

Package ‘r2redux’

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Title R2 statistic

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Description R2 statistic for significance test.

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olkin_beta1_2	<i>olkin_beta1_2 function</i>
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Description

This function derives Information matrix for β_1^2 and β_2^2 where β_1 and β_2 are regression coefficients from a multiple regression model, i.e. $y = x_1 \cdot \beta_1 + x_2 \cdot \beta_2 + e$, where y , x_1 and x_2 are column-standardised (see Olkin and Finn 1995).

Usage

```
olkin_beta1_2(omat, nv)
```

Arguments

omat	3 by 3 matrix having the correlation coefficients between y, x1 and x2, i.e. $\text{omat}=\text{cor}(\text{dat})$ where dat is N by 3 matrix having variables in the order of cbind(y,x1,x2)
nv	sample size

References

Olkin, I. and J.D. Finn, Correlations redux. Psychological Bulletin, 1995. 118(1): p. 155.

Examples

```
olkin_beta1_2 (omat,nv) (see example file)
dat=read.table("test_ukbb_thresholds_scaled") (see example files)
omat=cor(dat)
omat
1.000 0.195 0.197 0.197 0.197 0.194 0.191 0.183 0.168 0.130 0.103
0.195 1.000 0.998 0.995 0.992 0.984 0.964 0.936 0.860 0.757 0.669
0.197 0.998 1.000 0.998 0.996 0.988 0.969 0.942 0.867 0.764 0.674
0.197 0.995 0.998 1.000 0.998 0.991 0.972 0.946 0.872 0.769 0.679
0.197 0.992 0.996 0.998 1.000 0.995 0.978 0.953 0.879 0.777 0.687
0.194 0.984 0.988 0.991 0.995 1.000 0.988 0.966 0.897 0.796 0.706
0.191 0.964 0.969 0.972 0.978 0.988 1.000 0.987 0.928 0.831 0.742
0.183 0.936 0.942 0.946 0.953 0.966 0.987 1.000 0.959 0.869 0.780
0.168 0.860 0.867 0.872 0.879 0.897 0.928 0.959 1.000 0.942 0.859
0.130 0.757 0.764 0.769 0.777 0.796 0.831 0.869 0.942 1.000 0.951
0.103 0.669 0.674 0.679 0.687 0.706 0.742 0.780 0.859 0.951 1.000
```

olkin_beta_inf	<i>olkin_beta_inf function</i>
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Description

This function derives Information matrix for beta1 and beta2 where beta1 and 2 are regression coefficients from a multiple regression model, i.e. $y = x1 \cdot \text{beta1} + x2 \cdot \text{beta2} + e$, where y, x1 and x2 are column-standardised (see Olkin and Finn 1995).

Usage

```
olkin_beta_inf(omat, nv)
```

Arguments

omat	3 by 3 matrix having the correlation coefficients between y, x1 and x2, i.e. $\text{omat}=\text{cor}(\text{dat})$ where dat is N by 3 matrix having variables in the order of cbind(y,x1,x2)
nv	sample size

References

Olkin, I. and J.D. Finn, Correlations redux. Psychological Bulletin, 1995. 118(1): p. 155.

Examples

```

olkin_beta_inf (omat,nv) (see example file)
dat=read.table("test_ukbb_thresholds_scaled") (see example files)
omat=cor(dat)
omat
1.000 0.195 0.197 0.197 0.197 0.194 0.191 0.183 0.168 0.130 0.103
0.195 1.000 0.998 0.995 0.992 0.984 0.964 0.936 0.860 0.757 0.669
0.197 0.998 1.000 0.998 0.996 0.988 0.969 0.942 0.867 0.764 0.674
0.197 0.995 0.998 1.000 0.998 0.991 0.972 0.946 0.872 0.769 0.679
0.197 0.992 0.996 0.998 1.000 0.995 0.978 0.953 0.879 0.777 0.687
0.194 0.984 0.988 0.991 0.995 1.000 0.988 0.966 0.897 0.796 0.706
0.191 0.964 0.969 0.972 0.978 0.988 1.000 0.987 0.928 0.831 0.742
0.183 0.936 0.942 0.946 0.953 0.966 0.987 1.000 0.959 0.869 0.780
0.168 0.860 0.867 0.872 0.879 0.897 0.928 0.959 1.000 0.942 0.859
0.130 0.757 0.764 0.769 0.777 0.796 0.831 0.869 0.942 1.000 0.951
0.103 0.669 0.674 0.679 0.687 0.706 0.742 0.780 0.859 0.951 1.000

