# Package 'StageWise'

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Title Two-stage analysis of multi-environment trials for genomic selection
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<b>Description</b> Two-stage analysis of multi-environment trials for genomic selection
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R topics documented:
blup blup_prep class_geno-class class_prep-class class_var-class gwas_threshold manhattan_plot quantile

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blup BLUP

## **Description**

**BLUP** 

## Usage

```
blup(data, geno = NULL, what, index.coeff = 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.coeff index coefficients 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 argument what="marker", fixed effects are not included in the BLUP. Argument index.coeff should be a named vector, matching the names of the locations or traits. Index coefficients are assumed to imply relative weights for the different locations or traits; as such, they are divided by the square root of the genetic variance estimate for that location/trait and then rescaled to have unit sum.

#### Value

Data frames of BLUPs

blup\_prep 3

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 column "id" and optional columns "env", "trait"

#### **Details**

The argument mask can be used to mask all observations for a particular individual, or by including a column named 'env', only the observations in particular environments can be masked. This is useful for cross-validation to test the accuracy of predicting into new environments.

## Value

```
Object of class_prep
```

## 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
```

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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 logical whether additive effects predicted

loc.env data frame with loc, 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

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gwas_threshold	Compute GWAS discovery threshold
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## **Description**

Compute GWAS discovery threshold

## Usage

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

## **Arguments**

geno object of class\_geno
n.core number of cores to use

alpha genome-wide significance level

#### **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

## Description

Create Manhattan plot

## Usage

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

## Arguments

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

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#### **Details**

Assumes position in bp

#### Value

ggplot2 object

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)
```

## **Arguments**

filename Name of CSV file

ploidy 2,4,6,etc. (even numbers)

map TRUE/FALSE

eigen.tol 1e-9

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#### **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.

#### 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 = "asrem1",
  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

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#### **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

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"
)
```

Stage2

#### **Arguments**

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

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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 data frame of variances. For multi-location or multi-trait analysis, a list containing the variance data frame and the correlation matrix.

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