Package 'GMMAT'

July 21, 2025

Version 1.4.2

Date 2023-11-17

Title Generalized Linear Mixed Model Association Tests

Description Perform association tests using generalized linear mixed models (GLMMs) in genome-wide association studies (GWAS) and sequencing association studies. First, GM-MAT fits a GLMM with covariate adjustment and random effects to account for population structure and familial or cryptic relatedness. For GWAS, GMMAT performs score tests for each genetic variant as proposed in Chen et al. (2016) <DOI:10.1016/j.ajhg.2016.02.012>. For candidate gene studies, GMMAT can also perform Wald tests to get the effect size estimate for each genetic variant. For rare variant analysis from sequencing association studies, GMMAT performs the variant Set Mixed Model Association Tests (SMMAT) as proposed in Chen et al. (2019) <DOI:10.1016/j.ajhg.2018.12.012>, including the burden test, the sequence kernel association test (SKAT), SKAT-O and an efficient hybrid test of the burden test and SKAT, based on user-defined variant sets.

License GPL (>= 3)

Copyright See COPYRIGHTS for details.

Imports Rcpp, CompQuadForm, foreach, parallel, Matrix, methods, data.table

Suggests doMC, SeqArray, SeqVarTools, testthat

LinkingTo Rcpp, RcppArmadillo

Encoding UTF-8

NeedsCompilation yes

Depends R (>= 3.2.0)

Author Han Chen [aut, cre],

Matthew Conomos [aut],

Duy Pham [aut],

Arthor Gilly [ctb],

Robert Gentleman [ctb, cph] (Author and copyright holder of the C function Brent fmin),

Ross Ihaka [ctb, cph] (Author and copyright holder of the C function Brent fmin),

The R Core Team [ctb, cph] (Author and copyright holder of the C

2 GMMAT-package

function Brent_fmin),

The R Foundation [cph] (Copyright holder of the C function Brent_fmin),

Eric Biggers [ctb, cph] (Author and copyright holder of included libdeflate library),

Tino Reichardt [ctb, cph] (Author and copyright holder of threading code used in the included Zstandard (zstd) library),

Meta Platforms, Inc. and affiliates [cph] (Copyright holder of included Zstandard (zstd) library)

Maintainer Han Chen < han.chen.2@uth.tmc.edu>

Repository CRAN

Date/Publication 2023-11-17 10:20:33 UTC

Contents

GMMA ⁻	T-package	(Gei	ner	ral	ize	ed	Li	ne	ear	· N	1ix	кe	d I	Иd	od	el	As	SSC	oci	aı	io	n	Те	st.	S							
Index																																	29
	SMMAT.meta		•			•	•	•	•			•	•			•		•	•	•	•	•	•	•	•			•	•	•	•	•	25
	SMMAT																																
	glmmkin																																
	glmm.wald																																
	glmm.score.meta																																
	glmm.score																																
	example																																
	GMMAT-package																										 						2

Description

An R package for performing association tests using generalized linear mixed models (GLMMs) in genome-wide association studies (GWAS) and sequencing association studies. First, GMMAT fits a GLMM with covariate adjustment and random effects to account for population structure and familial or cryptic relatedness. For GWAS, GMMAT performs score tests for each genetic variant. For candidate gene studies, GMMAT can also perform Wald tests to get the effect size estimate for each genetic variant. For rare variant analysis from sequencing association studies, GMMAT performs the variant Set Mixed Model Association Tests (SMMAT), including the burden test, the sequence kernel association test (SKAT), SKAT-O and an efficient hybrid test of the burden test and SKAT, based on user-defined variant sets.

Details

Package: GMMAT Type: Package Version: 1.4.2 Date: 2023-11-17 License: GPL (>= 3)

Author(s)

Han Chen, Matthew P. Conomos, Duy T. Pham

Maintainer: Han Chen < Han. Chen. 2@uth.tmc.edu>

References

Brent, R.P. (1973) "Chapter 4: An Algorithm with Guaranteed Convergence for Finding a Zero of a Function", Algorithms for Minimization without Derivatives, Englewood Cliffs, NJ: Prentice-Hall, ISBN 0-13-022335-2.

Breslow, N.E. and Clayton, D.G. (1993) Approximate Inference in Generalized Linear Mixed Models. Journal of the American Statistical Association 88, 9-25.

Chen, H., Huffman, J.E., Brody, J.A., Wang, C., Lee, S., Li, Z., Gogarten, S.M., Sofer, T., Bielak, L.F., Bis, J.C., et al. (2019) Efficient variant set mixed model association tests for continuous and binary traits in large-scale whole-genome sequencing studies. The American Journal of Human Genetics 104, 260-274.

Chen, H., Wang, C., Conomos, M.P., Stilp, A.M., Li, Z., Sofer, T., Szpiro, A.A., Chen, W., Brehm, J.M., Celedón, J.C., Redline, S., Papanicolaou, G.J., Thornton, T.A., Laurie, C.C., Rice, K. and Lin, X. (2016) Control forpopulation structure and relatedness for binary traits in genetic association studies via logistic mixed models. The American Journal of Human Genetics 98, 653-666.

Gilmour, A.R., Thompson, R. and Cullis, B.R. (1995) Average Information REML: An Efficient Algorithm for Variance Parameter Estimation in Linear Mixed Models. Biometrics 51, 1440-1450.

Lee, S., Teslovich, T., Boehnke, M., Lin, X. (2013) General framework for meta-analysis of rare variants in sequencing association studies. The American Journal of Human Genetics 93, 42-53.

Lee, S., Wu, M.C., Lin, X. (2012) Optimal tests for rare variant effects in sequencing association studies. Biostatistics 13, 762-775.

Nelder, J.A. and Mead, R. (1965) A simplex algorithm for function minimization. Computer Journal 7, 308-313.

Sun, J., Zheng, Y., Hsu, L. (2013) A unified mixed-effects model for rare-variant association in sequencing studies. Genetic Epidemiology 37, 334-344.

Wu, M.C., Lee, S., Cai, T., Li, Y., Boehnke, M., Lin, X. (2011) Rare-variant association testing for sequencing data with the sequence kernel association test. The American Journal of Human Genetics 89, 82-93.

Yang, J., Lee, S.H., Goddard, M.E. and Visscher, P.M. (2011) GCTA: A Tool for Genome-wide Complex Trait Analysis. The American Journal of Human Genetics 88, 76-82.

Zhou, X. and Stephens, M. (2012) Genome-wide efficient mixed-model analysis for association studies. Nature Genetics 44, 821-824.

4 glmm.score

example

Example dataset

Description

Example dataset for GMMAT.

Format

Contains the following objects:

pheno a data frame of 400 observations from a cross-sectional study with 5 variables: id, disease, trait, age and sex.

pheno2 a data frame of 2,000 observations from a longitudinal study with 400 individuals and 5 variables: id, y.repeated, y.trend, time and sex.

GRM a genetic relationship matrix for 400 observations.

glmm.score

Performing GLMM based score tests

Description

Use a glmmkin class object from the null GLMM to perform score tests for association with genotypes in a plink .bed file (binary genotypes), a GDS file .gds, or a plain text file (or compressed .gz or .bz2 file).

