Package 'StageWise'

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Title Two-stage analysis of multi-environment trials for genomic selection and GWAS

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Description Two-stage analysis of multi-environment trials for genomic selection and GWAS

Depends R (>= 4.0)

License GPL-3

LazyData true

RoxygenNote 7.1.1

Encoding UTF-8

Imports Matrix, ggplot2, methods, ggrepel, rlang, ggpubr, SpATS, spam

Suggests knitr, rmarkdown, asreml

Collate 'Stage1.R'
'Stage2.R'
'blup.R'
'blup_prep.R'
'class_geno.R'
'class_prep.R'
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'gwas_threshold.R'
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'summary.var.R'

R topics documented:

`blup` *BLUP*

Description

BLUP

Usage

```
blup(data, geno = NULL, what, index.weights = NULL, gwas.ncore = 0L)
```

Arguments

`data`	object of `class_prep` from `blup_prep`
`geno`	object of `class_geno` from `read_geno`
`what`	"id" or "marker"
`index.weights`	named vector of index weights for the locations or traits
`gwas.ncore`	Integer indicating number of cores to use for GWAS (default is 0 for no GWAS). Requires `what="markers"`.

Details

Argument `what="id"` leads to prediction of breeding values (BV) and genotypic values (GV), including the average fixed effect of the environments and any fixed effect markers. For `what="marker"`, environment fixed effects are not included in the BLUP. Argument `index.weights` is a named vector (matching the names of the locations or traits), and the values are interpreted for standardized traits. An overall scaling factor is also applied so that the sum of the squared index coefficients equals 1.

Value

Data frames of BLUPs

blup_prep *Prepare data for BLUP*

Description

Prepare data for BLUP

Usage

```
blup_prep(data, vcov = NULL, geno = NULL, vars, mask = NULL)
```

Arguments

data	data frame of BLUEs from Stage 1
vcov	list of variance-covariance matrices for the BLUEs
geno	object of class_geno from read_geno
vars	object of class_var from Stage2
mask	(optional) data frame with possible columns "id","env","trait"

Value

Object of class_prep

class_geno-class *S4 class for marker genotype data*

Description

S4 class for marker genotype data

Slots

map Marker map positions

coeff Coefficients of the marker effects (dim: indiv x marker)

scale Scaling factor between markers and indiv

G Additive (genomic) relationship matrix

eigen.G list of eigenvalues and eigenvectors

class_prep-class *S4 class to prepare for blup*

Description

S4 class to prepare for blup

Slots

`y` Stage 1 BLUEs

`Z` Incidence matrix for random effects

`id` genotype identifiers

`var.u` variance of random effects

`Vinv` inverted covariance matrix of the Stage 1 BLUEs

`Pmat` P matrix from Searle

`fixed` fixed effect estimates

`random` random effect estimates

`add` matrix of additive variances from `class_var`

`loc.env` data frame with loc, env

`trait.env` data frame with trait, env

`index.scale` sqrt of genetic variances for the locations/traits

`fixed.marker` names of fixed effect markers

class_var-class *S4 class for variances*

Description

S4 class for variances

Slots

`add` additive

`g.resid` genetic residual

`resid` residual

`meanG` mean of diagonal of G

`meanOmega` mean of diagonal of Omega

`fixed.marker.var` variance of marker fixed effects

`fixed.marker.cov` contribution of marker fixed effects to additive covariance between locations

gwas_threshold *Compute GWAS discovery threshold*

Description

Compute GWAS discovery threshold

Usage

```
gwas_threshold(geno, alpha = 0.05, exclude.chrom = NULL, n.core = 1)
```

Arguments

geno	object of class_geno
alpha	genome-wide significance level
exclude.chrom	chromosomes to exclude
n.core	number of cores to use

Details

Uses a Bonferroni-type correction based on an effective number of markers that accounts for LD (Moskvina and Schmidt, 2008).