```

r2_diff

r2_diff function

Description

This function estimates $\text{var}(R2(y \sim x[,v1]) - R2(y \sim x[,v2]))$ where $R2$ is the R squared value of the model, y is N by 1 matrix having the dependent variable, and x is N by M matrix having M explanatory variables. $v1$ or $v2$ indicates the i th column in the x matrix ($v1$ or $v2$ can be multiple values between $1 - M$, see Arguments below)

Usage

```
r2_diff(dat, v1, v2, nv)
```

Arguments

dat	N by $(M+1)$ matrix having variables in the order of <code>cbind(y,x)</code>
v1	This can be set as <code>v1=c(1)</code> or <code>v1=c(1,2)</code>
v2	This can be set as <code>v2=c(2)</code> , <code>v2=c(3)</code> , <code>v2=c(1,3)</code> or <code>v2=c(3,4)</code>
nv	sample size

Examples

To get the test statistics for the difference between $R2(y \sim x[,v1])$ and $R2(y \sim x[,v2])$. (here we define `R2_1=R2(lm(y~x[,v1]))` and `R2_2=R2(lm(y~x[,v2]))`)

```

dat=read.table("test_ukbb_thresholds_scaled") (see example files)
nv=length(dat$V1)
v1=c(1)
v2=c(2)
output=r2_diff(dat,v1,v2,nv)
output

r2redux output

```

```
output$var1 (variance of R2_1)
0.0001437583
```

```
output$var2 (variance of R2_2)
0.0001452828
```

```
output$var_diff (variance of difference between R2_1 and R2_2)
5.678517e-07
```

```
output$r2_based_p (p-value for significant difference between R2_1 and R2_2)
0.5514562
```

```
output$mean_diff (differences between R2_1 and R2_2)
-0.0004488044
```

```
output$upper_diff (upper limit of 95% CI for the difference)
0.001028172
```

```
output$lower_diff (lower limit of 95% CI for the difference)
-0.001925781
```

To get the test statistics for the difference between $R2(y \sim x[,v1])$ and $R2(y \sim x[,v2])$.
(here $R2_1 = R2(lm(y \sim x[,v1] + x[,v2]))$ and $R2_2 = R2(lm(y \sim x[,v1]))$)

```
dat=read.table("test_ukbb_thresholds_scaled") (see example files)
nv=length(dat$V1)
v1=c(1,2)
v2=c(1)
output=r2_diff(dat,v1,v2,nv)
output
```

r2redux output

```
output$var1 (variance of R2_1)
0.0001475195
```

```
output$var2 (variance of R2_2)
0.0001437583
```

```
output$var_diff (variance of difference between R2_1 and R2_2)
2.321425e-06
```

```
output$r2_based_p (p-value for significant difference between R2_1 and R2_2)
0.4369177
```

```
output$mean_diff (differences between R2_1 and R2_2)
0.0006042383
```

```
output$upper_diff (upper limit of 95% CI for the difference)
0.004887989
```

```
output$lower_diff (lower limit of 95% CI for the difference)
-0.0005574975
```

r2_enrich

*r2_enrich***Description**

This function estimates $\text{var}(t1/(t1+t2))$ where $t1 = R2(y \sim x[,v1] + x[,v2]) - R2(y \sim x[,v1])$ and $t2 = R2(y \sim x[,v1] + x[,v2]) - R2(y \sim x[,v2])$ where $R2$ is the R squared value of the model, y is N by 1 matrix having the dependent variable, and x is N by M matrix having M explanatory variables. $v1$ or $v2$ indicates the i th column in the x matrix ($v1$ or $v2$ should be a single interger between $1 - M$, see Arguments below)

