

Package ‘powerGWASinteraction’

October 14, 2022

Version 1.1.3

Date 2015-07-02

Title Power Calculations for GxE and GxG Interactions for GWAS

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Depends R (>= 2.10), mvtnorm, pwr

Imports stats

Description Analytical power calculations for GxE and GxG interactions for case-control studies of candidate genes and genome-wide association studies (GWAS). This includes power calculation for four two-step screening and testing procedures. It can also calculate power for GxE and GxG without any screening.

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NeedsCompilation no

Repository CRAN

Date/Publication 2015-07-03 01:34:42

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powerGE	<i>Power for GxE interactions in genetic association studies</i>
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Description

This routine carries out (analytical, approximate) power calculations for identifying Gene-Environment interactions in Genome Wide Association Studies

Usage

```
powerGE(n, power, model, caco, alpha, alpha1, maintain.alpha)
```

Arguments

n	Sample size: combined number of cases and controls. Note: exactly one of n and power should be specified.
power	Power: targeted power. Note: exactly one of n and power should be specified.
model	List specifying the genetic model. This list contains the following objects: <ul style="list-style-type: none"> • prev Prevalence of the outcome in the <i>population</i>. Note that for case-only and empirical Bayes estimators to be valid, the prevalence needs to be low. • pGene Probability that a <i>binary</i> SNP is 1 (i.e. not the minor allele frequency for a three level SNP). • pEnv Frequency of the binary environmental variable. • orGE Odds ratio between the binary SNP and binary environmental variable. • beta.LOR Vector of length three with the odds ratios of the genetic, environmental, and GxE interaction effect, respectively. • nSNP Number of SNPs (genes) being tested.
caco	Fraction of the sample that are cases (default = 0.5).
alpha	Overall (family-wise) Type 1 error (default = 0.05).
alpha1	Significance level at which testing during the first stage (screening) takes place. If alpha1 = 1, there is no screening.
maintain.alpha	Some combinations of screening and GxE testing methods do not maintain the proper Type 1 error. Default is True: combinations that do not maintain the Type 1 error are not computed. If maintain.alpha is False all combinations are computed.

Details

The routine computes power for a variety of two-stage procedures. Five different screening procedures are used:

- **No screening** All SNPs are tested for interaction
- **Marginal screening** Only SNPs that are marginally significant at level alpha1 are screened for interaction. See Kooperberg and LeBlanc (2010).
- **Correlation screening** Only SNPs that are, combined over all cases and controls, associated with the environmental variable at level alpha 1 are screened for interaction. See Murcray et al. (2012).
- **Cocktail screening** SNPs are screened on the most significant of marginal and correlation screening. See Hsu et al. (2012).
- **Chi-square screening** SNPs are screened using a chi-square combination of correlation and marginal screening. See Gauderman et al. (2013).

After screening, the SNPs that pass the screen can be tested using

- **Case-control** The standard case-control estimator.
- **Case-only** The case-only estimator.
- **Empirical Bayes** The empirical Bayes estimator of Mukherjee and Chatterjee (2010).

If screening took place using the correlation or chi-square screening, the Type 1 error won't be maintained if the final GxE testing is carried out using either the case-only or empirical Bayes estimator. See Dai et al. (2012). The cocktail screening maintains the Type 1 family wise error rate, since only those SNPs that pass on to the second stage using marginal screening will use the case-only or empirical Bayes estimator, the SNPs that pass on to the second stage using correlation screening will always use the case-control estimator.

When SNP and environment are correlated in the population (i.e. ρ_{GE} does not equal 1) the case-only estimator does not maintain the Type 1 error. The empirical Bayes estimator may also have a moderately inflated Type 1 error. When the disease is common either the case-only estimator or the empirical Bayes estimator also may not estimate the GxE interaction.

Power calculations are described in Kooperberg, Dai, and Hsu (2014). Briefly, for a given genetic model we compute the expected p-values for all screening statistics. We then use a normal approximation to compute the probability that this SNP passes the screening (e.g., if α_1 equaled this expected p-value this probability would be exactly 0.5), and combine this with power calculations for the second stage of GxE testing.

Value

A list with three components.

power	A 5x3 matrix with estimated power for all testing approaches, only if n was specified.
samplesize	A 5x3 matrix with required sample sizes for all testing approaches, only if power was specified.
expected.p	A 5x3 matrix with the expected p value for the SNP to pass screening. This p-value depends on the sample size, but not on the second stage testing.
prob.select	A 5x3 matrix with the probability that the interacting SNP would pass the screening stage. This probability depends on the sample size, but not on the second stage testing.

