Package 'polyBreedR'

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Title Using genome-wide markers for polyploid breeding

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Description Using genome-wide markers for polyploid breeding
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R topics documented:
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A_mat

Additive relationship matrix from pedigree

Description

Additive relationship matrix from pedigree

Usage

```
A_mat(ped, ploidy, order.ped = TRUE)
```

Arguments

ped Pedigree in three column format: id, mother, father

ploidy 2 or 4

order.ped TRUE/FALSE does the pedigree need to be ordered so that progeny follow par-

ents

Details

This is a wrapper that prepares the pedigree in the format required for R package AGHmatrix by Amadeu et al. (2016) (cite them if you use this function). A random bivalents model for tetraploid meiosis is assumed.

Value

Additive relationship matrix (dim: indiv x indiv)

References

Amadeu et al. (2016) Plant Genome 9, doi:10.3835/plantgenome2016.01.0009

check_ploidy

Check ploidy

Description

Fraction of simplex or triplex markers

Usage

```
check_ploidy(geno, map)
```

Arguments

geno Genotype matrix (markers x indiv)

map Data frame with marker map (Marker, Chrom, Position)

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Details

For every indiv in the genotype matrix, the fraction of markers per chromosome called as simplex or triplex is calculated, which should be low for diploids. A small amount of missing genotype data can be tolerated.

Value

List containing

mat Matrix (indiv x chrom) of resultsplot ggplot2 barplot

check_trio

Check markers for parent-offspring trio

Description

Check markers for parent-offspring trio

Usage

```
check_trio(parentage, geno, ploidy)
```

Arguments

parentage Data frame with three columns: id, mother, father

geno Matrix of allele dosages: markers x indiv

ploidy 2 or 4

Details

Computes the percentage of markers at which the two parents and offspring have incompatible allele dosages (for tetraploids, the random bivalents model is used). For dihaploid offspring of a single tetraploid parent, use ploidy = 4 and "haploid" for the father in parentage, as well as a diploid (0,1,2) genotype for the offspring. A small amount of missing genotype data can be tolerated.

Value

Data frame with the percentage of incompatible markers for each trio

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 $\operatorname{direct_sum}$

Direct Sum

Description

Direct Sum

Usage

```
direct_sum(x)
```

Arguments

Х

list of matrices

Value

Sparse Matrix

D_mat

Dominance genomic relationships

Description

Coefficients and relationship matrix for digenic dominance effects with bi-allelic markers

Usage

```
D_mat(geno, ploidy)
```

Arguments

geno

Matrix of allele dosages: markers x indiv

ploidy

2 or 4

Details

Digenic dominance effects are based on the traditional orthogonal decomposition of genetic variance in panmictic populations (Fisher 1918; Kempthorne 1957; Endelman et al. 2018). The D matrix is computed from the coefficients and scaling factor according to D = tcrossprod(coeff/scale). Missing genotype data is replaced with the population mean.

Value

List containing

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

scale Scaling factor between markers and indiv

mat D matrix

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References

```
Fisher (1918) Trans. Roy. Soc. Edin. 52:399-433.
Kempthorne (1957) An Introduction to Genetic Statistics.
Endelman et al. (2018) Genetics 209:77-87.
```

geno_call

Genotype calls

Description

Genotype calls based on a normal mixture model

Usage

```
geno_call(
  data,
  filename,
  model.ploidy = 4,
  sample.ploidy = 4,
  min.posterior = 0,
  transform = TRUE
)
```

Arguments

data matrix (markers x id) of input values for the normal mixture model

filename CSV filename with the model parameters

model.ploidy 2 or 4 (default) sample.ploidy 2 or 4 (default)

min.posterior minimum posterior probability (default 0) for genotype call

transform TRUE (default) or FALSE whether to apply arcsin square root transformation

Details

The first column of the CSV input file should be the SNP ID, followed by columns for the normal distribution means, standard deviations, and mixture probabilities. Genotype calls are based on the maximum a posteriori (MAP) method. If the posterior probability of the MAP genotype is less than min.posterior, then NA is returned for that sample. By default, an arcsin square root transformation is applied to the input values to match the approach used by R package fitPoly. To use a tetraploid mixture model for diploid samples, set sample.ploidy = 2 and model.ploidy = 4.

