

# Package ‘jointsum’

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**Title** jointsum

**Version** 0.19

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**Description** Joint analysis using summary statistics. Build joint models. Test for pleiotropy or allelic heterogeneity.

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**License** MIT

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conAH	<i>conAH</i>
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## Description

Test Allelic Heterogeneity with summary statistics and intersection-union tests.

**Usage**

```
conAH(Z, ld, n)
```

**Arguments**

**Z** a  $q \times 1$  matrix containing the Z-scores for  $q$  SNPs.  
**ld** a  $q \times q$  correlation matrix of the  $q$  SNPs.  
**n** a  $q \times 1$  matrix containing the sample sizes used to get the Z-scores for the  $q$  SNPs.  
 Can be a number if the  $q$  sample sizes are the same.

**Value**

conAH returns the p-value of testing AH.

**Author(s)**

Yangqing Deng and Wei Pan.

**References**

Deng, Y., Pan, W. (2018b). Significance Testing for Allelic Heterogeneity. Genetics. September 1, 2018 vol. 210 no. 1 25-32; <https://doi.org/10.1534/genetics.118.301111>.

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conAHseq

*conAHseq*


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**Description**

Infer the number of causal SNPs with a sequential procedure using summary statistics and intersection-union tests.

**Usage**

```
conAHseq(Z, ld, n, k=6, alpha=0.05)
```

**Arguments**

**Z** a  $q \times 1$  matrix containing the Z-scores for  $q$  SNPs.  
**ld** a  $q \times q$  correlation matrix of the  $q$  SNPs.  
**n** a  $q \times 1$  matrix containing the sample sizes used to get the Z-scores for the  $q$  SNPs.  
 Can be a number if the  $q$  sample sizes are the same.  
**k** a maximum number of causal SNPs.  
**alpha** the significance threshold.

**Value**

conAHseq returns the p-value of each step as well as the predicted number of causal SNPs.

**Author(s)**

Yangqing Deng and Wei Pan.

## References

Deng, Y., Pan, W. (2018b). Significance Testing for Allelic Heterogeneity. *Genetics*. September 1, 2018 vol. 210 no. 1 25-32; <https://doi.org/10.1534/genetics.118.301111>.

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conAHseq2	<i>conAHseq2</i>
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## Description

A faster version of conAHseq, which considers less cases but gives the causal locations.

## Usage

```
conAHseq2(Z, ld, n, k=6, alpha=0.05)
```

## Arguments

Z	a q*1 matrix containing the Z-scores for q SNPs.
ld	a q*q correlation matrix of the q SNPs.
n	a q*1 matrix containing the sample sizes used to get the Z-scores for the q SNPs. Can be a number if the q sample sizes are the same.
k	a maximum number of causal SNPs.
alpha	the significance threshold.

## Value

conAHseq2 returns the significant causal SNPs, the p-value of each step and the predicted number of causal SNPs.

## Author(s)

Yangqing Deng and Wei Pan.

## References

Deng, Y., Pan, W. (2018b). Significance Testing for Allelic Heterogeneity. *Genetics*. September 1, 2018 vol. 210 no. 1 25-32; <https://doi.org/10.1534/genetics.118.301111>.

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JointSum	<i>JointSum</i>
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### Description

Use summary statistics to build a joint linear model for one trait vs. L SNPs or one trait vs. L SNPs + (K-1) traits.

### Usage

```
JointSum(B1,S1,B2=0,S2=0,N,XX=diag(1,nrow=1),YY0,adj_Y=1,lam=0)
```

### Arguments

B1	a L*1 matrix containing marginal effects on the trait treated as response.
S1	a L*1 matrix containing standard errors for B1.
B2	a L*(K-1) matrix containing marginal effects on the (K-1) traits to adjust for. If K=1, do not specify this.
S2	a L*(K-1) matrix containing standard errors for B2. If K=1, do not specify this.
N	a L*K matrix containing sample sizes for each coefficients in B1, B2.
XX	a L*L estimated covariance matrix for the L SNPs.
YY0	a K*K estimated correlation matrix for the K traits.
adj_Y	whether traits should be adjusted for. If it is 0, adjust for SNPs only. Otherwise adjust for both SNPs and traits.
lam	a modifying parameter in [0,1). It is used only if adj_Y=1.