Usage

```
glmm.score(obj, infile, outfile, BGEN.samplefile = NULL, center = T, select = NULL,
MAF.range = c(1e-7, 0.5), miss.cutoff = 1,
missing.method = "impute2mean", nperbatch = 100, tol = 1e-5,
infile.nrow = NULL, infile.nrow.skip = 0, infile.sep = "\t",
infile.na = "NA", infile.ncol.skip = 1, infile.ncol.print = 1,
infile.header.print = "SNP", is.dosage = FALSE, ncores = 1, verbose = FALSE)
```

Arguments

obj

a class glmmkin or class glmmkin.multi object, returned by fitting the null GLMM using glmmkin.

infile

the input file name or an object of class SeqVarGDSClass. Note that for plink binary genotype files only the prefix without .bed, .bim or .fam should be used. Only SNP major mode recognized in the binary file. Alternatively, it can be the full name of a BGEN file (including the suffix .bgen), a GDS file (including the suffix .gds), or a plain text file with some delimiters (comma, space, tab or something else), with one row for each SNP and one column for each individual.

glmm.score 5

In that case, SNPs should be coded as numeric values (0/1/2 or dosages allowed, A/C/G/T coding is not recognized). There can be additional rows and columns to skip at the beginning. The order of individuals can be different from obj in the null GLMM (see the argument select). Some compressed files (.gz and .bz2) also allowed. If infile is an object of class SeqVarGDSClass, the .gds file will be closed upon successful completion of the function.

outfile

the output file name.

BGEN.samplefile

path to the BGEN sample file. Required when the BGEN file does not contain sample identifiers or the select parameter is NULL (default = NULL).

center

a logical switch for centering genotypes before tests. If TRUE, genotypes will be centered to have mean 0 before tests, otherwise raw values will be directly used in tests (default = TRUE).

select

an optional vector indicating the order of individuals in infile. If supplied, the length must match the number of individuals in infile (default = NULL). Individuals to be excluded should be coded 0. For example, select = c(2, 3, 1, 0) means the 1st individual in infile corresponds to the 2nd individual in obj, the 2nd individual in infile corresponds to the 3rd individual in obj, the 3rd individual in infile corresponds to the 1st individual in obj, the 4th individual in infile is not included in obj. If there are any duplicated id_include in obj (longitudinal data analysis), indices in select should match the order of individuals with unique id_include in obj. For plink binary genotype files and GDS files, this argument is not required and the sample ID's are automatically matched.

MAF.range

a numeric vector of length 2 defining the minimum and maximum minor allele frequencies of variants that should be included in the analysis (default = c(1e-7, 0.5)).

miss.cutoff

the maximum missing rate allowed for a variant to be included (default = 1, including all variants).

missing.method

method of handling missing genotypes. Either "impute2mean" or "omit" (default = "impute2mean").

nperbatch

an integer for how many SNPs should be tested in a batch (default = 100). The computational time can increase dramatically if this value is either small or large. The optimal value for best performance depends on the user's system.

tol

the threshold for determining monomorphism. If a SNP has value range less than the tolerance, it will be considered monomorphic and its association test p-value will be NA (default = 1e-5). Only used when infile is a plain text file (or compressed .gz or .bz2 file).

infile.nrow

number of rows to read in infile, including number of rows to skip at the beginning. If NULL, the program will determine how many rows there are in infile automatically and read all rows (default = NULL). Only used when infile is a plain text file (or compressed .gz or .bz2 file).

infile.nrow.skip

number of rows to skip at the beginning of infile. Must be nonnegative integers. Useful when header or comment lines are present (default = 0). Only used when infile is a plain text file (or compressed .gz or .bz2 file).

6 glmm.score

infile.sep delimiter in infile (default = "\t"). Only used when infile is a plain text file

(or compressed .gz or .bz2 file).

infile.na symbol in infile to denote missing genotypes (default = "NA"). Only used when infile is a plain text file (or compressed .gz or .bz2 file).

infile.ncol.skip

number of columns to skip before genotype data in infile. These columns can be SNP name, alleles and/or quality measures and should be placed at the beginning in each line. After skipping these columns, the program will read in genotype data and perform score tests. Must be nonnegative integers. It is recommended that SNP name should be included as the first column in infile and genotype data should start from the second column or later (default = 1). Only used when infile is a plain text file (or compressed .gz or .bz2 file).

infile.ncol.print

a vector indicating which column(s) in infile should be printed to the output directly. These columns can be SNP name, alleles and/or quality measures placed at the beginning in each line. Must be nonnegative integers, no greater than infile.ncol.skip and sorted numerically in ascending order. By default, it is assumed that the first column is SNP name and genotype data start from the second column, and SNP name should be carried over to the output (default = 1). Should be set to NULL if infile.ncol.skip is 0. Only used when infile is a plain text file (or compressed .gz or .bz2 file).

infile.header.print

a character vector indicating column name(s) of column(s) selected to print by infile.ncol.print (default = "SNP"). Should be set to NULL if infile.ncol.skip

is 0. Only used when infile is a plain text file (or compressed .gz or .bz2 file).

is.dosage a logical switch for whether imputed dosage should be used from a GDS infile

(default = FALSE).

ncores a positive integer indicating the number of cores to be used in parallel computing

(default = 1).

verbose a logical switch for whether a progress bar should be shown for a GDS infile

(default = FALSE).

Value

NULL if infile is a BGEN file (.bgen) or a GDS file (.gds), otherwise computational time in seconds, excluding I/O time.

Author(s)

Han Chen, Duy T. Pham

References

Chen, H., Wang, C., Conomos, M.P., Stilp, A.M., Li, Z., Sofer, T., Szpiro, A.A., Chen, W., Brehm, J.M., Celedón, J.C., Redline, S., Papanicolaou, G.J., Thornton, T.A., Laurie, C.C., Rice, K. and Lin, X. (2016) Control forpopulation structure and relatedness for binary traits in genetic association studies via logistic mixed models. The American Journal of Human Genetics 98, 653-666.

glmm.score.meta 7

See Also

```
glmmkin, glmm.wald
```

Examples

```
data(example)
attach(example)
model0 <- glmmkin(disease ~ age + sex, data = pheno, kins = GRM, id = "id",
       family = binomial(link = "logit"))
plinkfiles <- strsplit(system.file("extdata", "geno.bed", package = "GMMAT"),</pre>
       ".bed", fixed = TRUE)[[1]]
outfile.bed <- tempfile()</pre>
glmm.score(model0, infile = plinkfiles, outfile = outfile.bed)
if(requireNamespace("SeqArray", quietly = TRUE) && requireNamespace("SeqVarTools",
        quietly = TRUE)) {
infile <- system.file("extdata", "geno.gds", package = "GMMAT")</pre>
outfile.gds <- tempfile()</pre>
glmm.score(model0, infile = infile, outfile = outfile.gds)
unlink(outfile.gds)
infile <- system.file("extdata", "geno.txt", package = "GMMAT")</pre>
outfile.text <- tempfile()</pre>
glmm.score(model0, infile = infile, outfile = outfile.text, infile.nrow.skip = 5,
infile.ncol.skip = 3, infile.ncol.print = 1:3,
infile.header.print = c("SNP", "Allele1", "Allele2"))
infile <- system.file("extdata", "geno.bgen", package = "GMMAT")</pre>
samplefile <- system.file("extdata", "geno.sample", package = "GMMAT")</pre>
outfile.bgen <- tempfile()</pre>
glmm.score(model0, infile = infile, BGEN.samplefile = samplefile,
        outfile = outfile.bgen)
unlink(c(outfile.bed, outfile.text, outfile.bgen))
```

glmm.score.meta

Performing meta-analysis for GLMM based score test results

Description

Use output files from GLMM based score tests to perform meta-analysis.

Usage

```
glmm.score.meta(files, outfile, SNP = rep("SNP", length(files)),
A1 = rep("A1", length(files)), A2 = rep("A2", length(files)))
```

8 glmm.score.meta

Arguments

files a vector of input file names. The input files should be the output files of glmm. score(

), or customized tab or space delimited files that include at least 8 columns: SNP, effect allele, noneffect allele, N, AF, SCORE, VAR and PVAL. The column names of SNP, effect allele and noneffect allele can be customized and

provided in SNP, A1 and A2.

outfile the output file name.

SNP a character vector of SNP column names in each input file. The length and order

must match the length and order of files (default = rep("SNP", length(files))).