Usage

```
r2_enrich(dat, v1, v2, nv, expl)
```

Arguments

dat	N by $(M+1)$ matrix having variables in the order of <code>cbind(y,x)</code>
v1	and v2 can be set as $v1=1$ and $v2=2$, $v1=2$ and $v2=1$, $v1=3$ and $v2=2$, or any combination as long as the value is between $1 - M$
nv	sample size
expl	The expectation of the ratio (e.g. ratio of # SNPs in genomic partitioning)

Examples

```
To get test statistics for the ratio, i.e. t1/(t1+t2).
t1 = R2(y~x[,v1]+x[,v2]) - R2(y~x[,v1]) and
t2 = R2(y~x[,v1]+x[,v2]) - R2(y~x[,v2])
(here we define R2_1=R2(y~x[,v1]), R2_2=R2(y~x[,v2]), R2_12=R2(y~x[,v1]+x[,v2]))

dat=read.table("test_ukbb_enrichment_choles") (see example file)
nv=length(dat$V1)
v1=c(1)
v2=c(2)
expected_ratio=0.04
output=r2_enrich(dat,v1,v2,nv,expected_ratio)
output

r2redux output

output$var1 (variance of R2_1)
8.758455e-05

output$var2 (variance of R2_2)
7.36385e-05

output$var12 (variance of R2_12)
0.000102236

output$var_diff1_2 (var of difference of R2(y~x[,v1]) - R2(y~x[,v2]))
6.074567e-05

output$var_diff12_1 (var of difference of R2(y~x[,v1]+x[,v2]) - R2(y~x[,v1]))
```

```

1.184853e-05

output$var_diff12_2 (var of difference of R2(y~x[,v1]+x[,v2]) - R2(y~x[,v2]))
2.650564e-05

output$mean_diff12_1 (difference of R2(y~x[,v1]+x[,v2]) - R2(y~x[,v1]))
0.003048595

output$mean_diff12_2 (difference of R2(y~x[,v1]+x[,v2]) - R2(y~x[,v2]))
0.006845484

output$ratio (ratio of t1/(t1+t2))
0.6918768

output$ratio_var (variance of ratio, (t1/(t1+t2))
0.1324076

output$enrich_p (p-value for testing the ratio significantly different
from the expectation (expl))
0.07321821

output$upper_ratio (upper limit of 95% CI for the ratio)
1.405079

output$lower_ratio (lower limit of 95% CI for the ratio)
-0.02132515

```

r2_enrich_beta	<i>r2_enrich_beta</i>
----------------	-----------------------

Description

This function estimates $\text{var}(t1/(t1+t2))$ where $t1 = \text{beta1}^2$ and $t2 = \text{beta2}^2$, and beta1 and 2 are regression coefficients from a multiple regression model, i.e. $y = x1 \cdot \text{beta1} + x2 \cdot \text{beta2} + e$, where y , $x1$ and $x2$ are column-standardised (see Olkin and Finn 1995). y is N by 1 matrix having the dependent variable, and $x1$ is N by 1 matrix having the i th explanatory variables. $x2$ is N by 1 matrix having the j th explanatory variables. $v1$ and $v2$ indicates the i th and j th column in the data ($v1$ or $v2$ should be a single interger between $1 - M$, see Arguments below)

Usage

```
r2_enrich_beta(dat, v1, v2, nv, expl)
```

Arguments

dat	N by $(M+1)$ matrix having variables in the order of <code>cbind(y,x)</code>
v1	and $v2$ can be set as $v1=1$ and $v2=2$, $v1=2$ and $v2=1$, $v1=3$ and $v2=2$, or any combination as long as the value is between $1 - M$
nv	sample size
expl	The expectation of the ratio (e.g. ratio of # SNPs in genomic partitioning)

References

Olkin, I. and J.D. Finn, Correlations redux. Psychological Bulletin, 1995. 118(1): p. 155.