Value

matrix of allele dosages (0,1,2,...ploidy) with dimensions markers x individuals

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get_pedigree

Generate pedigree

Description

Generate pedigree for a set of individuals

Usage

```
get_pedigree(id, pedfile, delim = ",", na.string = "NA", trim = TRUE)
```

Arguments

id Vector of names of individuals

pedfile Name of pedigree file

delim Delimiter for the pedigree file (default is "," for CSV)

na.string String used for NA in the pedigree file (default is "NA")

trim TRUE/FALSE whether to trim pedigree (see Details)

Details

Finds ancestors of individuals in a three-column pedigree file (id,mother,father). The id column can be the identifier for an individual or cross. String matches must be exact or based on the naming convention crossID-progenyID. The returned pedigree is ordered using R package pedigree so that offspring follow parents. When trim is TRUE (default), the pedigree is trimmed to remove ancestors with only one offspring (which are not needed to compute the pedigree relationship matrix).

Value

Data frame with columns id, mother, father

GvsA

Plot G vs. A

Description

Plot marker-based vs. pedigree-based additive relationship coefficients

Usage

```
GvsA(
  parentage,
  G,
  A,
  filename = NULL,
  thresh.G = Inf,
  thresh.A = 0.5,
  Gmax = NULL,
  Amax = NULL
)
```

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Arguments

parentage Data frame of individuals to plot, with 3 columns: id, mother, father

G Genomic relationship matrix
A Pedigree relationship matrix

filename Name of PDF file to save the results (optional for one individual)

thresh.G Threshold above which names are displayed (default Inf)
thresh.A Threshold above which names are displayed (default 0.5)

Gmax Upper limit for y-axis for plotting. If NULL, maximum value in G is used.

Amax Upper limit for x-axis for plotting. If NULL, maximum value in A is used.

Details

Useful for finding and correcting pedigree errors. If the G or A coefficient for an individual exceeds the threshold, its name is displayed in the figure. If parentage contains one individual, by default a ggplot2 variable will be returned, but the result can also be written to file. If multiple individuals are present, a filename is required.

G_mat Additive genomic relationships

Description

Coefficients and relationship matrix for additive effects with bi-allelic markers

Usage

```
G_mat(geno, ploidy)
```

Arguments

geno Matrix of allele dosages: markers x indiv

ploidy 2 or 4

Details

Additive effects are based on the traditional orthogonal decomposition of genetic variance in panmictic populations (Fisher 1918; Kempthorne 1957; Endelman et al. 2018). The G matrix is computed from the coefficients and scaling factor according to G = tcrossprod(coeff/scale). Missing genotype data is replaced with the population mean.

Value

List containing

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

scale Scaling factor between markers and indiv

mat G matrix

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References

```
Fisher (1918) Trans. Roy. Soc. Edin. 52:399-433.
Kempthorne (1957) An Introduction to Genetic Statistics.
Endelman et al. (2018) Genetics 209:77-87.
```

impute

Impute missing marker data

Description

Impute marker data based on the population mean or mode

Usage

```
impute(geno, method)
```

Arguments

geno Matrix of allele dosages with dimensions markers x indiv

method Either "mean" or "mode"

Details

Missing values are imputed with either the population mean or mode (most frequent value) for each marker

Value

Imputed genotype matrix (markers x indiv)

merge_impute

Merge two genotype matrices and impute missing data

Description

Merge two genotype matrices and impute missing data by BLUP

Usage

```
merge_impute(geno1, geno2, ploidy)
```

Arguments

geno1	Genotype matrix (coded 0ploidy) with dimensions markers x indiv
geno2	Genotype matrix (coded 0ploidy) with dimensions markers x indiv
ploidy	Either 2 or 4

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Details

Designed to impute from low to high density markers. The BLUP method is equivalent to Eq. 4 of Poland et al. (2012), but this function is not iterative. Additional shrinkage toward the mean is applied if needed to keep the imputed values within the range [0,ploidy]. Missing data in the input matrices are imputed with the population mean for each marker. If an individual appears in both input matrices, it is renamed with suffixes ".1" and ".2" and treated as two different individuals. Monomorphic markers are removed.