A1 a character vector of allele 1 column names in each input file. The length and or-

der must match the length and order of files (default = rep("A1", length(files))). Note that glmm.score.meta() does not define A1 as the effect allele or non-effect allele: it is the user's choice. However, the choice should be consistent across different studies, if A1 column is the effect allele in one study but the

noneffect allele in another, meta-analysis results will be incorrect.

A2 a character vector of allele 2 column names in each input file. The length and or-

der must match the length and order of files (default = rep("A2", length(files))). Note that glmm.score.meta() does not define A2 as the effect allele or non-effect allele: it is the user's choice. However, the choice should be consistent across different studies, if A2 column is the effect allele in one study but the

noneffect allele in another, meta-analysis results will be incorrect.

Value

a data frame containing the following:

SNP SNP name.

A1 allele 1.

A2 allele 2.

N total sample size.

AF effect allele frequency (user-defined: can be either allele 1 or allele 2).

SCORE the summary score of the effect allele.

VAR the variance of the summary score.

PVAL meta-analysis p-value.

Author(s)

Han Chen

See Also

glmm.score

Examples

```
infile1 <- system.file("extdata", "meta1.txt", package = "GMMAT")
infile2 <- system.file("extdata", "meta2.txt", package = "GMMAT")
infile3 <- system.file("extdata", "meta3.txt", package = "GMMAT")
outfile <- tempfile()
glmm.score.meta(files = c(infile1, infile2, infile3), outfile = outfile,
SNP = rep("SNP", 3), A1 = rep("A1", 3), A2 = rep("A2", 3))
unlink(outfile)</pre>
```

glmm.wald

Performing GLMM based Wald tests

Description

Fit a GLMM under the alternative hypothesis to perform Wald tests for association with genotypes in a plink .bed file (binary genotypes), a GDS file .gds, or a plain text file (or compressed .gz or .bz2 file).

Usage

```
glmm.wald(fixed, data = parent.frame(), kins = NULL, id, random.slope = NULL,
groups = NULL, family = binomial(link = "logit"), infile, snps,
method = "REML", method.optim = "AI", maxiter = 500, tol = 1e-5,
taumin = 1e-5, taumax = 1e5, tauregion = 10, center = T,
select = NULL, missing.method = "impute2mean", infile.nrow = NULL,
infile.nrow.skip = 0, infile.sep = "\t", infile.na = "NA",
snp.col = 1, infile.ncol.skip = 1, infile.ncol.print = 1,
infile.header.print = "SNP", is.dosage = FALSE, verbose = FALSE, ...)
```

Arguments

fixed	an object of class formula (or one that can be coerced to that class): a symbolic description of the fixed effects model (without including any snps to be tested) to be fitted.
data	a data frame or list (or object coercible by as.data.frame to a data frame) containing the variables in the model.
kins	a known positive semi-definite relationship matrix (e.g. kinship matrix in genetic association studies) or a list of known positive semi-definite relationship matrices. The rownames and colnames of these matrices must at least include all samples as specified in the id column of the data frame data. If not provided, glmmkin will switch to the generalized linear model with no random effects (default = NULL).
id	a column in the data frame data, indicating the id of samples. When there are duplicates in id, the data is assumed to be longitudinal with repeated measures.

random.slope an optional column indicating the random slope for time effect used in a mixed

> effects model for cross-sectional data with related individuals, and longitudinal data. It must be included in the names of data. There must be duplicates in id

and method.optim must be "AI" (default = NULL).

an optional categorical variable indicating the groups used in a heteroscedastic groups

> linear mixed model (allowing residual variances in different groups to be different). This variable must be included in the names of data, and family must be

"gaussian" and method.optim must be "AI" (default = NULL).

family a description of the error distribution and link function to be used in the model.

This can be a character string naming a family function, a family function or the result of a call to a family function. (See family for details of family functions.)

infile the input file name. Note that for plink binary genotype files only the prefix

without .bed, .bim or .fam should be used. Only SNP major mode recognized in the binary file. Alternatively, it can be the full name of a GDS file (including the suffix .gds) or a plain text file with some delimiters (comma, space, tab or something else), with one row for each SNP and one column for each individual. In that case, SNPs should be coded as numeric values (0/1/2 or dosages allowed, A/C/G/T coding is not recognized). There can be additional rows and columns to skip at the beginning. The order of individuals can be different from obj in the null GLMM (see the argument select). Some compressed files (.gz and

.bz2) also allowed.

a vector of SNP names to be tested. snps

method of fitting the generalized linear mixed model. Either "REML" or "ML"

(default = "REML").

method.optim optimization method of fitting the generalized linear mixed model. Either "AI",

"Brent" or "Nelder-Mead" (default = "AI").

a positive integer specifying the maximum number of iterations when fitting the

generalized linear mixed model (default = 500).

a positive number specifying tolerance, the difference threshold for parameter tol

> estimates below which iterations should be stopped. Also the threshold for determining monomorphism. If a SNP has value range less than the tolerance, it will be considered monomorphic and its association test p-value will be NA

(default = 1e-5).

taumin the lower bound of search space for the variance component parameter τ (default

= 1e-5), used when method.optim = "Brent". See glmmkin.

the upper bound of search space for the variance component parameter τ (default taumax

= 1e5), used when method.optim = "Brent". See glmmkin.

the number of search intervals for the REML or ML estimate of the variance tauregion

component parameter τ (default = 10), used when method.optim = "Brent".

See glmmkin.

a logical switch for centering genotypes before tests. If TRUE, genotypes will center

be centered to have mean 0 before tests, otherwise raw values will be directly

used in tests (default = TRUE).

method

maxiter

select

an optional vector indicating the order of individuals in infile. If supplied, the length must match the number of individuals in infile (default = NULL). Individuals to be excluded should be coded 0. For example, select = c(2, 3, 1, 0) means the 1st individual in infile corresponds to the 2nd individual in data, the 2nd individual in infile corresponds to the 3rd individual in data, the 3rd individual in infile corresponds to the 1st individual in data, the 4th individual in infile is not included in data. If there are any duplicated id in data (longitudinal data analysis), indices in select should match the order of individuals with unique id in data. For plink binary genotype files and GDS files, this argument is not required and the sample ID's are automatically matched.

missing.method

method of handling missing genotypes. Either "impute2mean" or "omit" (default = "impute2mean").

infile.nrow

number of rows to read in infile, including number of rows to skip at the beginning. If NULL, the program will determine how many rows there are in infile automatically and read all rows (default = NULL). Only used when infile is a plain text file (or compressed .gz or .bz2 file).

infile.nrow.skip

number of rows to skip at the beginning of infile. Must be nonnegative integers. Useful when header or comment lines are present (default = 0). Only used when infile is a plain text file (or compressed .gz or .bz2 file).

infile.sep

delimiter in infile (default = "\t"). Only used when infile is a plain text file (or compressed .gz or .bz2 file).

infile.na

symbol in infile to denote missing genotypes (default = "NA"). Only used when infile is a plain text file (or compressed .gz or .bz2 file).

snp.col

a positive integer specifying which column in infile is SNP names. Only used when infile is a plain text file (or compressed .gz or .bz2 file).

infile.ncol.skip

number of columns to skip before genotype data in infile. These columns can be SNP name, alleles and/or quality measures and should be placed at the beginning in each line. After skipping these columns, the program will read in genotype data and perform Wald tests. Must be positive integers. It is recommended that SNP name should be included as the first column in infile and genotype data should start from the second column or later (default = 1). Only used when infile is a plain text file (or compressed .gz or .bz2 file).

infile.ncol.print

a vector indicating which column(s) in infile should be shown in the results. These columns can be SNP name, alleles and/or quality measures placed at the beginning in each line. Must be positive integers, no greater than infile.ncol.skip and sorted numerically in ascending order. By default, it is assumed that the first column is SNP name and genotype data start from the second column, and SNP name should be carried over to the results (default = 1). Only used when infile is a plain text file (or compressed .gz or .bz2 file).

infile.header.print

a character vector indicating column name(s) of column(s) selected to print by infile.ncol.print (default = "SNP"). Only used when infile is a plain text file (or compressed .gz or .bz2 file).

is. dosage a logical switch for whether imputed dosage should be used from a GDS infile

(default = FALSE).

verbose a logical switch for printing a progress bar and detailed information (param-

eter estimates in each iteration) for testing and debugging purpose (default =

FALSE).

