# Package 'ASRgenomics' 

January 29, 2024
Title Complementary Genomic Functions
Version 1.1.4
Description Presents a series of molecular and genetic routines in the R environment with the aim of assisting in analytical pipelines before and after the use of 'asreml' or another library to perform analyses such as Genomic Selection or Genome-Wide Association Analyses. Methods and examples are described in Gezan, Oliveira, Galli, and Murray (2022) <https:
//asreml.kb.vsni.co.uk/wp-content/uploads/sites/3/ASRgenomics_Manual.pdf>.
License MIT + file LICENSE
Depends R (>=4.0.0)
Imports AGHmatrix, cowplot, crayon, data.table, ellipse, factoextra, ggplot2, Matrix, methods, scattermore, stats, superheat, utils

Suggests knitr, rmarkdown, spelling, V8, testthat (>= 3.0.0)
Config/testthat/edition 3
Encoding UTF-8
LazyData true
LazyDataCompression bzip2
RoxygenNote 7.3.1
Language en-US
NeedsCompilation no
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Repository CRAN
Date/Publication 2024-01-29 21:20:02 UTC

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full2sparse Generates a sparse form matrix from a full form matrix

## Description

Modifies the input square matrix into its sparse form with three columns per line, corresponding to the set: Row, Col, Value. This matrix defines the lower triangle row-wise of the original matrix and it is sorted as columns within row. Values of zero on the matrix are dropped by default. Individual names should be assigned to rownames and colnames.

## Usage

full2sparse (K = NULL, drop.zero = TRUE)

## Arguments

$\mathrm{K} \quad$ A square matrix in full form $(n \times n)($ default $=$ NULL $)$.
drop.zero If TRUE observations equal to zero are dropped (default $=$ TRUE $)$.

## Details

Based on the function published by Borgognone et al. (2016).

## Value

A matrix in sparse form with columns: Row, Col, Value together with the attributes rowNames and colNames. If attribute INVERSE is found this is also passed to the sparse matrix.

## References

Borgognone, M.G., Butler, D.G., Ogbonnaya, F.C and Dreccer, M.F. 2016. Molecular marker information in the analysis of multi-environment trial helps differentiate superior genotypes from promising parents. Crop Science 56:2612-2628.

## Examples

```
# Get G matrix.
G <- G.matrix(M = geno.apple, method = "VanRaden", sparseform = FALSE)$G
G[1:5, 1:5]
# Transform matrix into sparse.
G.sparse <- full2sparse(K = G)
head(G.sparse)
head(attr(G.sparse, "rowNames"))
head(attr(G.sparse, "colNames"))
```

G.inverse $\quad$ Obtains the inverse of the genomic relationship matrix $G$

## Description

Generates the inverse of a genomic relationship matrix $G$ that is provided. This input matrix should be of the full form $(n \times n)$ with individual names assigned to rownames and colnames. Several checks for the stability of the matrix are presented based on the reciprocal conditional number.

In case of an ill-conditioned matrix, options of blending, bending or aligning before inverting are available. These options will be deprecated (discontinued) in future versions of ASRgenomics as they can be better implemented in the function G. tuneup().
Based on procedures published by Nazarian and Gezan et al. (2016).

## Usage

```
G.inverse(
    \(\mathrm{G}=\mathrm{NULL}\),
    A = NULL,
    rcn.thr \(=1 \mathrm{e}-12\),
    blend = FALSE,
    pblend = 0.02,
    bend = FALSE,
    eig.tol = NULL,
    align = FALSE,
    digits = 8,
    sparseform = FALSE,
    message = TRUE
)
```


## Arguments

G

A Input of the pedigree relationship matrix $\boldsymbol{A}$ to perform blending or aligning in full form. It should be of the same dimension as the $\boldsymbol{G}$ matrix ( $n \times n$ ) (default $=$ NULL) (to be deprecated).
rcn.thr A threshold for identifying the $G$ matrix as an ill-conditioned matrix. Based on the reciprocal conditional number $($ default $=1 \mathrm{e}-12)$.
blend If TRUE a "blending" with identity matrix $\boldsymbol{I}$ or pedigree relationship matrix $\boldsymbol{A}$ (if provided) is performed (default $=$ FALSE) (to be deprecated).
pblend If blending is requested this is the proportion of the identity matrix $\boldsymbol{I}$ or pedigree relationship matrix $\boldsymbol{A}$ to blend for $($ default $=0.02)$ (to be deprecated).
bend If TRUE a "bending" is performed by making the matrix near positive definite (default $=$ FALSE) (to be deprecated).
eig.tol Defines relative positiveness (i.e., non-zero) of eigenvalues compared to the largest one. It determines which threshold of eigenvalues will be treated as zero $($ default $=$ NULL) (to be deprecated).
align If TRUE the genomic relationship matrix $G$ is aligned to the pedigree relationship matrix $\boldsymbol{A}$ (default = FALSE) (to be deprecated).
digits $\quad$ Set up the number of digits in used to round the output matrix (default $=8$ ).
sparseform If TRUE it generates an inverse matrix in sparse form to be used directly in asreml with required attributes (default $=$ FALSE).
message If TRUE diagnostic messages are printed on screen (default $=$ TRUE $)$.

## Value

A list with three of the following elements:

- Ginv: the inverse of $\boldsymbol{G}$ matrix in full form (only if sparseform = FALSE).
- Ginv. sparse: the inverse of $\boldsymbol{G}$ matrix in sparse form (only if sparseform = TRUE).
- status: the status (ill-conditioned or well-conditioned) of the inverse of $\boldsymbol{G}$ matrix.
- ren: the reciprocal conditional number of the inverse of $G$ matrix.


## References

Nazarian A., Gezan S.A. 2016. GenoMatrix: A software package for pedigree-based and genomic prediction analyses on complex traits. Journal of Heredity 107:372-379.

## Examples

```
# Example: An ill-conditioned matrix.
# Get G matrix.
G <- G.matrix(M = geno.apple, method = "VanRaden")$G
G[1:5, 1:5]
# Get the inverse of G.
GINV <- G.inverse(G = G, bend = FALSE, blend = FALSE, align = FALSE)
GINV$Ginv[1:5, 1:5]
GINV$status
```

```
G.matrix
```

Obtains the genomic matrix from SNP data for additive or dominant relationships

## Description

Generates the genomic numerator relationship matrix for additive (VanRaden or Yang) or dominant (Su or Vitezica) relationships. Matrix provided $\boldsymbol{M}$ is of the form $n \times p$, with $n$ individuals and $p$ markers. Individual and marker names are assigned to rownames and colnames, respectively. SNP data is coded as $0,1,2$ (integers or decimal numbers). Missing values, if present, need to be specified.

## Usage

G.matrix(
$M=N U L L$,
method = "VanRaden",
na.string = "NA",
sparseform = FALSE,
digits = 8,
message = TRUE
)

## Arguments

| M | A matrix with SNP data of form $n \times p$, with $n$ individuals and $p$ markers. Indi- <br> vidual and marker names are assigned to rownames and colnames, respectively. <br> SNP data is coded as 0, 1, 2 (integers or decimal numbers) (default $=$ NULL). |
| :--- | :--- |
| method | The method considered for calculation of genomic matrix. Options are: "VanRaden" <br> and "Yang" for additive matrix, and "Su" and "Vitezica" for dominant matrix <br> (default = "VanRaden"). |
| na.string | A character that is interpreted as missing values (default = "NA"). <br> sparseform <br> If TRUE it generates a matrix in sparse form to be used directly in asreml with <br> required attributes (default = FALSE). |
| digits | Set up the number of digits used to round the output matrix (default = 8). <br> message |
| If TRUE diagnostic messages are printed on screen (default = TRUE). |  |

## Details

Note: If data is provided with missing values, it will process calculations of relationships on pairwise non-missing data.
It uses function Gmatrix for calculations from R package AGHmatrix (Amadeu et al. 2019).

## Value

A list with one of these two elements:

- G: the $G$ matrix in full form (only if sparseform = FALSE).
- G. sparse: the $G$ matrix in sparse form (only if sparseform = TRUE).


## References

Amadeu, R.R., Cellon, C., Olmstead, J.W., Garcia, A.A.F, Resende, M.F.R. and P.R. Munoz. 2016. AGHmatrix: R package to construct relationship matrices for autotetraploid and diploid species: A blueberry example. The Plant Genome 9(3). doi: 10.3835/plantgenome2016.01.0009

## Examples

```
# Example: Requesting a full matrix by VanRanden.
# Get G matrix.
G <- G.matrix(M = geno.apple, method = "VanRaden")$G
G[1:5, 1:5]
# Example: Requesting a sparse form by VanRanden.
# Get G matrix.
G <- G.matrix(M = geno.apple, method = "VanRaden", sparseform = TRUE)$G.sparse
head(G)
head(attr(G, "rowNames"))
```

G.predict Generates the conditional predictions of random effects (BLUPs)

## Description

Predicts random effects values for individuals with unobserved responses (here called x , a vector of length $n x$ ) based on known random effect values for individuals with observed responses (here called y , a vector of length $n y$ ). This is done using the common genomic relationship matrix $\boldsymbol{G}$ for all individuals (full matrix of dimension $(n x+n y) \times(n x+n y)$ ).
The prediction of unobserved responses will be performed through the multivariante Normal conditional distribution. These predictions are identical to what would be obtained if the entire set of individuals $(n x+n y)$ were included into a GBLUP animal model fit with individuals in the set x coded as missing.
The user needs to provide the matrix $G$ in full form. Individual names ( $n x+n y$ ) should be assigned to rownames and colnames, and these can be in any order. If the variance-covariance matrix of the set y is provided, standard errors of random effects in set x are calculated.

