi A, Greenwood CMT. (2012) Multiple Regression Methods Show Great Potential for Rare Variant Association Tests. PLoS ONE 7(8): e41694. doi:10.1371/journal.pone.0041694

## See Also

[PCR](#), [SPLS](#)

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 RR

*Ridge Regression for RV Tests*


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## Description

Use ridge regression for testing rare variants association with disease traits.

## Usage

```
RR(x, y, z = NULL, scale = FALSE, weights = 1, lambda = 1,
  npermutation = 1000, npermutation.max, min.nonsignificant.counts = 100)
```

## Arguments

x	Genotype matrix
y	Phenotype vector
z	Covariate matrix
scale	If TRUE, scale x and y.
weights	Genotype weights
lambda	Regularization parameter
npermutation	Number of permutation
npermutation.max	Maximum permutation
min.nonsignificant.counts	Minimum nonsignificant counts

## Value

nonsignificant.counts	Counts of permuted data that have a higher score than unpermuted data.
total.permutation	Total permutation
score	Correlation between y and y_est if z=NULL.
pvalue.empirical	Empirical pvalue via permutation
pvalue.nominal	Theoretical pvalue, not available.



**Author(s)**

C. Xu

**References**

Xu C, Ladouceur M, Dastani Z, Richards JB, Ciampi A, Greenwood CMT. (2012) Multiple Regression Methods Show Great Potential for Rare Variant Association Tests. PLoS ONE 7(8): e41694. doi:10.1371/journal.pone.0041694

**See Also**

[PCR](#), [PLS](#)

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 sample.cgeno

*Genotype Counts dataset*


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**Description**

A list of genotype counts, phenotype, and polyphen weight

**Usage**

```
data(sample.cgeno)
```

**Format**

The format is: List of 3 \$ cgeno :'data.frame': 960 obs. of 3 variables: ..\$ indid: int [1:960] 16929 18167 18168 28671 31308 32037 33182 49716 53138 13206 ... ..\$ snpid: int [1:960] 57548466 57548466 57548466 57548466 57548466 57548466 57548466 57548466 57548466 57550649 ... ..\$ count: int [1:960] 1 1 1 1 1 1 1 1 1 1 ... \$ phen :'data.frame': 262 obs. of 2 variables: ..\$ indid: int [1:262] 32 90 101 109 129 133 225 236 253 349 ... ..\$ trait: num [1:262] 0.128 0.166 0.884 0.929 0.195 ... \$ polyphen.weight:'data.frame': 71 obs. of 2 variables: ..\$ snpid : int [1:71] 57548364 57548466 57550649 57550666 57556205 57556220 57556236 57567762 57569339 57569466 ... ..\$ weight: num [1:71] 0.5 0.5 0.5 0.5 0.055 0 0.706 0.5 0.995 0.5 ...

**Details**

The dataset was used in comparing VT and WOD methods.

**Examples**

```
data(sample.cgeno)
str(sample.cgeno)
```

SPLS

*Sparse PLS for RV Tests***Description**

Use SPLS for selecting significant variants and testing the variants associated with disease traits.

**Usage**

```
SPLS(x, y, scale = TRUE, ncomp, eta.grid, size.max, a = 2,
      npermutation = 0, npermutation.max, min.nonsignificant.counts)
```

**Arguments**

x	Genotype matrix, each row as an individual and each column as a snp
y	Phenotype vector
scale	see spls
ncomp	Number of components
eta.grid	see spls
size.max	Maximum number of variants included
a	Penalty parameter for information criterion, a=2 for AIC.
npermutation	Number of permutation, if less than 1, the permutation will not be run.
npermutation.max	Maximum permutation
min.nonsignificant.counts	Minimum nonsignificant counts

**Details**

Use spls package to implement SPLS and an information criterion (AIC, BIC, GIC) to select a set of variants.

**Value**

nonsignificant.counts	Counts of permuted data that have a higher score than unpermuted data.
pvalue.empirical	Empirical pvalue via permutation
pvalue.nominal	Not available
vs	The selected variants
total.permutation	Total permutation

**Author(s)**

C. Xu

**References**

Xu C, Ladouceur M, Dastani Z, Richards JB, Ciampi A, Greenwood CMT. (2012) Multiple Regression Methods Show Great Potential for Rare Variant Association Tests. PLoS ONE 7(8): e41694. doi:10.1371/journal.pone.0041694

**See Also**

[spls](#), [LASSO](#)

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 VTWOD

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*VT and WOD for RV Tests*


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**Description**

Include methods: T1, T5, WE, VT, and WOD.

**Usage**

```
VTWOD(x, y, polyphen.weight, flipPhenotype = 0,
      npermutation = 1000, npermutation.max, min.nonsignificant.counts)
```

**Arguments**

x	Genotype matrix
y	Phenotype vector
polyphen.weight	Polyphen weight
flipPhenotype	Logical, if TRUE, flip phenotype to opposite by multiplying -1
npermutation	Number of permutation, if less than 1, the permutation will not be run.
npermutation.max	Maximum permutation
min.nonsignificant.counts	Minimum nonsignificant counts

**Value**

score	Scores of T1, T5, WE, VT, and WOD
nonsignificant.counts	Counts of permuted data that have a higher score than unpermuted data.
pvalue.empirical	Empirical pvalue via permutation

pvalue.nominal  
Theoretical pvalue, not available now.  
total.permutation  
Total permutation

**Note**

This R implementation by Adam Kiezun, based on reference implementation in C by Alkes Price. Added WOD tests to the program in 2011 by Celia Greenwood

**Author(s)**

C. Xu, Celia Greenwood

**References**

Xu C, Ladouceur M, Dastani Z, Richards JB, Ciampi A, Greenwood CMT. (2012) Multiple Regression Methods Show Great Potential for Rare Variant Association Tests. PLoS ONE 7(8): e41694. doi:10.1371/journal.pone.0041694

Ladouceur M, Dastani Z, Aulchenko YS, Greenwood CM, Richards JB (2012) The empirical power of rare variant association methods: Results from sanger sequencing in 1,998 individuals. PLoS Genetics 8: e1002496.

Price AL, Kryukov GV, de Bakker PI, Purcell SM, Staples J, et al. (2010) Pooled association tests for rare variants in exon-resequencing studies. Am J Hum Genet 86: 832 - 838.

**See Also**

[RR](#), [PCR](#), [PLS](#)

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