Value

-log10(p) threshold

References

Moskvina V, Schmidt KM (2008) On multiple-testing correction in genome-wide association studies. Genetic Epidemiology 32:567-573. doi:10.1002/gepi.20331

manhattan_plot *Create Manhattan plot*

Description

Create Manhattan plot

Usage

```
manhattan_plot(data, chrom = NULL, thresh = NULL, rotate.label = FALSE)
```

Arguments

data	data frame with columns for marker, chrom, position, and gwas.score
chrom	optional, to plot only one chromosome
thresh	optional, to include horizontal line at discovery threshold
rotate.label	TRUE/FALSE whether to rotate x-axis labels to be perpendicular

Details

Assumes position in bp

Value

ggplot2 object

`predict` *Predict additive (breeding) values from marker effects*

Description

Predict additive (breeding) values from marker effects

Arguments

`object`	object of `class_geno`
`marker.effects`	data frame with columns "marker","add.effect"

Details

Use the `blup` function with `what="marker"` to generate the data frame for `marker.effects`.

Value

data frame with columns "id", "BV"

`quantile` *G matrix quantile*

Description

G matrix quantile

Arguments

`x`	object of `class_geno`
`prob`	probability

Details

Unlike the S3 method, `prob` must have length = 1

Value

data frame with the quantile of the G matrix coefficients for each id

read_geno *Read marker genotype data*

Description

Read marker genotype data

Usage

```
read_geno(filename, ploidy, map, eigen.tol = 1e-09, min.minor.allele = 5)
```

Arguments

filename	Name of CSV file
ploidy	2,4,6,etc. (even numbers)
map	TRUE/FALSE
eigen.tol	See Details. Default is 1e-9.
min.minor.allele	See Details. Default is 5.

Details

When map=TRUE, first three columns of the file are marker, chrom, position. When map=FALSE, the first column is marker. Subsequent columns contain the allele dosage for individuals/clones, coded 0,1,2,...ploidy (fractional values are allowed). The input file for diploids can also be coded using -1,0,1 (fractional values allowed). Additive coefficients are computed by subtracting the population mean from each marker, and the additive (genomic) relationship matrix is computed as G = tcrossprod(coeff)/scale. The scale parameter ensures the mean of the diagonal elements of G equals 1 under panmictic equilibrium. Missing genotype data is replaced with the population mean. For numerical conditioning, eigenvalues of G smaller than eigen.tol are replaced by eigen.tol.

The argument min.minor.allele specifies the minimum number of individuals that must contain the minor allele. Markers that do not meet this threshold are discarded.

Value

Variable of class class_geno.

Stage1 *Stage 1 analysis of multi-environment trials*

Description

Computes genotype BLUEs within each environment

Usage

```
Stage1(
  filename,
  traits,
  effects = NULL,
  solver = "asreml",
  spline = NULL,
  silent = TRUE,
  workspace = c("500mb", "500mb")
)
```

Arguments

`filename`	Name of CSV file
`traits`	trait names (see Details)
`effects`	data frame specifying other effects in the model (see Details)
`solver`	one of the following: "asreml","spats"
`spline`	vector of variable names for 2D spline with SpATS
`silent`	TRUE/FALSE, whether to suppress REML output
`workspace`	memory limits for ASRreml-R

Details

The input file must have one column labeled "id" for the individuals and one labeled "env" for the environments. The data for each environment are analyzed independently with a linear mixed model. Although not used in Stage1, to include a genotype x location effect in `Stage2`, a column labeled "loc" should be present in the input file.

Argument `effects` is used to specify other i.i.d. effects besides genotype and has three columns: name, fixed, factor. The "name" column is a string that must match a column in the input file. The fixed column is a logical variable to indicate whether the effect is fixed (TRUE) or random (FALSE). The factor column is a logical variable to indicate whether the effect is a factor (TRUE) or numeric (FALSE).