## Usage

G.predict(G = NULL, gy = NULL, vcov.gy = NULL)

## Arguments

G
Input of the genomic relationship matrix $\boldsymbol{G}$ in full form (of dimension $(n x+$ $n y) \times(n x+n y))($ default $=$ NULL $)$.
gy Input of random effects (e.g. breeding values) for individuals with known values. Individual names should be assigned to rownames of this vector and be found on the matrix $\boldsymbol{G}($ default $=$ NULL $)$.
vcov.gy The variance-covariance matrix associated with the random effects from the individuals with known values (set y , of dimension $n y \times n y)($ default $=$ NULL $)$.

## Value

A data frame with the predicted random effect values for individuals with unobserved responses in the set $x$. If the variance-covariance matrix is provided, standard errors are included in an additional column.

## Examples

```
## Not run:
library(asreml) # Load asreml.
# Example: Apple data creating 100 missing observations.
# Prepare G (nx + ny).
G <- G.matrix(M = geno.apple, method = "VanRaden", sparseform = FALSE)$G
dim(G)
```

```
# Prepare subset of data.
# Select only 147 from 247 individuals from pheno.apple and geno.apple.
Gy <- G[1:147, 1:147]
phenoy <- pheno.apple[1:147, ]
# Obtain the BLUPs for the 147 individuals using ASReml-R.
# Blend Gy.
Gyb <- G.tuneup(G = Gy, blend = TRUE, pblend = 0.02)$Gb
# Get the Gy inverse
Gyinv <- G.inverse(G = Gyb, sparseform = TRUE)$Ginv.sparse
# Fit a GBLUP model
phenoy$INDIV <- as.factor(phenoy$INDIV)
modelGBLUP <-
    asreml(
        fixed = JUI_MOT ~ 1,
        random = ~vm(INDIV, Gyinv),
    workspace = 128e06,
    data = phenoy)
# Obtain Predictions - BLUP (set y).
BLUP <- summary(modelGBLUP,coef = TRUE)$coef.random
head(BLUP)
gy <- as.matrix(BLUP[, 1])
rownames(gy) <- phenoy$INDIV
# Ready to make conditional predictions.
g.cond <- G.predict(G = G, gy = gy)
head(g.cond)
## End(Not run)
```

G. tuneup $\quad$ Tune-up the the genomic relationship matrix $G$

## Description

Generates a new matrix that can be blended, bended or aligned in order to make it stable for future use or matrix inversion. The input matrix should be of the full form $(n \times n)$ with individual names assigned to rownames and colnames.
This routine provides three options of tune-up:

- Blend. The $\boldsymbol{G}$ matrix is blended (or averaged) with another matrix. $\boldsymbol{G}^{*}=(1-p) \boldsymbol{G}+p \boldsymbol{A}$, where $G^{*}$ is the blended matrix, $\boldsymbol{G}$ is the original matrix, $p$ is the proportion of the identity matrix or pedigree-based relationship matrix to consider, and $\boldsymbol{A}$ is the matrix to blend. Ideally, the pedigree-based relationship matrix should be used, but if this is not available (or it is of poor quality), then it is replaced by an identity matrix $\boldsymbol{I}$.
- Bend. It consists on adjusting the original $G$ matrix to obtain a near positive definite matrix, which is done by making all negative or very small eigenvalues slightly positive.
- Align. The original $\boldsymbol{G}$ matrix is aligned to the matching pedigree relationship matrix $\boldsymbol{A}$ where an $\alpha$ and $\beta$ parameters are obtained. More information can be found in the manual or in Christensen et al. (2012).

The user should provide the matrices $\boldsymbol{G}$ and $\boldsymbol{A}$ in full form $(n \times n)$ and matching individual names should be assigned to the rownames and colnames of the matrices.
Based on procedures published by Nazarian and Gezan et al. (2016).

## Usage

```
G.tuneup(
    G = NULL,
    A = NULL,
    blend = FALSE,
    pblend = 0.02,
    bend = FALSE,
    eig.tol = 1e-06,
    align = FALSE,
    rcn = TRUE,
    digits = 8,
    sparseform = FALSE,
    determinant = TRUE,
    message = TRUE
)
```


## Arguments

G
A
blend If TRUE a blending with identity matrix $\boldsymbol{I}$ or pedigree relationship matrix $\boldsymbol{A}$ (if provided $)$ is performed $($ default $=$ FALSE $)$.
pblend If blending is requested this is the proportion of the identity matrix $I$ or pedigreebased relationship matrix $\boldsymbol{A}$ to blend for (default $=0.02$ ).
bend If TRUE a bending is performed by making the matrix near positive definite (default $=$ FALSE).
eig.tol Defines relative positiveness (i.e., non-zero) of eigenvalues compared to the largest one. It determines which threshold of eigenvalues will be treated as zero (default $=1 \mathrm{e}-06$ ).
align If TRUE the genomic matrix $G$ is aligned to the matching pedigree relationship matrix $\boldsymbol{A}$ (default = FALSE).
ren If TRUE the reciprocal conditional number of the original and the bended, blended or aligned matrix will be calculated $($ default $=T R U E)$.
digits $\quad$ Set up the number of digits used to round the output matrix $($ default $=8)$.

| sparseform | If TRUE it generates an inverse matrix in sparse form to be used directly in asreml <br> with required attributes (default = FALSE). |
| :--- | :--- |
| determinant | If TRUE the determinant will be calculated, otherwise, this is obtained for matri- <br> ces of a dimension of less than $1,500 \times 1,500$ (default $=$ TRUE). |
| message | If TRUE diagnostic messages are printed on screen (default = TRUE). |

## Value

A list with six of the following elements:

- Gb: the inverse of $\boldsymbol{G}$ matrix in full form (only if sparseform = FALSE).
- Gb. sparse: if requested, the inverse of $G$ matrix in sparse form (only if sparseform $=$ TRUE).
- rcn0: the reciprocal conditional number of the original matrix. Values near zero are associated with an ill-conditioned matrix.
- rcnb: the reciprocal conditional number of the blended, bended or aligned matrix. Values near zero are associated with an ill-conditioned matrix.
- $\operatorname{det} 0$ : if requested, the determinant of the original matrix.
- blend: if the matrix was blended.
- bend: if the matrix was bended.
- align: if the matrix was aligned.


## References

Christensen, O.F., Madsen, P., Nielsen, B., Ostersen, T. and Su, G. 2012. Single-step methods for genomic evaluation in pigs. Animal 6:1565-1571. doi:10.1017/S1751731112000742.
Nazarian A., Gezan S.A. 2016. GenoMatrix: A software package for pedigree-based and genomic prediction analyses on complex traits. Journal of Heredity 107:372-379.

## Examples

\# Example: Apple dataset.
\# Get $G$ matrix.
G <- G.matrix ( $M=$ geno.apple, method = "VanRaden") \$G
G[1:5, 1:5]
\# Blend G matrix.
G_blended <- G.tuneup $(G=G$, blend $=$ TRUE, pblend $=0.05$ )
G_blended\$Gb[1:5, 1:5]
\# Bend G matrix.
G_bended <- G.tuneup ( $G=G$, bend $=$ TRUE, eig.tol $=1 e-03$ )
G_bended\$Gb[1:5, 1:5]
\# Example: Loblolly Pine dataset with pedigree - Aligned G matrix.

A <- AGHmatrix::Amatrix(ped.pine)
geno.apple

```
dim(A)
# Read and filter genotypic data.
M.clean <- qc.filtering(
    M = geno.pine655,
    maf = 0.05,
    marker.callrate = 0.2, ind.callrate = 0.20,
    na.string = "-9")$M.clean
# Get G matrix.
G <- G.matrix(M = M.clean, method = "VanRaden", na.string = "-9")$G
G[1:5, 1:5]
dim(G)
# Match G and A.
Aclean <- match.G2A(A = A, G = G, clean = TRUE, ord = TRUE, mism = TRUE)$Aclean
# Align G with A.
G_align <- G.tuneup(G = G, A = Aclean, align = TRUE)
G_align$Gb[1:5, 1:5]
```

    geno.apple Genotypic data for apple dataset
    
## Description

Genotypic data on 247 apple clones (i.e., genotypes) with a total of 2,828 SNP markers (coded as $0,1,2$ and there are no missing records). Dataset obtained from supplementary material in Kumar et al. (2015).

## Usage

geno.apple

## Format

matrix

## References

Kumar S., Molloy C., Muñoz P., Daetwyler H., Chagné D., and Volz R. 2015. Genome-enabled estimates of additive and nonadditive genetic variances and prediction of apple phenotypes across environments. G3 Genes, Genomes, Genetics 5:2711-2718.

