# Package 'r2redux'

May 27, 2022

Version 1.0.4	
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<b>Description</b> R2	statistic for significance test.
License GPL (>	=3)
Encoding UTF-	3
Roxygen list(ma	rkdown = TRUE)
RoxygenNote 7.	1.2
NeedsCompilati	on no
R topics do	cumented:
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cc_trf	cc_trf function

# Description

Title R2 Statistic

This function transforms the predictive ability (R2) and its standard error (se) between the observed scale and liability scale

```
cc_trf(R2, se, K, P)
```

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#### **Arguments**

R2	R2 or Coefficient of determination on the observed or liability scale
se	Standard error of R2
K	Population prevalence
P	The ratio of cases in the study samples

#### References

Lee, S. H., Goddard, M. E., Wray, N. R., and Visscher, P. M. A better coefficient of determination for genetic profile analysis. Genetic epidemiology, (2012). 36(3): p. 214-224.

#### **Examples**

```
To get the transformed R2
output=cc_trf(0.06, 0.002, 0.05, 0.05)
output$R21 (transformed R2 to the liability scale)
 0.2679337
 output$sel (transformed se to the liability scale)
 0.008931123
 output$R2O (transformed R2 to the observed scale)
 0.01343616
 output$seO (transformed se to the observed scale)
 0.000447872
```

```
olkin_beta1_2
                       olkin_beta1_2 function
```

## Description

This function derives Information matrix for beta1<sup>2</sup> and beta2<sup>2</sup> where beta1 and 2 are regression coefficients from a multiple regression model, i.e.  $y = x1 \cdot beta1 + x2 \cdot beta2 + e$ , where y, x1 and x2 are column-standardised, (i.e. in the context of correlation coefficients, 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. omat=cor(dat) where dat is N by 3 matrix having variables in the order of cbind $(y,x1,x2)$
nv	sample size

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#### References

Olkin, I. and J.D. Finn, Correlations redux. Psychological Bulletin, 1995. 118(1): p. 155. Momin, M.M., Lee, S. Wray, N. and S. Lee, S.H. The variance and covariance of the coefficients of determination for genetic profile analysis (will be subbitted soon)

#### **Examples**

```
To get information (variance-covariance) matrix of beta1^2 and beta2^2 where beta1 and 2 are regression coefficients from a multiple regression model.
```

```
dat=read.table("test_ukbb_thresholds_scaled") (see example files)
omat=cor(dat)[1:3,1:3]
omat
1.0000000 0.1958636 0.1970060
0.1958636 1.0000000 0.9981003
0.1970060 0.9981003 1.0000000
nv=length(dat$V1)
output=olkin_beta1_2(omat,nv)
output
output$info (2x2 information (variance-covariance) matrix)
0.04146276 0.08158261
0.08158261 0.16111124
output$var1 (variance of beta1^2)
0.04146276
output$var2 (variance of beta2^2)
0.1611112
output$var1_2 (variance of difference between beta1^2 and beta2^2)
0.03940878
```

```
{\tt olkin\_beta\_inf} \qquad \qquad \textit{olkin\_beta\_inf function}
```

## **Description**

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

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

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#### **Arguments**

omat 3 by 3 matrix having the correlation coefficients between y, x1 and x2, i.e.

omat=cor(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**

```
To get information (variance-covariance) matrix of betal and beta2 where
beta1 and 2 are regression coefficients from a multiple regression model.
dat=read.table("test_ukbb_thresholds_scaled") (see example files)
omat=cor(dat)[1:3,1:3]
omat
1.0000000 0.1958636 0.1970060
0.1958636 1.0000000 0.9981003
0.1970060 0.9981003 1.0000000
nv=length(dat$V1)
output=olkin_beta_inf(omat,nv)
output
output$info (2x2 information (variance-covariance) matrix)
0.2531406 -0.2526212
-0.2526212 0.2530269
output$var1 (variance of beta1)
0.2531406
output$var2 (variance of beta2)
0.2530269
output$var1_2 (variance of difference between beta1 and beta2)
1.01141
```

r2 diff

r2\_diff function

#### **Description**

This function estimates  $var(R2(y\sim x[,v1]) - R2(y\sim x[,v2]))$  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 ith column in the x matrix (v1 or v2 can be multiple values between 1 - M, see Arguments below)