Examples

To get the test statistic for the ratio which is significantly different from the expectation.
 $\text{var}(t1/(t1+t2))$, where $t1 = \text{beta1}^2$ and $t2 = \text{beta2}^2$.
 beta1 and beta2 are regression coefficients from a multiple regression model,
 i.e. $y = x1 \cdot \text{beta1} + x2 \cdot \text{beta2} + e$, where y , $x1$ and $x2$ are column-standardised

```
dat=read.table("test_ukbb_enrichment_choles") (see example file)
nv=length(dat$V1)
v1=c(1)
v2=c(2)
expected_ratio=0.04
output=r2_enrich_beta(dat,v1,v2,nv,expected_ratio)
output
```

```
r2redux output
```

```
output$var1 (variance of t1)
7.072931e-05
```

```
output$var2 (variance of t2)
3.161929e-05
```

```
output$var1_2 (difference between t1 and t2)
0.000162113
```

```
output$beta1_sq (1st squared regression coefficient)
0.01118301
```

```
output$beta2_sq (2nd squared regression coefficient)
0.004980285
```

```
output$cov (covariance between t1 and t2)
-2.988221e-05
```

```
output$ratio (ratio of t1/(t1+t2_2))
0.6918768
```

```
output$ratio_var (variance of ratio)
0.1324076
```

```
output$enrich_p (p-value for testing the ratio significantly different from
the expectation (exp1))
0.07321821
```

```
output$upper_ratio (upper limit of 95% CI for the ratio)
1.405079
```

```
output$lower_ratio (lower limit of 95% CI for the ratio)
-0.02132515
```

r2_var

r2_var function

Description

This function estimates $\text{var}(R^2(y \sim x[,v1]))$ where R^2 is the R squared value of the model, where R^2 is the R squared value of the model, y is N by 1 matrix having the dependent variable, and x is N by M matrix having M explanatory variables. $v1$ indicates the i th column in the x matrix ($v1$ can be multiple values between $1 - M$, see Arguments below)

Usage

```
r2_var(dat, v1, nv)
```

Arguments

dat	N by $(M+1)$ matrix having variables in the order of <code>cbind(y,x)</code>
v1	This can be set as <code>v1=c(1)</code> , <code>v1=c(1,2)</code> or possibly with more values
nv	sample size

Examples

```
To get the test statistics for  $R^2(y \sim x[,v1])$ 

dat=read.table("test_ukbb_thresholds_scaled") (see example file)
nv=length(dat$V1)
v1=c(1) ##1= $R^2(lm(y \sim x[,v1]))$ 
output=r2_var(dat,v1,nv)

r2redux output

output$var (variance of  $R^2$ )
0.0001437583

output$r2_based_p (P-value under null hypothesis,  $R^2=0$ )
1.213645e-10

output$mean_r2 ( $R^2$ )
0.03836254

output$upper_r2 (upper limit of 95% CI for  $R^2$ )
0.06435214

output$lower_r2 (lower limit of 95% CI for  $R^2$ )
0.01763347

To get the test for  $R^2(y \sim x[,v1])$ 

dat=read.table("test_ukbb_thresholds_scaled") (see example file)
nv=length(dat$V1)
v1=c(1,2,3) ##1= $R^2(lm(y \sim x[,v1]+x[,v2]+x[,v3]))$ 
```



```
outout=r2_var(dat,v1,nv)
output

r2redux output

output$var (variance of R2)
0.0001499374

output$r2_based_p (R2 based P-value)
7.461267e-11

output$mean_r2 (R2)
0.03917668

output$upper_r2 (upper limit of 95% CI for R2)
0.06538839

output$lower_r2 (lower limit of 95% CI for R2)
0.01821657
```

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