## Examples

geno.apple[1:5, 1:5]

## Description

Genotypic data for a total of 4,853 SNPs (coded as $0,1,2$ and -9 for missing) on 655 genotypes of Loblolly Pine (Pinus taeda L.). Dataset modified from supplementary material from Resende et al. (2012). This dataset differs from the original as some genotypes were made artificially missing by full-sib family.

## Usage

geno.pine655

## Format

matrix

## References

Resende, M.F.R., Munoz, P. Resende, M.D.V., Garrick, D.J., Fernando, R.L., Davis, J.M., Jokela, E.J., Martin, T.A., Peter, G.F., and Kirst, M. 2012. Accuracy of genomic selection methods in a standard data set of loblolly pine (Pinus taeda L.). Genetics 190:1503-1510.

## Examples

```
geno.pine655[1:5, 1:5]
```

geno. pine926 Genotypic data of 926 genotypes for loblolly pine dataset

## Description

Genotypic data for a total of 4,853 SNPs (coded as $0,1,2$ and -9 for missing) on 926 genotypes of Loblolly Pine (Pinus taeda L.). Dataset obtained from supplementary material in Resende et al. (2012).

## Usage

geno.pine926

## Format

matrix

## References

Resende, M.F.R., Munoz, P. Resende, M.D.V., Garrick, D.J., Fernando, R.L., Davis, J.M., Jokela, E.J., Martin, T.A., Peter, G.F., and Kirst, M. 2012. Accuracy of genomic selection methods in a standard data set of loblolly pine (Pinus taeda L.). Genetics 190:1503-1510.

## Examples

geno.pine926[1:5, 1:5]
geno.salmon Genotypic data for Atlantic salmon dataset

## Description

Genotypic data on 1,481 Atlantic salmon samples. A total of 17,156 SNP markers (coded as 0,1 , 2 and NA for missing) are included in this dataset. Dataset obtained from supplementary material in Robledo et al. (2018).

## Usage

geno.salmon

## Format

matrix

## References

Robledo D., Matika O., Hamilton A., and Houston R.D. 2018. Genome-wide association and genomic selection for resistance to amoebic gill disease in Atlantic salmon. G3 Genes, Genomes, Genetics 8:1195-1203.

## Examples

```
geno.salmon[1:5, 1:5]
```


## Description

The single-step GBLUP approach combines the information from the pedigree relationship matrix $\boldsymbol{A}$ and the genomic relationship matrix $\boldsymbol{G}$ in one hybrid relationship matrix called $\boldsymbol{H}$. This function will calculate directly the inverse of this matrix $\boldsymbol{H}$. The user should provide the matrices $\boldsymbol{A}$ or its inverse (only one of these is required) and the inverse of the matrix $\boldsymbol{G}\left(\boldsymbol{G}_{\boldsymbol{i n v}}\right)$ in its full form. Individual names should be assigned to rownames and colnames, and individuals from $\boldsymbol{G}_{\boldsymbol{i n v}}$ are verified to be all a subset within individuals from $\boldsymbol{A}$ (or $\boldsymbol{A}_{\text {inv }}$ ).

## Usage

## H.inverse(

A = NULL, Ainv = NULL,
Ginv = NULL,
lambda = NULL,
tau = 1,
omega = 1,
sparseform = FALSE,
keep. order = TRUE,
digits $=8$,
inverse = TRUE,
message = TRUE
)

## Arguments

A

Ainv $\quad$ Input of the inverse of the pedigree relationship matrix $\boldsymbol{A}^{-1}$ in full form ( $n a \times$ $n a)($ default $=$ NULL $)$.
Ginv Input of the inverse of the genomic relationship matrix $G^{-1}$ in full form $(n g \times$ $n g)($ default $=$ NULL $)$.
lambda $\quad$ The scaling factor for $\left(\boldsymbol{G}^{-1}-\boldsymbol{A}_{22}^{-1}\right)($ default $=$ NULL $)$.
tau $\quad$ The scaling factor for $\boldsymbol{G}^{-1}$ (default $=1$ ).
omega $\quad$ The scaling factor for $\boldsymbol{A}_{22}^{-1}($ default $=1)$.
sparseform If TRUE it generates the requested matrix in sparse form to be used directly in asreml with required attributes (default $=$ FALSE).
keep. order If TRUE the original order of the individuals from the $\boldsymbol{A}$ or $\boldsymbol{A}_{\boldsymbol{i n v}}$ matrix is kept. Otherwise the non-genotyped individuals are placed first and then genotyped individuals (default $=$ TRUE) .
digits $\quad$ Set up the number of digits used to round the output matrix $($ default $=8)$.
inverse If TRUE it generates the inverse of $\boldsymbol{H}$ matrix (default = TRUE) (to be deprecated).
message If TRUE diagnostic messages are printed on screen (default $=$ TRUE).

## Details

The generation of the $\boldsymbol{H}^{-1}$ matrix contains a few scaling factors to help with the calculation of this inverse and to allow further exploration of the combination of the information from the $\boldsymbol{A}^{-\mathbf{1}}$ and $G^{-1}$. We follow the specifications described by Martini et. al (2018), which is done by specifying the parameters $\lambda$, or the pair $\tau$ and $\omega$.
If inverse $=$ FALSE the $\boldsymbol{H}$ matrix is provided instead of its inverse. This option will be deprecated and it is better to use the function H.matrix.

## Value

The inverse of the hybrid matrix $\boldsymbol{H}$ matrix, in full or sparse form with required attributes to be used in asreml.

## References

Christensen, O.F., Lund, M.S. 2010. Genomic prediction matrix when some animals are not genotyped. Gen. Sel. Evol. 42(2):1-8.
Christensen, O., Madsen, P., Nielsen, B., Ostersen, T., and Su, G. 2012. Single-step methods for genomic evaluation in pigs. Animal 6(10):1565-1571.
Legarra, A., Aguilar, I., and Misztal, I. 2009. A relationship matrix including full pedigree and genomic information. J. Dairy Sci. 92:4656-4663.
Martini, J.W.R., Schrauf, M.F., Garcia-Baccino, C.A., Pimentel, E.C.G., Munilla, S., RogbergMuñoz, A., Cantet, R.J.C., Reimer, C., Gao, N., Wimmer, V., and Simianer, H. 2018. The effect of the $H^{-1}$ scaling factors $\tau$ and $\omega$ on the structure of $H$ in the single-step procedure. Genet. Sel. Evol. 50:1-9.

## Examples

```
# Get A matrix.
A <- AGHmatrix::Amatrix(data = ped.pine)
A[1:5,1:5]
dim(A)
# Read and filter genotypic data.
M.clean <- qc.filtering(
    M = geno.pine655,
    maf = 0.05,
    marker.callrate = 0.2, ind.callrate = 0.20,
    na.string = "-9",
    plots = FALSE)$M.clean
# Get G matrix.
G <- G.matrix(M.clean, method = "VanRaden", na.string = "-9")$G
G[1:5, 1:5]
```

```
dim(G)
# Match G and A.
check <- match.G2A(
    A = A, G = G,
    clean = TRUE, ord = TRUE, mism = TRUE, RMdiff = TRUE)
# Align G matrix with A.
G_align <- G.tuneup(G = check$Gclean, A = check$Aclean, align = TRUE, sparseform = FALSE)$Gb
# Get Ginverse using the G aligned.
Ginv <- G.inverse(G = G_align, sparseform = FALSE)$Ginv
Ginv[1:5, 1:5]
dim(Ginv)
# Obtain Hinv.
Hinv <- H.inverse(A = A, Ginv = Ginv, lambda = 0.90, sparseform = TRUE)
head(Hinv)
attr(Hinv, "INVERSE")
```

    H.matrix Generates the hybrid H matrix
    
## Description

The single-step GBLUP approach combines the information from the pedigree relationship matrix $\boldsymbol{A}$ and the genomic relationship matrix $\boldsymbol{G}$ in one hybrid relationship matrix called $\boldsymbol{H}$. This function will calculate directly this matrix $\boldsymbol{H}$. The user should provide the matrices $\boldsymbol{A}$ or its inverse (only one of these is required) and the inverse of the matrix $\boldsymbol{G}\left(\boldsymbol{G}_{\boldsymbol{i n v}}\right)$ in its full form. Individual names should be assigned to rownames and colnames, and individuals from $\boldsymbol{G}_{\boldsymbol{i n v}}$ are verified to be all a subset within individuals from $\boldsymbol{A}$ ( or $\boldsymbol{A}_{\boldsymbol{i n v}}$ ). This function is a wrapper of the H.inverse function.