Author(s)

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References

- Dai JY, Kooperberg C, LeBlanc M, Prentice RL (2012). Two-stage testing procedures with independent filtering for genome-wide gene-environment interaction. *Biometrika*, **99**, 929-944.
- Gauderman WJ, Zhang P, Morrison JL, Lewinger JP (2013). Finding novel genes by testing GxE interactions in a genome-wide association study. *Genetic Epidemiology*, **37**, 603-613.
- Hsu L, Jiao S, Dai JY, Hutter C, Peters U, Kooperberg C (2012). Powerful cocktail methods for detecting genome-wide gene-environment interaction. *Genetic Epidemiology*, **36**, 183-194.

Kooperberg C, Dai, JY, Hsu L (2014). Two-stage procedures for the identification of gene x environment and gene x gene interactions in genome-wide association studies. *To appear*.

Kooperberg C, LeBlanc ML (2008). Increasing the power of identifying gene x gene interactions in genome-wide association studies. *Genetic Epidemiology*, **32**, 255-263.

Mukherjee B, Chatterjee N (2008). Exploiting gene-environment independence for analysis of case-control studies: an empirical Bayes-type shrinkage estimator to trade-off between bias and efficiency *Biometrics*, **64**, 685-694.

Murcray CE, Lewinger JP, Gauderman WJ (2009). Gene-environment interaction in genome-wide association studies. *American Journal of Epidemiology*, **169**, 219-226.

See Also

powerGG

Examples

```
mod1 <- list(prev=0.01,pGene=0.2,pEnv=0.2,beta.LOR=log(c(1.0,1.2,1.4)),orGE=1.2,nSNP=10^6)
results <- powerGE(n=20000, model=mod1,alpha1=.01)
print(results)
```

```
mod2 <- list(prev=0.01,pGene=0.2,pEnv=0.2,beta.LOR=log(c(1.0,1.0,1.4)),orGE=1,nSNP=10^6)
results <- powerGE(power=0.8, model=mod2,alpha1=.01)
print(results)
```

powerGG

Power for GxG interactions in genetic association studies

Description

This routine carries out (analytical, approximate) power calculations for identifying Gene-Gene interactions in Genome Wide Association Studies

Usage

```
powerGG(n, power, model, caco, alpha, alpha1)
```

Arguments

- | | |
|-------|--|
| n | Sample size: combined number of cases and controls. Note: exactly one of n and power should be specified. |
| power | Power: targeted power. Note: exactly one of n and power should be specified. |
| model | List specifying the genetic model. This list contains the following objects: <ul style="list-style-type: none"> • prev Prevalence of the outcome in the <i>population</i>. Note that for case-only and empirical Bayes estimators to be valid, the prevalence needs to be low. • pGene1 Probability that the first <i>binary</i> SNP is 1 (i.e. not the minor allele frequency for a three level SNP). |

- pGene2 Probability that the first *binary* SNP is 1 (i.e. not the minor allele frequency for a three level SNP).
- beta.LOR Vector of length three with the odds ratios of the first genetic, second genetic, and GxG interaction effect, respectively.
- nSNP Number of SNPs (genes) being tested.

caco	Fraction of the sample that are cases (default = 0.5).
alpha	Overall (family-wise) Type 1 error (default = 0.05).
alpha1	Significance level at which testing during the first stage (screening) takes place. If alpha1 = 1, there is no screening.

Details

The routine computes power calculations for a two-stage procedure with marginal screening followed by either case-control or case-only testing.

Value

A data frame consisting of two numbers: the power for the case-control and case-only approaches if *n* is specified or the required combined sample size for the case-control and case-only approaches if power is specified.

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References

Kooperberg C, LeBlanc M (2008). Increasing the power of identifying gene x gene interactions in genome-wide association studies. *Genetic Epidemiology*, **32**, 255-263.

See Also

powerGG

Examples

```
mod1 <- list(prev=0.05, pGene1=0.3, pGene2=0.3, beta.LOR=c(0,0,.6),nSNP=500000)
powerGG(n=10000,mod=mod1,caco=0.5,alpha=.05,alpha1=.001)
powerGG(power=0.8,mod=mod1,caco=0.5,alpha=.05,alpha1=.001)
```

powerGWASinteraction *Deprecated*

Description

This function is deprecated and has been replaced by powerGG and powerGE

Usage

```
powerGWASinteraction()
```

Value

An error message is printed

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See Also

powerGG, powerGE

Examples

```
powerGWASinteraction()
```

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