### Value

beta	coefficient estimates (SNPs first).
cov	the covariance matrix for coefficients.
pvalue	p-values for coefficients.
sigma2	estimated mean squared error.

### Author(s)

Yangqing Deng and Wei Pan.

### References

Deng, Y., Pan, W. (2017). Conditional analysis of multiple quantitative traits based on marginal GWAS summary statistics. Genet Epidemiol. doi: 10.1002/gepi.22046.

## Examples

```
#2 correlated SNPs, 2 traits
set.seed(13)
x1=rbinom(1000,1,0.3)
x2=c(x1[1:300],rbinom(1000-300,1,0.3))
y2=rnorm(1000)+x1
y1=rnorm(1000)+y2/2

#standardization
x1=x1-mean(x1)
x2=x2-mean(x2)
y1=y1-mean(y1)
y2=y2-mean(y2)

#summary statistics
a=summary(lm(y1~x1-1))$coefficients
b=summary(lm(y1~x2-1))$coefficients
c=summary(lm(y2~x1-1))$coefficients
d=summary(lm(y2~x2-1))$coefficients

B1=as.matrix(c(a[1],b[1]))
S1=as.matrix(c(a[2],b[2]))
B2=as.matrix(c(c[1],d[1]))
S2=as.matrix(c(c[2],d[2]))
N=matrix(1000,nrow=2,ncol=2)
XX=cov(cbind(x1,x2))
YY0=cor(cbind(y1,y2))

#model Y1 ~ X1 + X2
JointSum(B1,S1,B2,S2,N,XX,YY0,adj_Y=0)
#or
JointSum(B1,S1,N=N[,1],XX=XX,YY0=diag(1),adj_Y=0)

#model Y1 ~ X1 + X2 + Y2
JointSum(B1,S1,B2,S2,N,XX,YY0,adj_Y=1)

#may compare with joint models using individual level data
summary(lm(y1~x1+x2-1))
summary(lm(y1~x1+x2+y2-1))
```

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Plei

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Plei

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## Description

Test pleiotropy with union-intersection tests and individual level data.

## Usage

```
Plei(X,Y,pr=c(1:(nrow(YY0),nrow(YY0)*nrow(XX)+1),method="wald",ay=1,suby=c())
```

## Arguments

<code>X</code>	a $n \times q$ matrix containing $q$ SNPs for $n$ subjects.
<code>Y</code>	a $n \times p$ matrix containing $p$ traits for $n$ subjects.
<code>pr</code>	a vector indicating which tests should be included in the union-intersection test. It is suggested to only use the default setting, which includes 1, 2, ..., $p$ and $p \times q + 1$ . 1, 2, ..., $p$ correspond to testing whether the first SNP only influences one of the $p$ traits. $p \times q + 1$ correspond to testing whether the first SNP influences none of the $p$ traits.
<code>method</code>	the method to be used for each individual test. <code>method = "lrt"</code> uses the likelihood ratio test as described in Schaid et. al (2016); <code>method = "wald"</code> uses the Wald test with ordinary least square estimates and sandwich covariance matrices. It is recommended to use the Wald test when <code>ay = 1</code> .
<code>ay</code>	whether some of the traits should be adjusted for. <code>ay = 1</code> means to use conditional analysis adjusting for traits (and possibly SNPs if $q$ is bigger than 1); <code>ay = 0</code> means not to adjust for any traits.
<code>suby</code>	a vector indicating which of the traits should NOT be adjusted for. It is only effective when <code>ay = 1</code> . If <code>suby</code> is not specified, all of the traits will be adjusted for in the conditional analysis (which is not recommended). If <code>suby = c(1,3)</code> , the first and the third traits will not be adjusted for. (In the current version, <code>suby</code> only works for <code>method = "wald"</code> ; if <code>method = "lrt"</code> , <code>suby</code> is always considered as <code>c()</code> )

## Value

Plei returns a vector containing 2 p-values. The first p-value corresponds to testing whether the SNP of interest only influences one of the trait. The second p-value corresponds to testing whether the SNP does not have any effect. The p-value for the pleiotropy test is the maximum of these two values.

## Author(s)

Yangqing Deng and Wei Pan.