## Usage

```
H.matrix(
    A = NULL,
    Ainv = NULL,
    Ginv = NULL,
    lambda = NULL,
    tau = 1,
    omega = 1,
    sparseform = FALSE,
    keep.order = TRUE,
    digits = 8,
    message = TRUE
)
```


## Arguments

A

Ainv $\quad$ Input of the inverse of the pedigree relationship matrix $\boldsymbol{A}^{-1}$ in full form ( $n a \times$ $n a)($ default $=$ NULL $)$.
Ginv Input of the inverse of the genomic relationship matrix $\boldsymbol{G}^{-1}$ in full form ( $n g \times$ $n g)($ default $=$ NULL $)$.
lambda $\quad$ The scaling factor for $\left(\boldsymbol{G}^{-1}-\boldsymbol{A}_{22}^{-1}\right)($ default $=$ NULL $)$.
tau The scaling factor for $\boldsymbol{G}^{-1}($ default $=1)$.
omega $\quad$ The scaling factor for $\boldsymbol{A}_{22}^{-1}($ default $=1)$.
sparseform If TRUE it generates the requested matrix in sparse form to be used directly in asreml with required attributes $($ default $=$ FALSE $)$.
keep. order If TRUE the original order of the individuals from the $\boldsymbol{A}$ or $\boldsymbol{A}_{\text {inv }}$ matrix is kept. Otherwise the non-genotyped individuals are placed first and then genotyped individuals (default = TRUE).
digits $\quad$ Set up the number of digits used to round the output matrix $($ default $=8)$.
message If TRUE diagnostic messages are printed on screen (default $=$ TRUE).

## Details

This function is currently equivalent to using H.inverse with (inverse $=$ FALSE).

## Value

The hybrid matrix $\boldsymbol{H}$ matrix, in full or sparse form.

## References

Christensen, O.F., Lund, M.S. 2010. Genomic prediction matrix when some animals are not genotyped. Gen. Sel. Evol. 42(2):1-8.
Christensen, O., Madsen, P., Nielsen, B., Ostersen, T., and Su, G. 2012. Single-step methods for genomic evaluation in pigs. Animal 6(10):1565-1571.
Legarra, A., Aguilar, I., and Misztal, I. 2009. A relationship matrix including full pedigree and genomic information. J. Dairy Sci. 92:4656-4663.
Martini, J.W.R., Schrauf, M.F., Garcia-Baccino, C.A., Pimentel, E.C.G., Munilla, S., RogbergMuñoz, A., Cantet, R.J.C., Reimer, C., Gao, N., Wimmer, V., and Simianer, H. 2018. The effect of the $H^{-1}$ scaling factors $\tau$ and $\omega$ on the structure of $H$ in the single-step procedure. Genet. Sel. Evol. 50:1-9.

## Examples

```
# Get A matrix.
A <- AGHmatrix::Amatrix(data = ped.pine)
A[1:5,1:5]
```

```
dim(A)
# Read and filter genotypic data.
M.clean <- qc.filtering(
    M = geno.pine655,
    maf = 0.05,
    marker.callrate = 0.2, ind.callrate = 0.20,
    na.string = "-9",
    plots = FALSE)$M.clean
# Get G matrix.
G <- G.matrix(M = M.clean, method = "VanRaden", na.string = "-9")$G
G[1:5, 1:5]
dim(G)
# Match G2A.
check <- match.G2A(
    A = A, G = G,
    clean = TRUE, ord = TRUE, mism = TRUE, RMdiff = TRUE)
# Align G matrix with A.
G_align <- G.tuneup(G = check$Gclean, A = check$Aclean, align = TRUE, sparseform = FALSE)$Gb
# Get Ginverse using the aligned G.
Ginv <- G.inverse(G = G_align, sparseform = FALSE)$Ginv
Ginv[1:5, 1:5]
dim(Ginv)
# Obtaining H.
H <- H.matrix(A = A, G = Ginv, lambda = 0.90, sparseform = FALSE)
H[1:5, 1:5]
```

kinship.diagnostics
Reports summary statistics, plots and filter options for a given kinship matrix $K$

## Description

It reports summary statistics, plots and allows for some filter options for diagonal and off-diagonal elements for a given kinship matrix. The input matrix can be a pedigree-based relationship matrix
$\boldsymbol{A}$, a genomic relationship matrix $\boldsymbol{G}$ or a hybrid relationship matrix $\boldsymbol{H}$. Individual names should be assigned to rownames and colnames.

## Usage

kinship.diagnostics(
K = NULL,
diagonal.thr.large $=1.2$,

```
    diagonal.thr.small = 0.8,
    duplicate.thr = 0.95,
    clean.diagonal = FALSE,
    clean.duplicate = FALSE,
    plots = TRUE,
    sample.plot = 1,
    message = TRUE
)
```


## Arguments

K
Input of a kinship matrix in full format $(n \times n)($ default $=$ NULL $)$.
diagonal.thr.large
A threshold value to flag large diagonal values $($ default $=1.2)$.
diagonal.thr.small
A threshold value to flag small diagonal values $($ default $=0.8)$.
duplicate.thr A threshold value to flag possible duplicates. Any calculation larger than the threshold based on $\boldsymbol{k}_{i, i} / \sqrt{\boldsymbol{k}_{i, i} \times \boldsymbol{k}_{j, j}}$ is identified as a duplicate (default $=$ $0.95)$.
clean. diagonal If TRUE returns a kinship matrix filtered by values smaller than diagonal. thr. large and larger than diagonal. thr. small $($ default $=$ FALSE $)$.
clean.duplicate
If TRUE return a kinship matrix without the flagged duplicate individuals. All individuals involved are removed $($ default $=$ FALSE $)$.
plots If TRUE generates graphical output of the diagonal and off-diagonal values of the kinship matrix $($ default $=$ TRUE $)$.
sample.plot A numeric value between 0 and 1 indicating the proportion of the data points to be sampled for fast plotting of off-diagonal values. Note that for proportions other than 1 , the method is not exact and low proportions are only recommended for large kinship matrices $($ default $=1)$.
message $\quad$ If TRUE diagnostic messages are printed on screen $($ default $=T R U E)$.

## Value

A list with the following elements:

- list. diagonal: a data frame with the list of flagged large or small diagonal values.
- list. duplicate: a data frame with the list of possible duplicates.
- clean. kinship: output of kinship matrix filtered without the flagged diagonal and/or duplicate individuals.
- plot. diagonal: histogram with the distribution of diagonal values from the kinship matrix.
- plot.offdiag: histogram with the distribution of off-diagonal values from kinship matrix.


## Examples

```
# Get G matrix.
G <- G.matrix(M = geno.apple, method = "VanRaden")$G
# Diagnose G.
G_summary <- kinship.diagnostics(
    K = G,
    diagonal.thr.large = 1.3, diagonal.thr.small = 0.7, clean.diagonal = TRUE,
    duplicate.thr = 0.8, clean.duplicate = TRUE,
    sample.plot = 0.50)
ls(G_summary)
dim(G_summary$clean.kinship)
G_summary$clean.kinship[1:5, 1:5]
G_summary$list.duplicate
G_summary$list.diagonal
G_summary$plot.diag
G_summary$plot.offdiag
```

kinship.heatmap Enhanced heatmap plot for a kinship matrix $K$

## Description

Generates a heatmap with dendrogram based on a provided kinship matrix. This matrix can be a pedigree relationship matrix $\boldsymbol{A}$, a genomic relationship matrix $G$ or a hybrid relationship matrix $\boldsymbol{H}$. Individual names should be assigned to rownames and colnames. It sorts individuals according to dendrogram in both columns and rows.

## Usage

```
kinship.heatmap(
    K = NULL,
    dendrogram = TRUE,
    clustering.method = c("hierarchical", "kmeans"),
    dist.method = c("euclidean", "maximum", "manhattan", "canberra", "binary", "minkowski"),
    row.label = TRUE,
    col.label = FALSE
)
```


## Arguments

K
dendrogram
clustering.method
The clustering method considered for the dendrogram. Options are: "hierarchical" and "kmeans" (default = "hierarchical").
dist.method The method considered to calculate the distance matrix between individuals used for hierarchical clustering. Options are: "euclidean", "maximum", "manhattan", "canberra", "binary" and "minkowski" (default = "euclidean").
row. label If TRUE the individual names (rownames) are added as labels to the left of the heatmap $($ default $=$ TRUE $)$.
col.label If TRUE the individual names (colnames) are added as labels to the bottom of the heatmap $($ default $=$ FALSE $)$.

## Details

Uses the library superheat from Barter and Yu (2018) to generate plots.

## Value

A plot with the properties specified by the above arguments.

## References

Barter, R.L. and Yu, B. 2018. Superheat: An R package for creating beautiful and extendable heatmaps for visualizing complex data. J. Comput. Graph. Stat. 27(4):910-922.

## Examples

```
# Get G matrix.
G <- G.matrix(M = geno.apple, method = "VanRaden")$G
G[1:5, 1:5]
# Plot a subset of the individuals.
kinship.heatmap(K = G[1:10, 1:10], dendrogram = TRUE, row.label = TRUE, col.label = TRUE)
```

| kinship.pca | Performs a Principal Component Analysis (PCA) based on a kinship <br> matrix $K$ |
| :--- | :--- |

## Description

Generates a PCA and summary statistics from a given kinship matrix for population structure. This matrix can be a pedigree-based relationship matrix $\boldsymbol{A}$, a genomic relationship matrix $\boldsymbol{G}$ or a hybrid relationship matrix $\boldsymbol{H}$. Individual names should be assigned to rownames and colnames. There is additional output such as plots and other data frames to be used on other downstream analyses (such as GWAS).