... additional arguments that could be passed to glm.

Value

if infile is a plain text file, a data frame containing variables included in infile.header.print and the following:

N number of individuals with non-missing genotypes for each SNP.

AF effect allele frequency for each SNP.

BETA effect size estimate for each SNP from the GLMM under the alternative hypoth-

esis.

SE standard error of the effect size estimate for each SNP.

PVAL Wald test p-value for each SNP.

converged a logical indicator for convergence for each SNP.

if infile is the prefix of plink binary files (.bed, .bim and .fam), a data frame containing the following:

CHR Chromosome, copied from .bim file.

SNP SNP name, as supplied in snps.

cM genetic location in centi Morgans, copied from .bim file.

POS physical position in base pairs, copied from .bim file.

A1 allele 1, copied from .bim file.
A2 allele 2, copied from .bim file.

N number of individuals with non-missing genotypes for each SNP.

AF effect allele frequency for each SNP.

BETA effect size estimate for each SNP from the GLMM under the alternative hypoth-

esis.

SE standard error of the effect size estimate for each SNP.

PVAL Wald test p-value for each SNP.

converged a logical indicator for convergence for each SNP.

if infile is a GDS file (.gds), a data frame containing the following:

SNP SNP name, as supplied in snps.

CHR Chromosome, copied from .gds file.

POS physical position in base pairs, copied from .gds file.

REF reference allele, copied from .gds file.

ALT alternate allele, copied from .gds file.

	1 6: 1: 1 1 1.1		C 1 CATE
N	number of individuals with	non-missing genotynes	for each SNP

AF ALT allele frequency for each SNP.

BETA effect size estimate for each SNP from the GLMM under the alternative hypoth-

esis.

SE standard error of the effect size estimate for each SNP.

PVAL Wald test p-value for each SNP.

converged a logical indicator for convergence for each SNP.

Author(s)

Han Chen, Matthew P. Conomos

References

Brent, R.P. (1973) "Chapter 4: An Algorithm with Guaranteed Convergence for Finding a Zero of a Function", Algorithms for Minimization without Derivatives, Englewood Cliffs, NJ: Prentice-Hall, ISBN 0-13-022335-2.

Breslow, N.E. and Clayton, D.G. (1993) Approximate Inference in Generalized Linear Mixed Models. Journal of the American Statistical Association 88, 9-25.

Chen, H., Wang, C., Conomos, M.P., Stilp, A.M., Li, Z., Sofer, T., Szpiro, A.A., Chen, W., Brehm, J.M., Celedón, J.C., Redline, S., Papanicolaou, G.J., Thornton, T.A., Laurie, C.C., Rice, K. and Lin, X. (2016) Control forpopulation structure and relatedness for binary traits in genetic association studies via logistic mixed models. The American Journal of Human Genetics 98, 653-666.

Gilmour, A.R., Thompson, R. and Cullis, B.R. (1995) Average Information REML: An Efficient Algorithm for Variance Parameter Estimation in Linear Mixed Models. Biometrics 51, 1440-1450.

Nelder, J.A. and Mead, R. (1965) A simplex algorithm for function minimization. Computer Journal 7, 308-313.

Yang, J., Lee, S.H., Goddard, M.E. and Visscher, P.M. (2011) GCTA: A Tool for Genome-wide Complex Trait Analysis. The American Journal of Human Genetics 88, 76-82.

Zhou, X. and Stephens, M. (2012) Genome-wide efficient mixed-model analysis for association studies. Nature Genetics 44, 821-824.

See Also

```
glmmkin, glmm.score
```

Examples

glmmkin

Fit generalized linear mixed model with known relationship matrices

Description

Fit a generalized linear mixed model with a random intercept, or a random intercept and an optional random slope of time effect for longitudinal data. The covariance matrix of the random intercept is proportional to a known relationship matrix (e.g. kinship matrix in genetic association studies). Alternatively, it can be a variance components model with multiple random effects, and each component has a known relationship matrix.

Usage

```
glmmkin(fixed, data = parent.frame(), kins = NULL, id, random.slope = NULL,
groups = NULL, family = binomial(link = "logit"), method = "REML",
method.optim = "AI", maxiter = 500, tol = 1e-5, taumin = 1e-5,
taumax = 1e5, tauregion = 10, verbose = FALSE, ...)
```

Arguments

_	
fixed	an object of class formula (or one that can be coerced to that class): a symbolic description of the fixed effects model to be fitted.
data	a data frame or list (or object coercible by as.data.frame to a data frame) containing the variables in the model.
kins	a known positive semi-definite relationship matrix (e.g. kinship matrix in genetic association studies) or a list of known positive semi-definite relationship matrices. The rownames and colnames of these matrices must at least include all samples as specified in the id column of the data frame data. If not provided, glmmkin will switch to the generalized linear model with no random effects (default = NULL).
id	a column in the data frame data, indicating the id of samples. When there are duplicates in id, the data is assumed to be longitudinal with repeated measures.
random.slope	an optional column indicating the random slope for time effect used in a mixed effects model for cross-sectional data with related individuals, and longitudinal data. It must be included in the names of data. There must be duplicates in id and method.optim must be "AI" (default = NULL).

on ontional actagorical variable indicating the groups used in a heterogoadastic

groups	an optional categorical variable indicating the groups used in a heteroscedastic linear mixed model (allowing residual variances in different groups to be different). This variable must be included in the names of data, and family must be "gaussian" and method.optim must be "AI" (default = NULL).
family	a description of the error distribution and link function to be used in the model. This can be a character string naming a family function, a family function or the result of a call to a family function. (See family for details of family functions.)
method	method of fitting the generalized linear mixed model. Either "REML" or "ML" (default = "REML").
method.optim	optimization method of fitting the generalized linear mixed model. Either "AI", "Brent" or "Nelder-Mead" (default = "AI").
maxiter	a positive integer specifying the maximum number of iterations when fitting the generalized linear mixed model (default = 500).
tol	a positive number specifying tolerance, the difference threshold for parameter estimates below which iterations should be stopped (default = 1e-5).
taumin	the lower bound of search space for the variance component parameter τ (default = 1e-5), used when method.optim = "Brent". See Details.
taumax	the upper bound of search space for the variance component parameter τ (default = 1e5), used when method.optim = "Brent". See Details.
tauregion	the number of search intervals for the REML or ML estimate of the variance component parameter τ (default = 10), used when method.optim = "Brent". See Details.
verbose	a logical switch for printing detailed information (parameter estimates in each iteration) for testing and debugging purpose (default = FALSE).
	additional arguments that could be passed to glm.

Details

Generalized linear mixed models (GLMM) are fitted using the penalized quasi-likelihood (PQL) method proposed by Breslow and Clayton (1993). Generally, fitting a GLMM is computationally expensive, and by default we use the Average Information REML algorithm (Gilmour, Thompson and Cullis, 1995; Yang et al., 2011) to fit the model. If only one relationship matrix is specified (kins is a matrix), iterations may be accelerated using the algorithm proposed by Zhou and Stephens (2012) for linear mixed models. An eigendecomposition is performed in each outer iteration and the estimate of the variance component parameter τ is obtained by maximizing the profiled log restricted likelihood (or likelihood) in a search space from taumin to taumax, equally divided into tauregion intervals on the log scale, using Brent's method (1973). If kins is a list of matrices and method = "Nelder-Mead", iterations are performed as a multi-dimensional maximization problem solved by Nelder and Mead's method (1965). It can be very slow, and we do not recommend using this method unless the likelihood function is badly behaved. Both Brent's method and Nelder and Mead's method are derivative-free. When the Average Information REML algorithm fails to converge, a warning message is given and the algorithm is default to derivative-free approaches: Brent's method if only one relationship matrix is specified, Nelder and Mead's method if more than one relationship matrix is specified.