Value

Imputed genotype matrix (markers x indiv)

References

Poland et al. (2012) Plant Genome 5:103-113.

readXY

Read SNP array intensity data

Description

Read SNP array intensity data

Usage

```
readXY(filename, skip, output = "ratio")
```

Arguments

filename filename

skip number of lines to skip before the header line with the column names

output Either "ratio" or "theta"

Details

The first two columns of the tab-delimited input file should be the SNP and Sample ID. Columns labeled "X" and "Y" contain the signal intensities for the two alleles. Use output to specify whether to return the ratio = Y/(X+Y) or theta = atan(Y/X)*2/pi.

Value

matrix with dimensions markers x individuals

10 Step2

Step2	Genomic prediction of marker effects from multi-environment trials

Description

Additive and dominance marker effects predicted by BLUP, based on variance components estimated with ASReml-R (must have license)

Usage

```
Step2(pheno, G, D = NULL, AIC = TRUE, silent = FALSE, workspace = "128mb")
```

Arguments

pheno	List of matrices containing BLUEs and variance-covariance matrix
G	Returned object from G_mat for additive effects
D	Optional, returned object from D_mat for dominance effects (default is NULL)
AIC	Boolean variable, whether to select A vs. AD model based on AIC (default is TRUE)
silent	Boolean variable, whether to suppress ASReml-R convergence monitoring (default is FALSE)
workspace	Workspace memory for ASReml-R (default is "128mb")

Details

Best practice for the analysis of datasets comprising multiple experiments follows a two-step approach (Damesa et al. 2017). Step 1 generates a vector of genotype BLUEs (and variancecovariance matrix) for each experiment, taking its design and potentially spatial factors into account. In Step 2, the BLUEs are used as a response variable in a mixed model, and the covariance matrix for the residuals is constrained to equal the direct sum of the covariance matrices of the BLUEs. In the current implementation, the Step 2 model contains three genetic effects: additive, digenic dominant, and independent (to capture higher-order non-additive effects). The digenic dominant term has nonzero mean to allow for heterosis/inbreeding depression (Varona et al. 2018). The model also contains a fixed effect for each experiment and independent random effects for the genotype x environment interaction (an environment may contain multiple experiments). Each element of pheno corresponds to an experiment and is a list with components named "env" and "y". Env is the name of the environment for that experiment, and "y" is a matrix of dimensions n x (n+1), where n is the number of genotypes in the trial. The first column of the matrix contains the BLUEs, and the remaining n columns are the variance-covariance matrix for the BLUEs. ASReml-R is used to estimate variance components, and BLUPs are computed based on standard formulas (Searle et al. 1992). When D is not NULL, variances are first estimated for the strictly additive (A) model and used as initial values when fitting the additive + dominance (AD) model. If AIC = TRUE, then whichever model (A vs. AD) has lower AIC is used for BLUP; otherwise, BLUPs are computed based on the AD model regardless of the AIC values. Individuals in pheno but not in G are removed from the analysis; to make predictions for unphenotyped individuals, include them in G (and D).

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Value

List containing

params Matrix with AIC and estimated parametersindiv Matrix of predicted values for the individuals in genomarkers Matrix of predicted effects for the markers in geno

References

Damesa et al. 2017. Agronomy Journal 109: 845-857. doi:10.2134/agronj2016.07.0395 Varona et al. 2018. Frontiers in Genetics 9: 78. doi:10.3389/fgene.2018.00078 Searle et al. 1992. Variance Components. doi:10.1002/9780470316856

update_alias

Update names based on alias

Description

Update names based on data frame with alias and preferred name

Usage

```
update_alias(x, alias, remove.space = TRUE)
```

Arguments

x Vector of names to update

alias Data frame with two columns: first is the preferred name and second is the alias

remove.space TRUE/FALSE

Details

Parameter remove. space indicates whether blank spaces should be removed before string matching

Value

Vector with updated names

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