## Usage

```
kinship.pca(
    K = NULL,
    scale = TRUE,
    label = FALSE,
    ncp = 10,
    groups = NULL,
    ellipses = FALSE
    )
```


## Arguments

K
scale If TRUE the PCA analysis will scale the kinship matrix, otherwise it is used in its original scale (default = TRUE).
label If TRUE then includes in output individuals names $($ default $=$ FALSE $)$.
ncp $\quad$ The number of PC dimensions to be shown in the screeplot, and to provide in the output data frame $($ default $=10)$.
groups $\quad$ Specifies a vector of class factor that will be used to define different colors for individuals in the PCA plot. It must be presented in the same order as the individuals in the kinship matrix $($ default $=$ NULL $)$.
ellipses If TRUE, ellipses will will be drawn around each of the define levels in groups (default $=$ FALSE).

## Details

It calls function eigen() to obtain eigenvalues and later generate the PCA and the factoextra R package to extract and visualize results.

## Value

A list with the following four elements:

- eigenvalues: a data frame with the eigenvalues and its variances associated with each dimension including only the first ncp dimensions.
- pca. scores: a data frame with scores (rotated observations on the new components) including only the first ncp dimensions.
- plot.pca: a scatterplot with the first two-dimensions (PC1 and PC2) and their scores.
- plot. scree: a barchart with the percentage of variances explained by the ncp dimensions.


## Examples

```
# Get G matrix.
G <- G.matrix(M = geno.apple, method = "VanRaden")$G
G[1:5, 1:5]
# Perform the PCA.
```

```
G_pca <- kinship.pca(K = G, ncp = 10)
ls(G_pca)
G_pca$eigenvalues
head(G_pca$pca.scores)
G_pca$plot.pca
G_pca$plot.scree
# PCA plot by family (17 groups).
grp <- as.factor(pheno.apple$Family)
G_pca_grp <- kinship.pca(K = G, groups = grp, label = FALSE, ellipses = FALSE)
G_pca_grp$plot.pca
```

match.G2A Check the genomic relationship matrix $G$ against the pedigree rela-
tionship matrix A or vice versa

## Description

Assesses a given genomic relationship matrix $\boldsymbol{G}$ against the pedigree relationship matrix $\boldsymbol{A}$, or vice versa, to determine the matched and mismatched individuals. If requested, it provides the cleaned versions containing only the matched individuals between both matrices. The user should provide the matrices $\boldsymbol{G}$ and $\boldsymbol{A}$ in full form ( $n g \times n g$ and $n a \times n a$, respectively). Individual names should be assigned to rownames and colnames for both matrices.

## Usage

match.G2A(
A = NULL,
$G=N U L L$,
clean = TRUE,
ord = TRUE,
mism = FALSE,
RMdiff = FALSE,
message $=$ TRUE
)

## Arguments

A

G
clean If TRUE generates new clean $\boldsymbol{G}$ and $\boldsymbol{A}$ matrices in full form containing only matched individuals $($ default $=$ TRUE $)$.
ord
Input of the pedigree relationship matrix $\boldsymbol{A}$ in full form $(n a \times n a)($ default $=$ NULL).

Input of the genomic relationship matrix $\boldsymbol{G}$ in full form $(n g \times n g)$ (default $=$ NULL).

If TRUE it will order by ascending order of individual names both of the clean $\boldsymbol{A}$ and $\boldsymbol{G}$ matrices $($ default $=$ TRUE).

| mism | If TRUE generates two data frames with mismatched individual names from the |
| :--- | :--- |
| $\boldsymbol{G}$ and $\boldsymbol{A}$ matrices (default = FALSE). |  |
| RMdiff | If TRUE it generates the matrix (in lower diagonal row-wise sparse form) of <br> matched observations from both the $\boldsymbol{G}$ and $\boldsymbol{A}$ matrices. This matrix can be <br> used to identify inconsistent values between matched matrices, but it can be <br> very large (default = FALSE). |
| message | If TRUE diagnostic messages are printed on screen (default = TRUE). |

## Value

A list with the following elements:

- Gclean: a matrix with the portion of $\boldsymbol{G}$ containing only matched individuals.
- Aclean: a matrix with the portion of $\boldsymbol{A}$ containing only matched individuals.
- mismG: a vector containing the names of the individuals from matrix $G$ that are missing in matrix $\boldsymbol{A}$.
- mismA: a vector containing the names of the individuals from matrix $\boldsymbol{A}$ that are missing in matrix $\boldsymbol{G}$.
- RM: a data frame with the observations from both the $\boldsymbol{G}$ and $\boldsymbol{A}$ matched matrices, together with their absolute relationship difference.
- plotG2A: scatterplot with the pairing of matched pedigree- against genomic-based relationship values. This graph might take a long to plot with large datasets.


## Examples

```
# Get A matrix.
A <- AGHmatrix::Amatrix(data = ped.pine)
A[1:5,1:5]
dim(A)
# Read and filter genotypic data.
M.clean <- qc.filtering(
    M = geno.pine655,
    maf = 0.05,
    marker.callrate = 0.2, ind.callrate = 0.20,
    na.string = "-9",
    plots = FALSE)$M.clean
# Get G matrix.
G <- G.matrix(M = M.clean, method = "VanRaden", na.string = "-9")$G
G[1:5, 1:5]
dim(G)
# Match G2A.
check <- match.G2A(
    A = A, G = G,
    clean = TRUE, ord = TRUE, mism = TRUE, RMdiff = TRUE)
ls(check)
```

match.kinship2pheno

```
dim(check$Aclean)
dim(check$Gclean)
check$Aclean[1:5, 1:5]
check$Gclean[1:5, 1:5]
head(check$mismG)
head(check$mismA)
check$plotG2A
head(check$RM)
```

match.kinship2pheno Check any kinship matrix K against phenotypic data

## Description

Assesses a given kinship matrix against the provided phenotypic data to determine if all genotypes are in the kinship matrix or not. It also reports which individuals match or are missing from one set or another. If requested, a reduced kinship matrix is generated that has only the matched individuals. The input kinship matrix can be a pedigree-based relationship matrix $\boldsymbol{A}$, a genomic-based relationship matrix $\boldsymbol{G}$, or a hybrid relationship matrix $\boldsymbol{H}$. Individual names should be assigned to rownames and colnames of input matrix.

## Usage

match.kinship2pheno(
K = NULL,
pheno.data $=$ NULL,
indiv = NULL,
clean = FALSE,
ord = TRUE,
mism = FALSE,
message = TRUE
)

## Arguments

K
Input of a kinship matrix in full form $(n \times n)($ default $=$ NULL $)$;
pheno.data A data fame with the phenotypic data to assess (for n individuals) (default $=$ NULL).
indiv The string for the column name for genotypes/individuals in the phenotypic data (default $=$ NULL).
clean If TRUE, generates a new clean kinship matrix containing only the matched phenotyped individuals (default $=$ FALSE).
ord If TRUE, it will order the kinship matrix as in the phenotypic data, which is recommended for some downstream genomic analyses (default $=$ TRUE $)$.

| mism | If TRUE, it generates data frames with matched and mismatched individual's |
| :--- | :--- |
|  | names from the kinship matrix and the phenotypic data $($ default = FALSE). |
| message | If TRUE diagnostic messages are printed on screen (default = TRUE). |

## Value

A list with the following elements:

- mismatchesK: a vector containing the names of the individuals from the provided kinship matrix that mismatch with the phenotypic data.
- matchesK: a vector containing the names of the individuals from the provided kinship matrix that match with the phenotypic data.
- mismatchesP: a vector containing the names of phenotyped individuals that mismatch with those from the kinship matrix.
- matchesP: a vector containing the names of phenotyped individuals that match with those from the kinship matrix.
- Kclean: a clean kinship matrix containing only the matched phenotyped individuals.


## Examples

```
## Not run:
# Get G matrix.
G <- G.matrix(M = geno.pine655, method = "VanRaden", na.string = "-9", sparseform = FALSE)$G
dim(G)
# Match G and the phenotype.
check <-
    match.kinship2pheno(
        K = G, pheno.data = pheno.pine,
        indiv = "Genotype",
        clean = TRUE, mism = TRUE)
ls(check)
length(check$matchesK)
length(check$mismatchesK)
length(check$matchesP)
length(check$mismatchesP)
dim(check$Kclean)
## End(Not run)
```

ped.pine Pedigree data for loblolly pine dataset

## Description

Individual pedigree data for a total of 2,034 records of loblolly pine (Pinus taeda L.). Missing parental information coded as 0 . Dataset obtained from supplementary material in Resende et al. (2012).

## Usage

ped. pine

## Format

data.frame

## References

Resende, M.F.R., Munoz, P. Resende, M.D.V., Garrick, D.J., Fernando, R.L., Davis, J.M., Jokela, E.J., Martin, T.A., Peter, G.F., and Kirst, M. 2012. Accuracy of genomic selection methods in a standard data set of loblolly pine (Pinus taeda L.). Genetics 190:1503-1510.

## Examples

```
ped.pine |> head()
```

ped.salmon Pedigree data for Atlantic salmon dataset

## Description

Pedigree data of 1,481 Atlantic salmon samples. Missing parental information coded as 0 . Dataset obtained from supplementary material in Robledo et al. (2018).