For longitudinal data (with duplicated id), two types of models can be applied: random intercept only models, and random intercept and random slope models. The random intercept only model

is appropriate for analyzing repeated measures with no time trends, and observations for the same individual are assumed to be exchangeable. The random intercept and random slope model is appropriate for analyzing longitudinal data with individual-specific time trends (therefore, a random slope for time effect). Typically, the time effect should be included in the model as a fixed effect covariate as well. Covariances of the random intercept and the random slope are estimated.

For multiple phenotype analysis, formula recognized by lm, such as cbind(y1, y2, y3) ~ x1 + x2, can be used in fixed as fixed effects. For each matrix in kins, variance components corresponding to each phenotype, as well as their covariance components, will be estimated. Currently, family must be "gaussian" and method.optim must be "AI".

Value

theta

a vector or a list of variance component parameter estimates. See below.

For cross-sectional data, if kins is not provided (unrelated individuals), theta is the dispersion parameter estimate from the generalized linear model; if kins is a matrix and groups is not provided, theta is a length 2 vector, with theta[1] being the dispersion parameter estimate and theta[2] being the variance component parameter estimate for kins; if kins is a list and groups is not provided, theta is a length 1 + length(kins) vector, with theta[1] being the dispersion parameter estimate and theta[2:(1 + length(kins))] being the variance component parameter estimates, corresponding to the order of matrices in the list kins; if kins is a matrix and groups is provided (a heteroscedastic linear mixed model with n. groups residual variance groups), theta is a length 1 + n. groups vector, with theta[1:n.groups] being the residual variance estimates for each group and theta[1 + n.groups] being the variance component parameter estimate for kins; if kins is a list and groups is provided (a heteroscedastic linear mixed model with n.groups residual variance groups), theta is a length length(kins) + n.groups vector, with theta[1:n.groups] being the residual variance estimates for each group and theta[(1 + n.groups):(length(kins) + n.groups)] being the variance component parameter estimates, corresponding to the order of matrices in the list kins.

For longitudinal data (with duplicated id) in a random intercept only model, if kins is not provided (unrelated individuals) and groups is not provided, theta is a length 2 vector, with theta[1] being the dispersion parameter estimate and theta[2] being the variance component parameter estimate for the random individual effects; if kins is a matrix and groups is not provided, theta is a length 3 vector, with theta[1] being the dispersion parameter estimate, theta[2] being the variance component parameter estimate for the random individual effects attributable to relatedness from kins, and theta[3] being the variance component parameter estimate for the random individual effects not attributable to relatedness from kins; if kins is a list and groups is not provided, theta is a length 2 + length(kins) vector, with theta[1] being the dispersion parameter estimate, theta[2:(1 + length(kins))] being the variance component parameter estimates for the random individual effects attributable to relatedness from kins, corresponding to the order of matrices in the list kins, and theta[2 + length(kins)] being the variance component parameter estimate for the random individual effects not attributable to relatedness from kins; if kins is not provided (unrelated individuals) and groups is pro-

vided (a heteroscedastic linear mixed model with n.groups residual variance groups), theta is a length 1 + n.groups vector, with theta[1:n.groups] being the residual variance estimates for each group and theta[1 + n.groups] being the variance component parameter estimate for the random individual effects; if kins is a matrix and groups is provided (a heteroscedastic linear mixed model with n.groups residual variance groups), theta is a length 2 + n.groups vector, with theta[1:n.groups] being the residual variance estimates for each group, theta[1 + n.groups] being the variance component parameter estimate for the random individual effects attributable to relatedness from kins, and theta[2 + n.groups] being the variance component parameter estimate for the random individual effects not attributable to relatedness from kins; if kins is a list and groups is provided (a heteroscedastic linear mixed model with n.groups residual variance groups), theta is a length 1 + length(kins) + n.groups vector, with theta[1:n.groups] being the residual variance estimates for each group, theta[(1 + n.groups):(length(kins) + n.groups)] being the variance component parameter estimates for the random individual effects attributable to relatedness from kins, corresponding to the order of matrices in the list kins, and theta[1 + length(kins) + n.groups] being the variance component parameter estimate for the random individual effects not attributable to relatedness from kins.

For longitudinal data (with duplicated id) in a random intercept and random slope (for time effect) model, if kins is not provided (unrelated individuals) and groups is not provided, theta is a length 4 vector, with theta[1] being the dispersion parameter estimate, theta[2] being the variance component parameter estimate for the random individual effects of the intercept, theta[3] being the covariance estimate for the random individual effects of the intercept and the random individual effects of the time slope, and theta[4] being the variance component parameter estimate for the random individual effects of the time slope; if kins is a matrix and groups is not provided, theta is a length 7 vector, with theta[1] being the dispersion parameter estimate, theta[2] being the variance component parameter estimate for the random individual effects of the intercept attributable to relatedness from kins, theta[3] being the variance component parameter estimate for the random individual effects of the intercept not attributable to relatedness from kins, theta[4] being the covariance estimate for the random individual effects of the intercept and the random individual effects of the time slope attributable to relatedness from kins, theta[5] being the covariance estimate for the random individual effects of the intercept and the random individual effects of the time slope not attributable to relatedness from kins, theta[6] being the variance component parameter estimate for the random individual effects of the time slope attributable to relatedness from kins, and theta[7] being the variance component parameter estimate for the random individual effects of the time slope not attributable to relatedness from kins; if kins is a list and groups is not provided, theta is a length 4 + 3 * length(kins) vector, with theta[1] being the dispersion parameter estimate, theta[2:(1 + length(kins))] being the variance component parameter estimates for the random individual effects of the intercept attributable to relatedness from kins, corresponding to the order of matrices in the list kins, theta[2 + length(kins)] being the variance

component parameter estimate for the random individual effects of the intercept not attributable to relatedness from kins, theta[(3 + length(kins)):(2 + 2 * length(kins))] being the covariance estimates for the random individual effects of the intercept and the random individual effects of the time slope attributable to relatedness from kins, corresponding to the order of matrices in the list kins, theta[3 + 2 * length(kins)] being the covariance estimate for the random individual effects of the intercept and the random individual effects of the time slope not attributable to relatedness from kins, theta[(4 + 2 * length(kins): (3 + 3 * length(kins))] being the variance component parameter estimates for the random individual effects of the time slope attributable to relatedness from kins, corresponding to the order of matrices in the list kins, theta[4 + 3 * length(kins)] being the variance component parameter estimate for the random individual effects of the time slope not attributable to relatedness from kins; if kins is not provided (unrelated individuals) and groups is provided (a heteroscedastic linear mixed model with n. groups residual variance groups), theta is a length 3 + n.groups vector, with theta[1:n.groups] being the residual variance estimates for each group, theta[1 + n.groups] being the variance component parameter estimate for the random individual effects of the intercept, theta[2 + n.groups] being the covariance estimate for the random individual effect of the intercept and the random individual effects of the time slope, and theta[3 + n.groups] being the variance component parameter estimate for the random individual effects of the time slope; if kins is a matrix and groups is provided (a heteroscedastic linear mixed model with n.groups residual variance groups), theta is a length 6 + n.groups vector, with theta[1:n.groups] being the residual variance estimates for each group, theta[1 + n.groups] being the variance component parameter estimate for the random individual effects of the intercept attributable to relatedness from kins, theta[2 + n.groups] being the variance component parameter estimate for the random individual effects of the intercept not attributable to relatedness from kins, theta[3 + n.groups] being the covariance estimate for the random individual effects of the intercept and the random individual effects of the time slope attributable to relatedness from kins, theta[4 + n.groups] being the covariance estimate for the random individual effects of the intercept and the random individual effects of the time slope not attributable to relatedness from kins, theta[5 + n.groups] being the variance component parameter estimate for the random individual effects of the time slope attributable to relatedness from kins, and theta[6+n.groups] being the variance component parameter estimate for the random individual effects of the time slope not attributable to relatedness from kins; if kins is a list and groups is provided (a heteroscedastic linear mixed model with n. groups residual variance groups), theta is a length 3 + 3 * length(kins) + n.groups vector, with theta[1:n.groups] being the residual variance estimates for each group, theta[(1 + n.groups):(length(kins) + n.groups)] being the variance component parameter estimates for the random individual effects of the intercept attributable to relatedness from kins, corresponding to the order of matrices in the list kins, theta[1 + length(kins) + n.groups] being the variance component parameter estimate for the random individual effects of the intercept not attributable to relatedness from kins, theta[(2 + length(kins) + n.groups):(1 + 2 * length(kins) + n.groups)]