## References

- Deng, Y., Pan, W. (2017). Testing Genetic Pleiotropy with GWAS Summary Statistics for Marginal and Conditional Analyses. *Genetics*. 2017 Dec; 207(4): 1285-1299. doi: 10.1534/genetics.117.300347.
- Schaid, D. J., Tong, X., Larrabee, B., Kennedy, R. B., Poland, G. A., & Sinnwell, J.P. (2016). Statistical Methods for Testing Genetic Pleiotropy. *Genetics* 204(2): 483-497.

## Examples

```
#2 correlated SNPs, 2 traits
set.seed(13)
x1=rbinom(1000,1,0.3)
x2=c(x1[1:300],rbinom(1000-300,1,0.3))
y2=rnorm(1000)+x1
y1=rnorm(1000)+y2/2
X=cbind(x1,x2)
Y=cbind(y1,y2)
```

```

#do not condition on traits (p-values for pleiotropy tests)
#y1 ~ x1 + x2; y2 ~ x1 + x2 (H0: none or only one of x1's coefficients is nonzero)
qq=Plei(X,Y,method="lrt",ay=0,suby=c())
max(qq)
qq=Plei(X,Y,method="wald",ay=0,suby=c())
max(qq)

#condition on both traits
#y1 ~ x1 + x2 + y2; y2 ~ x1 + x2 + y1
qq=Plei(X,Y,method="wald",ay=1,suby=c())
max(qq)

#condition on trait 2
#y1 ~ x1 + x2 + y2; y2 ~ x1 + x2
qq=Plei(X,Y,method="wald",ay=1,suby=c(1))
max(qq)

```

PleiSum

*PleiSum*

## Description

Test pleiotropy with union-intersection tests and summary statistics.

## Usage

```
PleiSum(BM,SM,XX=diag(1),YY0,n,pr=c(1:(nrow(YY0),nrow(YY0)*nrow(XX)+1),method="wald",ay=1,suby=c(1))
```

## Arguments

BM	a $p \times q$ matrix containing marginal effect sizes (or Z-scores) from summary statistics. $q$ is the number of SNPs, and $p$ is the number of traits. The first SNP is to be tested, while the other $(q-1)$ SNPs are treated as covariates to adjust for.
SM	a $p \times q$ matrix containing marginal standard errors corresponding to BM from summary statistics. If Z-scores are used for BM, all entries of SM should be set to 1.
XX	a $q \times q$ estimated covariance matrix for the $q$ SNPs, the order of which should be consistent with BM.
YY0	a $p \times p$ estimated correlation matrix for the $p$ traits, the order of which should be consistent with SM.
n	a number indicating the total sample size.
pr	a vector indicating which tests should be included in the union-intersection test. It is suggested to only use the default setting, which includes 1, 2, ..., $p$ and $p \times q + 1$ . 1, 2, ..., $p$ correspond to testing whether the first SNP only influences one of the $p$ traits. $p \times q + 1$ correspond to testing whether the first SNP influences none of the $p$ traits.
method	the method to be used for each individual test. method = "lrt" uses the likelihood ratio test as described in Schaid et. al (2016); method = "wald" uses the Wald test with ordinary least square estimates and sandwich covariance matrices. It is recommended to use the Wald test when $ay = 1$ .

ay	whether some of the traits should be adjusted for. ay = 1 means to use conditional analysis adjusting for traits (and possibly SNPs if q is bigger than 1); ay = 0 means not to adjust for any traits.
suby	a vector indicating which of the traits should NOT be adjusted for. It is only effective when ay = 1. If suby is not specified, all of the traits will be adjusted for in the conditional analysis (which is not recommended). If suby = c(1,3), the first and the third traits will not be adjusted for. (In the current version, suby only works for method = "wald"; if method = "lrt", suby is always considered as c())

### Value

PleiSum returns a vector containing 2 p-values. The first p-value corresponds to testing whether the SNP does not have any effect. The second p-value corresponds to testing whether the SNP of interest only influences one of the trait. The p-value for the pleiotropy test is the maximum of these two values.

### Author(s)

Yangqing Deng and Wei Pan.

### References

- Deng, Y., Pan, W. (2017). Testing Genetic Pleiotropy with GWAS Summary Statistics for Marginal and Conditional Analyses. *Genetics*. 2017 Dec; 207(4): 1285-1299. doi: 10.1534/genetics.117.300347.
- Schaid, D. J., Tong, X., Larrabee, B., Kennedy, R. B., Poland, G. A., & Sinnwell, J.P. (2016). Statistical Methods for Testing Genetic Pleiotropy. *Genetics* 204(2): 483-497.