## Usage

ped.salmon

## Format

data.frame

## References

Robledo D., Matika O., Hamilton A., and Houston R.D. 2018. Genome-wide association and genomic selection for resistance to amoebic gill disease in Atlantic salmon. G3 Genes, Genomes, Genetics 8:1195-1203.

## Examples

ped.salmon |> head()

## Description

Phenotypic data on 247 apple clones (i.e., genotypes) evaluated for several fruit quality traits at two New Zealand sites, Motueka (MOT) and Hawkes Bay (HB). Dataset obtained from supplementary material in Kumar et al. (2015).

## Usage

pheno.apple

## Format

data.frame

## References

Kumar S., Molloy C., Muñoz P., Daetwyler H., Chagné D., and Volz R. 2015. Genome-enabled estimates of additive and nonadditive genetic variances and prediction of apple phenotypes across environments. G3 Genes, Genomes, Genetics 5:2711-2718.

## Examples

pheno.apple |> head()
pheno.pine
Phenotypic data for loblolly pine dataset

## Description

Deregressed estimated breeding values (DEBV) for the trait diameter at breast height (DBH) at 6 years of age from trees grown at site Nassau. The dataset contains a total of 861 genotypes of loblolly pine (Pinus taeda L.). Dataset obtained from supplementary material in Resende et al. (2012).

## Usage

pheno.pine

## Format

data.frame

## References

Resende, M.F.R., Munoz, P. Resende, M.D.V., Garrick, D.J., Fernando, R.L., Davis, J.M., Jokela, E.J., Martin, T.A., Peter, G.F., and Kirst, M. 2012. Accuracy of genomic selection methods in a standard data set of loblolly pine (Pinus taeda L.). Genetics 190:1503-1510.

## Examples

```
pheno.pine |> head()
```

pheno.salmon Phenotypic data for Atlantic salmon dataset

## Description

Phenotypic data on 1,481 Atlantic salmon individuals. All fish were phenotyped for mean gill score (mean of the left gill and right gill scores) and amoebic load (qPCR values using Neoparamoeba perurans specific primers, amplified from one of the gills). Dataset obtained from supplementary material in Robledo et al. (2018).

## Usage

pheno.salmon

## Format

data.frame

## References

Robledo D., Matika O., Hamilton A., and Houston R.D. 2018. Genome-wide association and genomic selection for resistance to amoebic gill disease in Atlantic salmon. G3 Genes, Genomes, Genetics 8:1195-1203.

## Examples

```
pheno.salmon |> head()
```

Quality control filtering of molecular matrix $M$ for downstream analyses

## Description

Reads molecular data in the format $0,1,2$ and performs some basic quality control filters and simple imputation. Matrix provided is of the full form $(n \times p)$, with $n$ individuals and $p$ markers. Individual and marker names are assigned to rownames and colnames, respectively. Filtering can be done with the some of the following options by specifying thresholds for: missing values on individuals, missing values on markers, minor allele frequency, inbreeding Fis value (of markers), and observed heterozygosity (of markers). String used for identifying missing values can be specified. If requested, missing values will be imputed based on the mean of each SNP.

## Usage

```
qc.filtering(
        M = NULL,
        base = FALSE,
        na.string = NA,
        map \(=\) NULL,
        marker = NULL,
        chrom \(=\) NULL,
        pos \(=\) NULL,
        ref \(=\) NULL,
        marker.callrate \(=1\),
        ind.callrate \(=1\),
        maf = 0,
        heterozygosity \(=1\),
        Fis = 1,
        impute = FALSE,
        Mrecode = FALSE,
        plots = TRUE,
        digits \(=2\),
        message \(=\) TRUE
    )
```


## Arguments

M
A matrix with SNP data of full form $(n \times p)$, with $n$ individuals and $p$ markers Individual and marker names are assigned to rownames and colnames, respectively. Data in matrix is coded as $0,1,2$ (integer or numeric) $($ default $=N U L L)$.
base If TRUE matrix $M$ is considered as bi-allele SNP data format (character) and the SNPs are recoded to numerical values before performing the quality control filters $($ default $=$ FALSE $)($ currently deprecated $)$.
na.string A character that will be interpreted as NA values (default = "NA").

| map | $($ Optional $)$ A data frame with the map information with $p$ rows (default $=$ NULL). |
| :---: | :---: |
| marker | A character indicating the name of the column in data frame map with the identification of markers. This is mandatory if map is provided $($ default $=$ NULL $)$. |
| chrom | A character indicating the name of the column in data frame map with the identification of chromosomes (default $=$ NULL). |
| pos | A character indicating the name of the column in data frame map with the identification of marker positions $($ default $=$ NULL $)$. |
| ref | A character indicating the name of the column in the map containing the reference allele for recoding. If absent, then conversion will be based on the major allele (most frequent). The marker information of a given individuals with two of the specified major alleles in ref will be coded as $2($ default $=$ NULL $)$. |
| marker.callrate |  |
|  | A numerical value between 0 and 1 used to remove SNPs with a rate of missing values equal or larger than this value (default $=1$, i.e. no removing). |
| ind.callrate | A numerical value between 0 and 1 used to remove individuals with a rate of missing values equal or larger than this value (default $=1$, i.e. no removing). |
| maf | A numerical value between 0 and 1 used to remove SNPs with a Minor Allele Frequency (MAF) below this value (default $=0$, i.e. no removing). |
| heterozygosity | A numeric value indicating the maximum value of accepted observed heterozygosity $(\mathrm{Ho})($ default $=1$, i.e. no removing). |
| Fis | A numeric value indicating the maximum value of accepted inbreeding (Fis) following the equation $\|1-(H o / H e)\|$ (default $=1$, i.e. no removing). |
| impute | If TRUE imputation of missing values is done using the mean of each SNP (default $=$ FALSE). |
| Mrecode | If TRUE it provides the recoded $\boldsymbol{M}$ matrix from the bi-allelic to numeric SNP $($ default $=$ FALSE $)($ currently deprecated $)$. |
| plots | If TRUE generates graphical output of the quality control based on the original input matrix $($ default $=T R U E)$. |
| digits | Set up the number of digits used to round the output matrix ( default $=2$ ). |
| message | If TRUE diagnostic messages are printed on screen ( default $=$ TRUE $)$. |

## Details

Warning: The arguments base, ref, and Mrecode currently are deprecated and will be removed on the next version of ASRgenomics. Use function snp.recode to recode the matrix prior to using qc.filtering.
The filtering process is carried out as expressed in the following simplified pseudo-code that consists on a loop repeated twice:

## for $\mathbf{i}$ in 1 to 2

Filter markers based on call rate.
Filter individuals based on call rate.
Filter markers based on minor allele frequency.
Filter markers based on observed heterozygosity.

Filter markers based on inbreeding.
end for

## Value

A list with the following elements:

- M. clean: the cleaned $\boldsymbol{M}$ matrix after the quality control filters have been applied.
- map: if provided, a cleaned map data frame after the quality control filters have been applied.
- plot.missing.ind: a plot of missing data per individual (original marker matrix).
- plot.missing. SNP: a plot of missing data per SNP (original marker matrix).
- plot. heteroz: a plot of observed heterozygocity per SNP (original marker matrix).
- plot.Fis: a plot of Fis per SNP (original marker matrix).
- plot.maf: a plot of the minor allele frequency (original marker matrix).


## Examples

```
# Example: Pine dataset from ASRgenomics (coded as 0,1,2 with missing as -9).
M.clean <- qc.filtering(
    M = geno.pine926,
    maf = 0.05,
    marker.callrate = 0.9, ind.callrate = 0.9,
    heterozygosity = 0.9, Fis = 0.6,
    na.string = "-9")
ls(M.clean)
M.clean$M.clean[1:5, 1:5]
dim(M.clean$M.clean)
head(M.clean$map)
M.clean$plot.maf
M.clean$plot.missing.ind
M.clean$plot.missing.SNP
M.clean$plot.heteroz
M.clean$plot.Fis
# Example: Salmon dataset (coded as 0,1,2 with missing as NA).
M.clean <- qc.filtering(
    M = geno.salmon,
    maf = 0.02,
    marker.callrate = 0.10, ind.callrate = 0.20,
    heterozygosity = 0.9, Fis = 0.4)
M.clean$M.clean[1:5, 1:5]
dim(M.clean$M.clean)
head(M.clean$map)
M.clean$plot.maf
M.clean$plot.missing.ind
M.clean$plot.missing.SNP
M.clean$plot.heteroz
```

```
M.clean\$plot.Fis
```

Performs a Principal Component Analysis (PCA) based on a molecular matrix $M$

## Description

Generates a PCA and summary statistics from a given molecular matrix for population structure. Matrix provided is of full form $(n \times p)$, with n individuals and p markers. Individual and marker names are assigned to rownames and colnames, respectively. SNP data is coded as $0,1,2$ (integers or decimal numbers). Missing values are not accepted and these need to be imputed (see function qc.filtering() for implementing mean imputation). There is additional output such as plots and other data frames to be used on other downstream analyses (such as GWAS).