being the covariance estimates for the random individual effects of the intercept and the random individual effects of the time slope attributable to relatedness from kins, corresponding to the order of matrices in the list kins, theta[2+2*length(kins) + n.groups] being the covariance estimate for the random individual effects of the intercept and the random individual effects of the time slope not attributable to relatedness from kins, theta[(3+2*length(kins)+n.groups)] being the variance component parameter estimates for the random individual effects of the time slope attributable to relatedness from kins, corresponding to the order of matrices in the list kins, and theta[3+3*length(kins)+n.groups] being the variance component parameter estimate for the random individual effects of the time slope not attributable to relatedness from kins.

For multiple phenotype analysis, theta is a list of variance-covariance matrices. If kins is not provided (unrelated individuals), theta is an n.pheno by n.pheno variance-covariance matrix for the residuals of the multiple phenotypes from the linear model; if kins is a matrix and groups is not provided, theta is a length 2 list, with theta[[1]] being the variance-covariance matrix for the residuals and theta[[2]] being the variance-covariance matrix for kins; if kins is a list and groups is not provided, theta is a length 1 + length(kins) list, with theta[[1]] being the variance-covariance matrix for the residuals and theta[[2]] to theta[[1 + length(kins)]] being the variance-covariance matrices, corresponding to the order of matrices in the list kins; if kins is a matrix and groups is provided (a heteroscedastic linear mixed model with n. groups residual variance groups), theta is a length 1 + n. groups list, with theta[[1]] to theta[[n.groups]] being the variance-covariance matrices for the residuals in each group and theta[[1 + n.groups]] being the variance-covariance matrix for kins; if kins is a list and groups is provided (a heteroscedastic linear mixed model with n.groups residual variance groups), theta is a length length(kins) + n.groups list, with theta[[1]] to theta[[n.groups]] being the variance-covariance matrices for the residuals in each group and theta[[1 + n.groups]] to theta[[length(kins) + n.groups]] being the variance-covariance matrices, corresponding to the order of matrices in the list kins.

n.pheno

an integer indicating the number of phenotypes in multiple phenotype analysis (for single phenotype analysis, n. pheno = 1).

n.groups

an integer indicating the number of distinct residual variance groups in heteroscedastic linear mixed models (for other models, n. groups = 1).

coefficients

a vector or a matrix for the fixed effects parameter estimates (including the intercept).

linear.predictors

a vector or a matrix for the linear predictors.

fitted.values a vector or a matrix for the fitted mean values on the original scale.

Y a vector or a matrix for the final working vector.

X model matrix for the fixed effects.

P the projection matrix with dimensions equal to the sample size multiplied by n.pheno. Used in glmm.score and SMMAT for dense matrices.

residuals a vector or a matrix for the residuals on the original scale. NOT rescaled by the

dispersion parameter.

scaled.residuals

a vector or a matrix for the scaled residuals, calculated as the original residuals divided by the dispersion parameter (in heteroscedastic linear mixed models,

corresponding residual variance estimates by each group).

cov covariance matrix for the fixed effects (including the intercept).

Sigma_i the inverse of the estimated covariance matrix for samples, with dimensions

equal to the sample size multiplied by n. pheno. Used in glmm. score and SMMAT

for sparse matrices.

Sigma_iX Sigma_i multiplied by X. Used in glmm. score and SMMAT for sparse matrices.

converged a logical indicator for convergence.

call the matched call.

id_include a vector indicating the id of rows in data with nonmissing outcome and covari-

ates, thus are included in the model fit.

Author(s)

Han Chen, Matthew P. Conomos

References

Brent, R.P. (1973) "Chapter 4: An Algorithm with Guaranteed Convergence for Finding a Zero of a Function", Algorithms for Minimization without Derivatives, Englewood Cliffs, NJ: Prentice-Hall, ISBN 0-13-022335-2.

Breslow, N.E. and Clayton, D.G. (1993) Approximate Inference in Generalized Linear Mixed Models. Journal of the American Statistical Association 88, 9-25.

Chen, H., Wang, C., Conomos, M.P., Stilp, A.M., Li, Z., Sofer, T., Szpiro, A.A., Chen, W., Brehm, J.M., Celedón, J.C., Redline, S., Papanicolaou, G.J., Thornton, T.A., Laurie, C.C., Rice, K. and Lin, X. (2016) Control for population structure and relatedness for binary traits in genetic association studies via logistic mixed models. The American Journal of Human Genetics 98, 653-666.

Gilmour, A.R., Thompson, R. and Cullis, B.R. (1995) Average Information REML: An Efficient Algorithm for Variance Parameter Estimation in Linear Mixed Models. Biometrics 51, 1440-1450.

Nelder, J.A. and Mead, R. (1965) A simplex algorithm for function minimization. Computer Journal 7, 308-313.

Yang, J., Lee, S.H., Goddard, M.E. and Visscher, P.M. (2011) GCTA: A Tool for Genome-wide Complex Trait Analysis. The American Journal of Human Genetics 88, 76-82.

Zhou, X. and Stephens, M. (2012) Genome-wide efficient mixed-model analysis for association studies. Nature Genetics 44, 821-824.

Examples

```
data(example)
attach(example)
model0 <- glmmkin(disease ~ age + sex, data = pheno, kins = GRM, id = "id",</pre>
```

SMMAT

Variant Set Mixed Model Association Tests (SMMAT)

Description

Variant Set Mixed Model Association Tests (SMMAT-B, SMMAT-S, SMMAT-O and SMMAT-E) for multiple user-defined test units and a null generalized linear mixed model. SMMAT.prep and SMMAT.lowmem are the two-step low-memory version of SMMAT.SMMAT.lowmem takes the returned R object from SMMAT.prep and uses less memory (if the returned R object from SMMAT.prep is saved to an R data file, the R session is terminated, and this R object is loaded into a new R session for running SMMAT.lowmem), especially when group.file contains only a subset of variants from geno.file.