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

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#### **Arguments**

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

#### **Examples**

```
To get the test statistics for the difference between R2(y\sim x[,v1]) and
 \mbox{R2} \ (\mbox{$y$}\mbox{$\times$}\mbox{$x[,v2]$}) \mbox{. (here we define } \mbox{$R2$}\mbox{$1$=$R2} \ (\mbox{$y$}\mbox{$\times$}\mbox{$x[,v1]$})) \mbox{ and } \mbox{$R2$}\mbox{$2$=$R2} \ (\mbox{$y$}\mbox{$\times$}\mbox{$x[,v2]$}))) \mbox{}
dat=read.table("test_ukbb_thresholds_scaled") (see example files)
nv=length(dat$V1)
v1=c(1)
v2=c(2)
output=r2_diff(dat,v1,v2,nv)
output
r2redux output
output$rsq1 (R2_1)
0.03836254
output$rsq2 (R2_2)
0.03881135
output$var1 (variance of R2_1)
0.0001437583
output$var2 (variance of R2_2)
0.0001452828
output$var_diff (variance of difference between R2_1 and R2_2)
5.678517e-07
outputr2_based_p (p-value for significant difference between R2_1 and R2_2)
0.5514562
output$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]+x[,v2]) and
R2(y\sim x[,v2]). (here R2_1=R2(y\sim x[,v1]+x[,v2]) and R2_2=R2(y\sim x[,v1]))
dat=read.table("test_ukbb_thresholds_scaled") (see example files)
nv=length(dat$V1)
```

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```
v1=c(1,2)
v2=c(1)
output=r2_diff(dat,v1,v2,nv)
output
r2redux output
output$rsq1 (R2_1)
0.03896678
output$rsq2 (R2_2)
0.03836254
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
outputmean\_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\_beta

r2\_enrich\_beta

## **Description**

This function estimates var(t1/exp) - (t2/(1-exp)), where  $t1 = beta1^2$  and  $t2 = beta2^2$ , and beta1 and 2 are regression coefficients from a multiple regression model, i.e.  $y = x1 \cdot beta1 + x2 \cdot beta2 + e$ , where y, x1 and x2 are column-standardised (see Olkin and Finn 1995). y is y is y 1 matrix having the dependent variable, and y 1 is y 1 matrix having the ith explanatory variables. y 2 indicates the ith and jth column in the data y 1 or y 2 should be a single interger between 1 - y 3. We have y 3. Note that y 2 enrich (above) and y 2 enrich beta is equivalent (identical p-value derived).

```
r2_enrich_beta(dat, v1, v2, nv, exp1)
```

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#### **Arguments**

dat	N by (M+1) matrix having variables in the order of cbind(y,x)
nv	sample size
exp1	The expectation of the ratio (e.g. ratio of # SNPs in genomic partitioning)
v1/v2	These 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

#### 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.
var[(t1/exp) - (t2/(1-exp))], where t1 = beta1^2 and t2 = beta2^2.
beta1 and beta2 are regression coefficients from a multiple regression model,
i.e. y = x1 \cdot beta1 + x2 \cdot beta2 + e, where y, x1 and x2 are column-standardised
dat=read.table("test_ukbb_enrichment_choles") (see example file)
nv=length(dat$V1)
v1=c(1)
v2=c(2)
expected_ratio=0.04
output=r2_enrich_beta(dat,v1,v2,nv,expected_ratio)
output
r2redux output
output$beta1_sq (t1)
0.01118301
output$beta2_sq (t2)
0.004980285
output$var1 (variance of t1)
7.072931e-05
output$var2 (variance of t2)
3.161929e-05
output$var1_2 (variance of difference between t1 and t2)
0.000162113
output$cov (covariance between t1 and t2)
-2.988221e-05
output\$enrich_p2 (p-value for testing the difference between t1/exp and t2/(1-exp))
0.1997805
output$mean_diff (difference between t1/exp and t2/(1-exp))
0.2743874
output$var_diff (variance of difference, t1/exp - t2/(1-exp))
```

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```
0.04579649

output$upper_diff (upper limit of 95% CI for the mean difference)
0.6938296

output$lower_diff (lower limit of 95% CI for the mean difference)
-0.1450549
```

r2\_var

r2\_var function

## **Description**

This function estimates  $var(R2(y\sim x[,v1]))$  where R2 is the R squared value of the model, 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 indicates the ith 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 cbind(y,x) v1 This can be set as v1=c(1), v1=c(1,2) or possibly with more values nv sample size

## **Examples**

```
To get the test statistics for R2(y~x[,v1])

dat=read.table("test_ukbb_thresholds_scaled") (see example file)
nv=length(dat$V1)
v1=c(1)
output=r2_var(dat,v1,nv)

r2redux output

output$rsq (R2)
0.03836254

output$var (variance of R2)
0.0001437583

output$r2_based_p (P-value under the null hypothesis, i.e. R2=0)
1.213645e-10

output$upper_r2 (upper limit of 95% CI for R2)
0.06435214
```

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```
output$lower_r2 (lower limit of 95% CI for R2)
0.01763347
To get the test statistic for R2(y \sim x[,v1]+x[,v2]+x[,v3])
dat=read.table("test_ukbb_thresholds_scaled") (see example file)
nv=length(dat$V1)
v1=c(1,2,3)
outout=r2_var(dat,v1,nv)
output
r2redux output
output$rsq (R2)
0.03917668
output$var (variance of R2)
0.0001499374
output$r2_based_p (R2 based P-value)
7.461267e-11
output$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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