### Examples

```
#2 correlated SNPs, 2 traits
set.seed(13)
x1=rbinom(1000,1,0.3)
x2=c(x1[1:300],rbinom(1000-300,1,0.3))
y2=rnorm(1000)+x1
y1=rnorm(1000)+y2/2

#standardization
x1=x1-mean(x1)
x2=x2-mean(x2)
y1=y1-mean(y1)
y2=y2-mean(y2)

#summary statistics
a=summary(lm(y1~x1-1))$coefficients
b=summary(lm(y1~x2-1))$coefficients
c=summary(lm(y2~x1-1))$coefficients
d=summary(lm(y2~x2-1))$coefficients

B1=as.matrix(c(a[1],b[1]))
S1=as.matrix(c(a[2],b[2]))
B2=as.matrix(c(c[1],d[1]))
S2=as.matrix(c(c[2],d[2]))
BM=t(cbind(B1,B2))
```



```

SM=t(cbind(S1,S2))
n=1000
XX=cov(cbind(x1,x2))
YY0=cor(cbind(y1,y2))

#do not condition on traits (p-values for pleiotropy tests)
#y1 ~ x1 + x2; y2 ~ x1 + x2 (H0: none or only one of x1's coefficients is nonzero)
qq=PleiSum(BM,SM,XX,YY0,n,method="lrt",ay=0,suby=c())
max(qq)
qq=PleiSum(BM,SM,XX,YY0,n,method="wald",ay=0,suby=c())
max(qq)

#condition on both traits
#y1 ~ x1 + x2 + y2; y2 ~ x1 + x2 + y1
qq=PleiSum(BM,SM,XX,YY0,n,method="wald",ay=1,suby=c())
max(qq)

#condition on trait 2
#y1 ~ x1 + x2 + y2; y2 ~ x1 + x2
qq=PleiSum(BM,SM,XX,YY0,n,method="wald",ay=1,suby=c(1))
max(qq)

```

SMI

*SMI***Description**

Build a joint linear model for one trait vs.  $q$  SNPs using summary statistics and the MI-type approach.

**Usage**

```
SMI(BM,SM,N,Bref,mult=30)
```

**Arguments**

BM	a $q \times 1$ matrix containing marginal effects of the $q$ SNPs on the trait.
SM	a $q \times 1$ matrix containing standard errors for BM.
N	a $q \times 1$ matrix containing sample sizes for each coefficient in BM.
Bref	a $nref \times q$ matrix. Reference data with $nref$ subjects and the $q$ SNPs.
mult	the number of imputations for the MI-type approach. When $mult=0$ , do not use the MI-type approach.

**Value**

beta	coefficient estimates.
cov	the covariance matrix for coefficients.
chisq	the test statistic for the Wald test (jointly testing the $q$ SNPs).
df	the degree of freedom for the test statistic.
pvalue	the p-value.

**Author(s)**

Yangqing Deng and Wei Pan.

**References**

Deng, Y., Pan, W. (2018a). Improved Use of Small Reference Panels for Conditional and Joint Analysis with GWAS Summary Statistics. *Genetics*. June 1, 2018 vol. 209 no. 2 401-408; <https://doi.org/10.1534/genetics.118.300813>.

**Examples**

```
#2 SNPs, 1 trait
set.seed(190)
x1=rbinom(1000,1,0.3)
x2=c(x1[1:400],rbinom(1000-400,1,0.3))
y1=rnorm(1000)
Bref0=cbind(x1,x2)

#standardization
x1=x1-mean(x1)
x2=x2-mean(x2)
y1=y1-mean(y1)

#summary statistics
a=summary(lm(y1~x1-1))$coefficients
b=summary(lm(y1~x2-1))$coefficients

BM=as.matrix(c(a[1],b[1]))
SM=as.matrix(c(a[2],b[2]))
N=matrix(1000,nrow=2,ncol=1)

#reference data
x1b=rbinom(500,1,0.3)
x2b=c(x1b[1:220],rbinom(500-220,1,0.3))
Bref=cbind(x1b,x2b)

#models using reference data
SMI(BM,SM,N,Bref,mult=1)
SMI(BM,SM,N,Bref,mult=30)

#model using original data
SMI(BM,SM,N,Bref0,mult=1)
```

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