Usage

```
SMMAT(null.obj, geno.file, group.file, group.file.sep = "\t",
meta.file.prefix = NULL, MAF.range = c(1e-7, 0.5),
MAF.weights.beta = c(1, 25), miss.cutoff = 1,
missing.method = "impute2mean", method = "davies";
tests = "E", rho = c(0, 0.1^2, 0.2^2, 0.3^2, 0.4^2,
0.5^2, 0.5, 1), use.minor.allele = FALSE,
auto.flip = FALSE, Garbage.Collection = FALSE,
is.dosage = FALSE, ncores = 1, verbose = FALSE)
SMMAT.prep(null.obj, geno.file, group.file, group.file.sep = "\t",
auto.flip = FALSE)
SMMAT.lowmem(SMMAT.prep.obj, geno.file = NULL, meta.file.prefix = NULL,
                           MAF.range = c(1e-7, 0.5), MAF.weights.beta = c(1, 25),
miss.cutoff = 1, missing.method = "impute2mean",
method = "davies", tests = "E", rho = c(0, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2, 0.1^2,
0.2^2, 0.3^2, 0.4^2, 0.5^2, 0.5, 1),
use.minor.allele = FALSE, Garbage.Collection = FALSE,
is.dosage = FALSE, ncores = 1, verbose = FALSE)
```

Arguments

null.obj a class glmmkin or a class glmmkin.multi object, returned by fitting the null

GLMM using glmmkin.

the .gds file name or an object of class SeqVarGDSClass for the full genotypes. geno.file The sample.id in geno.file should overlap id_include in null.obj. It is recommended that sample.id in geno.file include the full samples (at least

all samples as specified in id_include of null.obj). It is not necessary for the user to take a subset of geno. file before running the analysis. If geno. file is an object of class SeqVarGDSClass, the .gds file will be closed upon successful

completion of the function.

group.file a plain text file with 6 columns defining the test units. There should be no

headers in the file, and the columns are group name, chromosome, position,

reference allele, alternative allele and weight, respectively.

group.file.sep the delimiter in group.file (default = "\t").

meta.file.prefix

prefix of intermediate files (.score.* and .var.*) required in a meta-analysis. If

NULL, such intermediate files are not generated (default = NULL).

MAF.range a numeric vector of length 2 defining the minimum and maximum minor allele frequencies of variants that should be included in the analysis (default = c(1e-7),

0.5)).

MAF.weights.beta

a numeric vector of length 2 defining the beta probability density function parameters on the minor allele frequencies. This internal minor allele frequency weight is multiplied by the external weight given by the group.file. To turn off internal minor allele frequency weight and only use the external weight given

by the group file, use c(1, 1) to assign flat weights (default = c(1, 25)).

miss.cutoff the maximum missing rate allowed for a variant to be included (default = 1, including all variants).

method of handling missing genotypes. Either "impute2mean" or "impute2zero" missing.method

(default = "impute2mean").

method a method to compute p-values for SKAT-type test statistics (default = "davies").

> "davies" represents an exact method that computes a p-value by inverting the characteristic function of the mixture chisq distribution, with an accuracy of 1e-6. When "davies" p-value is less than 1e-5, it defaults to method "kuonen". "kuonen" represents a saddlepoint approximation method that computes the tail probabilities of the mixture chisq distribution. When "kuonen" fails to compute a p-value, it defaults to method "liu". "liu" is a moment-matching approximation

method for the mixture chisq distribution.

a character vector indicating which SMMAT tests should be performed ("B" for tests

> the burden test, "S" for SKAT, "O" for SKAT-O and "E" for the efficient hybrid test of the burden test and SKAT). The burden test and SKAT are automatically included when performing "O", and the burden test is automatically included

when performing "E" (default = "E").

rho a numeric vector defining the search grid used in SMMAT-O for SKAT-O (see the SKAT-O paper for details). Not used for SMMAT-B for the burden test,

SMMAT-S for SKAT or SMMAT-E for the efficient hybrid test of the burden test and SKAT (default = $c(0, 0.1^2, 0.2^2, 0.3^2, 0.4^2, 0.5^2, 0.5, 1)$).

use.minor.allele

a logical switch for whether to use the minor allele (instead of the alt allele) as the coding allele (default = FALSE). It does not change SMMAT-S results, but SMMAT-B (as well as SMMAT-O and SMMAT-E) will be affected. Along with the MAF filter, this option is useful for combining rare mutations, assuming rare

allele effects are in the same direction.

auto.flip a logical switch for whether to enable automatic allele flipping if a variant with

alleles ref/alt is not found at a position, but a variant at the same position with alleles alt/ref is found (default = FALSE). Use with caution for whole genome sequence data, as both ref/alt and alt/ref variants at the same position are not uncommon, and they are likely two different variants, rather than allele flipping.

Garbage.Collection

a logical switch for whether to enable garbage collection in each test (default =

FALSE). Pay for memory efficiency with slower computation speed.

is.dosage a logical switch for whether imputed dosage should be used from geno.file

(default = FALSE).

ncores a positive integer indicating the number of cores to be used in parallel computing

(default = 1).

verbose a logical switch for whether a progress bar should be shown (default = FALSE).

SMMAT.prep.obj a class SMMAT.prep object, returned by SMMAT.prep.

Value

SMMAT and SMMAT. lowmem return a data frame with the following components:

group name of the test unit group.

n.variants number of variants in the test unit group that pass the missing rate and allele

frequency filters.

miss.min minimum missing rate for variants in the test unit group.

miss.mean mean missing rate for variants in the test unit group.

miss.max maximum missing rate for variants in the test unit group.

freq.min minimum coding allele frequency for variants in the test unit group.

freq.mean mean coding allele frequency for variants in the test unit group.

freq.max maximum coding allele frequency for variants in the test unit group.

B. score burden test score statistic.

B. var variance of burden test score statistic.

B. pval burden test p-value.S. pval SKAT p-value.O. pval SKAT-O p-value.

0.minp minimum p-value in the SKAT-O search grid.

0.minp.rho rho value at the minimum p-value in the SKAT-O search grid.

E. pval SMMAT efficient hybrid test of the burden test and SKAT p-value.

SMMAT.prep return a list with the following components:

null.obj a class glmmkin or a class glmmkin.multi object from the null model, after pre-

processing.

geno.file the name of the .gds file for the full genotypes.

group.file the name of the plain text file with 6 columns defining the test units.

group.file.sep the delimiter in group.file.

auto.flip a logical indicator showing whether automatic allele flipping is enabled in pre-

processing if a variant with alleles ref/alt is not found at a position, but a variant

at the same position with alleles alt/ref is found.

residuals residuals from the null model, after pre-processing.

sample.id sample.id from geno.file, after pre-processing.

group.info group.info read from group.file, after pre-processing.

groups unique groups in group.info, after pre-processing.

group.idx.start

a vector of the start variant index for each group, after pre-processing.

group.idx.end a vector of the end variant index for each group, after pre-processing.

Author(s)

Han Chen

References

Wu, M.C., Lee, S., Cai, T., Li, Y., Boehnke, M., Lin, X. (2011) Rare-variant association testing for sequencing data with the sequence kernel association test. The American Journal of Human Genetics 89, 82-93.

Lee, S., Wu, M.C., Lin, X. (2012) Optimal tests for rare variant effects in sequencing association studies. Biostatistics 13, 762-775.

Sun, J., Zheng, Y., Hsu, L. (2013) A unified mixed-effects model for rare-variant association in sequencing studies. Genetic Epidemiology 37, 334-344.

Chen, H., Huffman, J.E., Brody, J.A., Wang, C., Lee, S., Li, Z., Gogarten, S.M., Sofer, T., Bielak, L.F., Bis, J.C., et al. (2019) Efficient variant set mixed model association tests for continuous and binary traits in large-scale whole-genome sequencing studies. The American Journal of Human Genetics 104, 260-274.

See Also

glmmkin, SMMAT.meta

Examples

```
if(requireNamespace("SeqArray", quietly = TRUE) && requireNamespace("SeqVarTools",
        quietly = TRUE)) {
data(example)
attach(example)
model0 <- glmmkin(disease ~ age + sex, data = pheno, kins = GRM, id = "id",</pre>
        family = binomial(link = "logit"))
geno.file <- system.file("extdata", "geno.gds", package = "GMMAT")</pre>
group.file <- system.file("extdata", "SetID.withweights.txt",</pre>
package = "GMMAT")
out <- SMMAT(model0, geno.file, group.file, MAF.range = c(0, 0.5),
        miss.cutoff = 1, method = "davies")
print(out)
}
## Not run:
obj <- SMMAT.prep(model0, geno.file, group.file)</pre>
save(obj, file = "SMMAT.prep.tmp.Rdata")
# quit R session
# open a new R session
obj <- get(load("SMMAT.prep.tmp.Rdata"))</pre>
out <- SMMAT.lowmem(obj, MAF.range = c(0, 0.5), miss.cutoff = 1,
        method = "davies")
print(out)
unlink("SMMAT.prep.tmp.Rdata")
## End(Not run)
```

SMMAT.meta

Meta-analysis for variant Set Mixed Model Association Tests (SM-MAT)

Description

Variant Set Mixed Model Association Tests (SMMAT-B, SMMAT-S, SMMAT-O and SMMAT-E) in the meta-analysis.