## Usage

snp.pca( $\mathrm{M}=\mathrm{NULL}$, label $=$ FALSE, ncp $=10$, groups $=$ NULL, ellipses = FALSE)

## Arguments

M A matrix with SNP data of full form $(n \times p)$, with $n$ individuals and $p$ markers (default = NULL).
label If TRUE then includes in output individuals names $($ default $=$ FALSE $)$.
ncp The number of PC dimensions to be shown in the screeplot, and to provide in the output data frame $($ default $=10)$.
groups Specifies a vector of class factor that will be used to define different colors for individuals in the PCA plot. It must be presented in the same order as the individuals in the molecular $M$ matrix (default $=$ NULL).
ellipses If TRUE, ellipses will will be drawn around each of the define levels in groups (default $=$ FALSE).

## Details

It calls function prcomp() to generate the PCA and the factoextra R package to extract and visualize results. Methodology uses normalized allele frequencies as proposed by Patterson et al. (2006).

## Value

A list with the following four elements:

- eigenvalues: a data frame with the eigenvalues and its variances associated with each dimension including only the first ncp dimensions.
- pca. scores: a data frame with scores (rotated observations on the new components) including only the first ncp dimensions.
- plot.pca: a scatterplot with the first two-dimensions (PC1 and PC2) and their scores.
- plot. scree: a barchart with the percentage of variances explained by the ncp dimensions.


## References

Patterson N., Price A.L., and Reich, D. 2006. Population structure and eigenanalysis. PLoS Genet 2(12):e190. doi:10.1371/journal.pgen. 0020190

## Examples

```
# Perform the PCA.
SNP_pca <- snp.pca(M = geno.apple, ncp = 10)
ls(SNP_pca)
SNP_pca$eigenvalues
head(SNP_pca$pca.scores)
SNP_pca$plot.pca
SNP_pca$plot.scree
# PCA plot by family (17 groups).
grp <- as.factor(pheno.apple$Family)
SNP_pca_grp <- snp.pca(M = geno.apple, groups = grp, label = FALSE)
SNP_pca_grp$plot.pca
```

Reduces the number of redundant markers on a molecular matrix $M$ by pruning

## Description

For a given molecular dataset $\boldsymbol{M}$ (in the format 0,1 and 2 ) it produces a reduced molecular matrix by eliminating "redundant" markers using pruning techniques. This function finds and drops some of the SNPs in high linkage disequilibrium (LD).

## Usage

```
snp.pruning(
    \(M=N U L L\),
    map \(=\) NULL,
    marker = NULL,
    chrom = NULL,
    pos = NULL,
    method = c("correlation"),
    criteria = c("callrate", "maf"),
    pruning.thr \(=0.95\),
    by.chrom = FALSE,
```

```
    window.n = 50,
    overlap.n = 5,
    iterations = 10,
    seed = NULL,
    message = TRUE
)
```


## Arguments

M
map
marker
chrom A character indicating the name of the column in data frame map with the identification of chromosomes. This is mandatory if map is provided (default = NULL).
pos A character indicating the name of the column in data frame map with the identification of marker positions $($ default $=$ NULL $)$.
method A character indicating the method (or algorithm) to be used as reference for identifying redundant markers. The only method currently available is based on correlations (default = "correlation").
criteria A character indicating the criteria to choose which marker to drop from a detected redundant pair. Options are: "callrate" (the marker with fewer missing values will be kept) and "maf" (the marker with higher minor allele frequency will be kept) $($ default $=$ "callrate" $)$.
pruning. thr A threshold value to identify redundant markers with Pearson's correlation larger than the value provided (default $=0.95$ ).
by.chrom If TRUE the pruning is performed independently by chromosome $($ default $=$ FALSE).
window.n A numeric value with number of markers to consider in each window to perform pruning $($ default $=50)$.
overlap.n A numeric value with number of markers to overlap between consecutive windows (default $=5$ ).
iterations An integer indicating the number of sequential times the pruning procedure should be executed on remaining markers. If no markers are dropped in a given iteration/run, the algorithm will stop $($ default $=10)$.
seed An integer to be used as seed for reproducibility. In case the criteria has the same values for a given pair of markers, one will be dropped at random (default $=$ NULL).
message If TRUE diagnostic messages are printed on screen (default $=$ TRUE).

## Details

Pruning is recommended as redundancies can affect the quality of matrices used for downstream analyses. The algorithm used is based on the Pearson's correlation between markers as a proxy for LD. In the event of a pairwise correlation higher than the selected threshold markers will be eliminated as specified by: call rate, minor allele frequency. In case of tie, one marker will be dropped at random.
Filtering markers (qc.filtering) is of high relevance before pruning. Poor quality markers (e.g., monomorphic markers) may prevent correlations from being calculated and may affect eliminations.

## Value

- Mpruned: a matrix containing the pruned marker $M$ matrix.
- map: an data frame containing the pruned map.


## Examples

```
# Read and filter genotypic data.
M.clean <- qc.filtering(
    M = geno.pine655,
    maf = 0.05,
    marker.callrate = 0.20, ind.callrate = 0.20,
    Fis = 1, heterozygosity = 0.98,
    na.string = "-9",
    plots = FALSE)$M.clean
# Prune correlations > 0.9.
Mpr <- snp.pruning(
    M = M.clean, pruning.thr = 0.90,
    by.chrom = FALSE, window.n = 40, overlap.n = 10)
head(Mpr$map)
Mpr$Mpruned[1:5, 1:5]
```


## Description

Reads molecular data in format of bi-allelic nucleotide bases (AA, AG, GG, CC, etc.) and recodes them as $0,1,2$ and NA to be used in other downstream analyses.

## Usage

```
snp.recode(
    M = NULL,
    map = NULL,
    marker = NULL,
```

```
    ref = NULL,
    alt = NULL,
    recoding = c("ATGCto012"),
    na.string = NA,
    rename.markers = TRUE,
    message = TRUE
)
```


## Arguments

| M | A character matrix with SNP data of full form $(n \times p)$, with $n$ individuals and $p$ <br> markers Individual and marker names are assigned to rownames and colnames, <br> respectively. Data in matrix is coded as AA, AG, GG, CC, etc (default = NULL). <br> (Optional) A data frame with the map information with $p$ rows. If NULL a dummy <br> map is generated considering a single chromosome and sequential positions for <br> markers and includes reference allele and alternative allele (default = NULL). <br> A character indicating the name of the column in data frame map with the iden- <br> tification of markers. This is mandatory if map is provided (default = NULL). |
| :--- | :--- |
| marker | A character indicating the name of the column in the map containing the refer- <br> ence allele for recoding. If absent, then conversion will be based on the major <br> allele (most frequent). The marker information of a given individual with two <br> of the specified major alleles in ref will be coded as 2. This is mandatory if map <br> is provided (default = NULL). |
| ref $\quad$A character indicating the name of the column in the map containing the alter- <br> native allele for recoding. If absent, then it will be inferred from the data. The <br> marker information of a given individual with two of the specified alleles in alt |  |
| will be coded as 0 (default = NULL). |  |

## Value

A list with the following two elements:

- Mrecode: the molecular matrix $M$ recoded to $0,1,2$ and NA.
- mapr: the data frame with the map information including reference allele and alternative allele.


## Examples

\# Create bi-allelic base data set.
Mnb <- matrix (c(

```
"A-", NA, "GG", "CC", "AT", "CC", "AA", "AA",
"AAA", NA, "GG", "AC", "AT", "CG", "AA", "AT",
"AA", NA, "GG", "CC", "AA", "CG", "AA", "AA",
"AA", NA, "GG", "AA", "AA", NA, "AA", "AA",
"AT", NA, "GG", "AA", "TT", "CC", "AT", "TT",
"AA", NA, NA, "CC", NA, "GG", "AA", "AA",
"AA", NA, NA, "CC", "TT", "CC", "AA", "AT",
"TT", NA, "GG", "AA", "AA", "CC", "AA", "AA"),
ncol = 8, byrow = TRUE, dimnames = list(paste0("ind", 1:8),
paste0("m", 1:8)))
Mnb
# Recode without map (but map is created).
Mr <- snp.recode(M = Mnb, na.string = NA)
Mr$Mrecode
Mr$map
# Create map.
mapnb <- data.frame(
    marker = paste0("m", 1:8),
    reference = c("A", "T", "G", "C", "T", "C", "A", "T"),
    alternative = c("T", "G", "T", "A", "A", "G", "T", "A")
    )
    mapnb
# Recode with map without alternative allele.
Mr <- snp.recode(M = Mnb, map = mapnb, marker = "marker", ref = "reference",
            na.string = NA, rename.markers = TRUE)
Mr$Mrecode
Mr$map
# Notice that the alternative allele is in the map as a regular variable,
# but in the names it is inferred from data (which might be 0 (missing)).
# Recode with map with alternative allele.
Mr <- snp.recode(M = Mnb, map = mapnb, marker = "marker",
    ref = "reference", alt = "alternative",
    na.string = NA, rename.markers = TRUE)
Mr$Mrecode
Mr$map # Now the alternative is also on the names.
# We can also recode without renaming the markers.
Mr <- snp.recode(M = Mnb, map = mapnb, marker = "marker", ref = "reference",
    na.string = NA, rename.markers = FALSE)
Mr$Mrecode
Mr$map # Now the alternative is also on the names.
```


## Description

Modifies the input sparse form matrix into its full form. The sparse form has three columns per line, corresponding to the set: Row, Col, Value, and is defined by a lower triangle row-wise of the full matrix and is sorted as columns within row. Individual names should be assigned as attributes: attr (K, "rowNames") and attr(K, "colNames"). If these are not provided they are considered as 1 to $n$.