Argument `solver` specifies which software to use for REML. Current options are "asreml" and "spats". For "spats", the argument `spline` must be a vector of length two, with the names of the x and y variables (respectively) for the 2D spline.

The heritability and residuals in the output are based on a random effects model for id.

Missing response values are omitted for single-trait analysis but retained for multi-trait analysis (unless both traits are missing), to allow for prediction in Stage 2.

Argument `workspace` is a vector of length two containing the workspace and pworkspace limits for ASReml-R, with default values of 500mb. If you get an error about insufficient memory, try increasing the appropriate value (workspace for variance estimation and pworkspace for BLUE computation).

For multiple traits, only "asreml" is supported, and only the BLUE model is run, so the returned object does not contain H2 or resid.

Value

List containing

blue data frame of BLUEs for all environments

vcov list of variance-covariance matrices for the BLUEs, one per env

H2 broad-sense H2 on a plot basis

resid list containing diagnostic plots and data frame of residuals

aic AIC for the BLUE model (only with asreml)

`Stage2` *Stage 2 analysis of multi-environment trials*

Description

Stage 2 analysis of multi-environment trials

Usage

```
Stage2(
  data,
  vcov = NULL,
  geno = NULL,
  fix.eff.marker = NULL,
  silent = TRUE,
  workspace = "500mb"
)
```

Arguments

`data`	data frame of BLUEs from Stage 1 (see Details)
`vcov`	named list of variance-covariance matrices for the BLUEs
`geno`	output from `read_geno`
`fix.eff.marker`	markers in `geno` to include as additive fixed effect covariates
`silent`	TRUE/FALSE, whether to suppress ASReml-R output
`workspace`	Memory limit for ASRreml-R variance estimation

Details

Stage 2 of the two-stage approach described by Damesa et al. 2017, using ASReml-R for variance component estimation. The variable `data` has three mandatory column: id, env, BLUE. Optionally, `data` can have a column labeled "loc", which changes the main effect for genotype into a separable genotype-within-location effect, using a FA2 covariance model for the locations. Optionally, `data` can have a column labeled "trait", which uses an unstructured covariance model. The multi-location and multi-trait analyses cannot be combined. Missing data are allowed in the multi-trait but not the single-trait analysis. The argument `geno` is used to partition genetic values into additive and non-additive (g.resid) components. Any individuals in `data` that are not present in `geno` are discarded.

The argument `vcov` is used to partition the macro- and micro-environmental variation, which are called GxE and residual in the output. `vcov` is a named list of variance-covariance matrices for the BLUEs within each environment, with id for rownames (single trait) or id:trait. The order in `vcov` and `data` should match. Both `data` and `vcov` can be created using the function `Stage1`.

Because ASReml-R can only use relationship matrices defined in the global environment, this function creates and then removes global variables when either `vcov` or `geno` is used. By default, the workspace memory for ASReml-R is set at 500mb. If you get an error about insufficient memory, try increasing it. ASReml-R version 4.1.0.148 or later is required.

Value

List containing

aic AIC

vars variances, as variable of class `class_var`

fixed Fixed effect estimates for env and markers

random Random effect predictions

uniplot uniplot of the genetic correlation between locations

References

Damesa et al. 2017. Agronomy Journal 109: 845-857. doi:10.2134/agronj2016.07.0395

`summary` *Summarize variances and correlations*

Description

Summarize variances and correlations

Arguments

`object`	object of `class_var`

Details

When reporting the partitioning of variance, the variance component for the additive effect is multiplied by the mean diagonal of the G matrix.

Value

For univariate analysis, a matrix with the proportion of variance (R2). For multi-location or multi-trait analysis, a correlation matrix is also returned. If a G matrix was included in the analysis, the correlation is between additive values; otherwise, it is between genotypic values. If fixed effect markers are present, the additive correlation is above the diagonal, and below the diagonal is the proportion of additive covariance due to the fixed effect markers.

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