Usage

```
SMMAT.meta(meta.files.prefix, n.files = rep(1, length(meta.files.prefix)), cohort.group.idx = NULL, group.file, group.file.sep = "\t", MAF.range = c(1e-7, 0.5), MAF.weights.beta = c(1, 25), miss.cutoff = 1, method = "davies", tests = "E", rho = c(0, 0.1^2, 0.2^2, 0.3^2, 0.4^2, 0.5^2, 0.5, 1), use.minor.allele = FALSE, verbose = FALSE)
```

Arguments

meta.files.prefix

a character vector for prefix of intermediate files (.score.* and .var.*) required in a meta-analysis. Each element represents the prefix of .score.* and .var.* from one cohort. The length of vector should be equal to the number of cohorts.

n.files

an integer vector with the same length as meta.files.prefix, indicating how many sets of intermediate files (.score.* and .var.*) are expected from each cohort, usually as the result of multi-threading in creating the intermediate files (default = rep(1, length(meta.files.prefix))).

cohort.group.idx

a vector with the same length as meta.files.prefix, indicating which cohorts should be grouped together in the meta-analysis assuming homogeneous genetic effects. For example, c("a","b","a","a","b") means cohorts 1, 3, 4 are assumed to have homogeneous genetic effects, and cohorts 2, 5 are in another group with homogeneous genetic effects (but possibly heterogeneous with group "a"). If NULL, all cohorts are in the same group (default = NULL).

group.file

a plain text file with 6 columns defining the test units. There should be no headers in the file, and the columns are group name, chromosome, position, reference allele, alternative allele and weight, respectively.

group.file.sep the delimiter in group.file (default = "\t").

MAF.range

a numeric vector of length 2 defining the minimum and maximum minor allele frequencies of variants that should be included in the analysis (default = c(1e-7, 0.5)). Filter applied to the combined samples.

MAF.weights.beta

a numeric vector of length 2 defining the beta probability density function parameters on the minor allele frequencies. This internal minor allele frequency weight is multiplied by the external weight given by the group.file. To turn off internal minor allele frequency weight and only use the external weight given by the group.file, use c(1, 1) to assign flat weights (default = c(1, 25)). Applied to the combined samples.

miss.cutoff

the maximum missing rate allowed for a variant to be included (default = 1, including all variants). Filter applied to the combined samples.

method

a method to compute p-values for SKAT-type test statistics (default = "davies"). "davies" represents an exact method that computes a p-value by inverting the characteristic function of the mixture chisq distribution, with an accuracy of 1e-6. When "davies" p-value is less than 1e-5, it defaults to method "kuonen". "kuonen" represents a saddlepoint approximation method that computes the tail probabilities of the mixture chisq distribution. When "kuonen" fails to compute a p-value, it defaults to method "liu". "liu" is a moment-matching approximation method for the mixture chisq distribution.

tests

a character vector indicating which SMMAT tests should be performed ("B" for the burden test, "S" for SKAT, "O" for SKAT-O and "E" for the efficient hybrid test of the burden test and SKAT). The burden test and SKAT are automatically included when performing "O", and the burden test is automatically included when performing "E" (default = "E").

rho

a numeric vector defining the search grid used in SMMAT-O for SKAT-O (see the SKAT-O paper for details). Not used for SMMAT-B for the burden test, SMMAT-S for SKAT or SMMAT-E for the efficient hybrid test of the burden test and SKAT (default = $c(0, 0.1^2, 0.2^2, 0.3^2, 0.4^2, 0.5^2, 0.5, 1)$).

use.minor.allele

a logical switch for whether to use the minor allele (instead of the alt allele) as the coding allele (default = FALSE). It does not change SMMAT-S results, but SMMAT-B (as well as SMMAT-O and SMMAT-E) will be affected. Along with the MAF filter, this option is useful for combining rare mutations, assuming rare allele effects are in the same direction. Use with caution, as major/minor alleles may flip in different cohorts. In that case, minor allele will be determined based on the allele frequency in the combined samples.

verbose a logical switch for whether a progress bar should be shown (default = FALSE).

Value

a data frame with the following components:

group name of the test unit group.

n.variants number of variants in the test unit group that pass the missing rate and allele

frequency filters.

B. score burden test score statistic.

B. var variance of burden test score statistic.

B. pval burden test p-value.
S. pval SKAT p-value.
O. pval SKAT-O p-value.

0.minp minimum p-value in the SKAT-O search grid.

0.minp.rho rho value at the minimum p-value in the SKAT-O search grid.

E. pval SMMAT efficient hybrid test of the burden test and SKAT p-value.

Author(s)

Han Chen

References

Wu, M.C., Lee, S., Cai, T., Li, Y., Boehnke, M., Lin, X. (2011) Rare-variant association testing for sequencing data with the sequence kernel association test. The American Journal of Human Genetics 89, 82-93.

Lee, S., Wu, M.C., Lin, X. (2012) Optimal tests for rare variant effects in sequencing association studies. Biostatistics 13, 762-775.

Sun, J., Zheng, Y., Hsu, L. (2013) A unified mixed-effects model for rare-variant association in sequencing studies. Genetic Epidemiology 37, 334-344.

Chen, H., Huffman, J.E., Brody, J.A., Wang, C., Lee, S., Li, Z., Gogarten, S.M., Sofer, T., Bielak, L.F., Bis, J.C., et al. (2019) Efficient variant set mixed model association tests for continuous and binary traits in large-scale whole-genome sequencing studies. The American Journal of Human Genetics 104, 260-274.

See Also

```
glmmkin, SMMAT
```

Examples

```
if(requireNamespace("SeqArray", quietly = TRUE) && requireNamespace("SeqVarTools",
        quietly = TRUE)) {
data(example)
attach(example)
model0 \leftarrow glmmkin(disease \sim age + sex, data = pheno, kins = GRM, id = "id",
        family = binomial(link = "logit"))
geno.file <- system.file("extdata", "geno.gds", package = "GMMAT")</pre>
group.file <- system.file("extdata", "SetID.withweights.txt",</pre>
package = "GMMAT")
metafile <- tempfile()</pre>
out <- SMMAT(model0, geno.file, group.file, meta.file.prefix = metafile,</pre>
        MAF.range = c(0, 0.5), miss.cutoff = 1, method = "davies")
print(out)
out1 <- SMMAT.meta(metafile, group.file = group.file)</pre>
print(out1)
unlink(paste0(metafile, c(".score", ".var"), ".1"))
}
```

Index

```
* Wald test
    glmm.wald, 9
* generalized linear mixed model
    glmm.score, 4
    glmm.wald, 9
    glmmkin, 14
    GMMAT-package, 2
    SMMAT, 21
    SMMAT.meta, 25
* meta-analysis
    glmm.score.meta, 7
* package
    GMMAT-package, 2
* score test
    glmm.score, 4
    glmm.score.meta, 7
* variant set-based test
    SMMAT, 21
    SMMAT.meta, 25
as.data.frame, 9, 14
example, 4
family, 10, 15
formula, 9, 14, 16
glm, 12, 15
glmm.score, 4, 8, 13, 19, 20
glmm.score.meta, 7
glmm.wald, 7, 9
glmmkin, 7, 10, 13, 14, 24, 28
GMMAT (GMMAT-package), 2
GMMAT-package, 2
lm, 16
SMMAT, 19, 20, 21, 28
SMMAT.meta, 24, 25
```