## Usage

sparse2full (K = NULL)

## Arguments

K A square matrix in sparse form $($ default $=$ NULL $)$.

## Details

Based on a function from ASReml-R 3 library by Butler et al. (2009).

## Value

A full square matrix where individual names are assigned to rownames and colnames. If attribute INVERSE is found this is also passed to the full matrix.

## References

Butler, D.G., Cullis, B.R., Gilmour, A.R., and Gogel, B.J. 2009. ASReml-R reference manual. Version 3. The Department of Primary Industries and Fisheries (DPI\&F).

## Examples

```
# Get G matrix.
Gsp <- G.matrix(M = geno.apple, method = "VanRaden", sparseform = TRUE)$G.sparse
head(Gsp)
head(attr(Gsp, "rowNames"))
# Transform into full matrix.
G <- sparse2full(K = Gsp)
G[1:5, 1:5]
```


## Description

This function generates (or imputes) a molecular matrix for offspring from hypothetical crosses based on the genomic information from the parents. This is a common procedure in species such as maize, where only the parents (inbred lines) are genotyped, and this information is used to generate/impute the genotypic data of each of the hybrid offspring. This function can be also used for bulked DNA analyses, in order to obtain an bulked molecular matrix for full-sib individuals were only parents are genotyped.

## Usage

synthetic.cross(
M = NULL, ped $=$ NULL, indiv = NULL, mother = NULL, father = NULL, heterozygote.action = c("useNA", "exact", "fail", "expected"), na.action = c("useNA", "expected"), message $=$ TRUE
)

## Arguments

M
A matrix with marker data of full form $(n \times p)$, with $n$ individuals (mothers and fathers) and $p$ markers. Individual and marker names are assigned to rownames and colnames, respectively. Data in matrix is coded as $0,1,2$ (integer or numeric) $($ default $=$ NULL $)$.
ped A data frame with three columns containing only the pedigree of the hypothetical offspring. (not pedigree of parents) It should include the three columns for individual, mother and father $($ default $=$ NULL $)$.
indiv A character indicating the column in ped data frame containing the identification of the offspring $($ default $=$ NULL $)$.
mother A character indicating the column in ped data frame containing the identification of the mother (default = NULL).
father A character indicating the column in ped data frame containing the identification of the father $($ default $=N U L L)$.
heterozygote.action
Indicates the action to take when heterozygotes are found in a marker. Options are: "useNA", "exact", "fail", and "expected". See details for more information (default = "useNA")

| na.action | Indicates the action to take when missing values are found in a marker. Options <br> are: "useNA" and "expected". See details for more information (default = <br> "useNA"). |
| :--- | :--- |
| message | If TRUE diagnostic messages are printed on screen (default = TRUE). |

## Details

For double-haploids, almost the totality of the markers (except for genotyping errors) will be homozygotic reads. But in many other cases (including recombinant inbred lines) there will be a proportion of heterozygotic reads. In these case, it is very difficult to infer (impute) the exact genotype of a given offspring individual. For example, if parents are $0(\mathrm{AA})$ and 1 (AC) then offsprings will differ given this Mendelian sampling. However, different strategies exist to determine the expected value for that specific cross (if required), which are detailed below using the option heterozygote. action.

- If heterozygote. action = "useNA", the generated offspring will have, for the heterozygote read, an NA, and no markers are removed. Hence, no attempt will be done to impute/estimate this value.
- If heterozygote.action = "exact", any marker containing one or more heterozygote reads will be removed. Hence, inconsistent markers are fully removed from the $M$ matrix.
- If heterozygote.action = "fail", function stops and informs of the presence of heterozygote reads.
- If heterozygote.action = "expected", then an algorithm is implemented, on the heterozygote read to determine its expected value. For example, if parents are 0 and 1 , then the expected value (with equal probability) is 0.5 . For a cross between two heterozygotes, the expected value is: $0(1 / 4)+1(1 / 2)+2(1 / 4)=1$. And for a cross between 1 and 2 , the expected value is: $1(1 / 2)+2(1 / 2)=1.5$

Missing value require special treatment, and an imputation strategy is detailed below as indicated using the option na. action.

- If na.action = "useNA", if at least one of the parental reads is missing values for a given marker then it will be assigned as missing for the hypothetical cross. Hence, no attempt will be done to impute/estimate this value.
- If na. action = "expected", then an algorithm is implemented that will impute the expected read of the cross if the genotype of one of the parents is missing (e.g., cross between 0 and NA). Calculations are based on parental allelic frequencies $p$ and $q$ for the given marker. The expressions for expected values are detailed below.
- If the genotype of the non-missing parent read is 0 .
$q^{2}$ (probability that the missing parent is 0 ) x 0 (expected value of the offspring from a 0 x 0 cross: $0(1 / 1)$ ) +
$2 p q$ (probability that the missing parent is 1 ) $\times 0.5$ (expected offspring from a $0 \times 1$ cross: $0(1 / 2)+1(1 / 2))+$
$q^{2}$ (probability that the missing parent is 2 ) $\times 1$ (expected offspring from a $0 \times 2$ cross: $1(1 / 1)$ )
- If the genotype of the non-missing parent read is 1 .
$q^{2}$ (probability that the missing parent is 0 ) $\times 0.5$ (offspring: $\left.0(1 / 2)+1(1 / 2)\right)+$
$2 p q$ (probability that the missing parent is 1 ) $\times 1$ (offspring: $0(1 / 4)+1(1 / 2)+2(1 / 4)$ ) $+$
$q^{2}$ (probability that the missing parent is 2 ) x 1.5 (offspring: $1(1 / 2)+2(1 / 2)$ )
- If the genotype of the non-missing parent read is 2 .
$q^{2}$ (probability that the missing parent is 0$) \times 1$ (offspring: $\left.1(1 / 1)\right)+$
$2 p q$ (probability that the missing parent is 1 ) $\times 1.5$ (offspring: $1(1 / 2)+2(1 / 2))+$ $q^{2}$ (probability that the missing parent is 2 ) $\times 2$ (offspring: $2(1 / 1)$ )
Similarly, the calculation of the expected read of a cross when both parents are missing is also based on population allelic frequencies for the given marker. The expressions for expected values are detailed below.
$q^{2} \times q^{2}$ (probability that both parents are 0 ) $\times 0$ (expected value of the offspring from a $0 \times 0$ cross: $0(1 / 1))+$
$2 \times\left(q^{2} \times 2 p q\right)$ (probability that the first parent is 0 and the second is 1 ; this requires the multiplication by 2 because it is also possible that the first parent is 1 and the second is 0 ) x 0.5 (offspring: $0(1 / 2)+1(1 / 2))+$
$2 \times\left(q^{2} \times p^{2}\right)$ (this could be $0 \times 2$ or $\left.2 \times 0\right) \times 1$ (offspring: $\left.1(1 / 1)\right)+$
$2 p q \times 2 p q$ (both parents are 1) x 1 (offspring: $0(1 / 4)+1(1 / 2)+2(1 / 4))+$ $2 \times(2 p q \times q 2)$ (this could be $1 \times 2$ or $2 \times 1$ ) x 1.5 (offspring: $1(1 / 2)+2(1 / 2))+$ $p^{2} \times p^{2}$ (both parents are 2 ) $\times 2$ (offspring: $2(1 / 1)$ )
Note that the use of na.action = "expected" is recommended when a large number of offspring will conform the hybrid cross (such as with bulked DNA analyses) for family groups with reasonable number of individuals.
Warning. If "expected" is used for heterozygote.action or na.action, direct transformation of the molecular data to other codings (e.g., dominance matrix coded as $c(0,1,0)$ ) is not recommended.


## Value

A molecular matrix $M$ containing the genotypes generated/imputed for the hypothetical cross.

## Examples

```
# Create dummy pedigree (using first 10 as parents).
ped <- data.frame(
    male = rownames(geno.apple)[1:5],
    female = rownames(geno.apple)[6:10])
ped$offs <- paste(ped$male, ped$female, sep = "_")
ped
# Select portion of M for parents.
Mp <- geno.apple[c(ped$male, ped$female), 1:15]
# Get genotype of crosses removing markers with heterozygotes.
synthetic.cross(
    M = Mp, ped = ped,
    indiv = "offs", mother = "female", father = "male",
    heterozygote.action = "exact",
    na.action = "useNA")
```

```
# Request the synthetic cross to be NA in the respective samples.
synthetic.cross(
    M = Mp, ped = ped,
    indiv = "offs", mother = "female", father = "male",
    heterozygote.action = "useNA",
    na.action = "useNA")
# Get genotype of crosses and use expected values.
synthetic.cross(
M = Mp, ped = ped,
indiv = "offs", mother = "female", father = "male",
heterozygote.action = "expected", na.action = "expected")
```


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