Package 'GenEval'

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```
Type Package
Title A package to simulate and evaluate genomic data
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Description GenEval is a package to simulate and evaluate genomic data.
      Genotypes can be simulated either in complete linkage equilibrium (LE) or in linkage disequilib-
      rium (LD). Phenotypes can be simulated as both continuous or discrete.
Depends R (>= 3.1.0),
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      MM4LMM,
      boot,
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Imports base,
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      simpleboot,
      car,
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```

2 BLUPsolve

R topics documented:

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Description

This function performs BLUP in a mixed model. If not provided, the function will estimate the variance components using REML.

Usage

```
BLUPsolve(
   y,
   X = NULL,
   FixedEff = NULL,
   Z = NULL,
   VarMat = NULL,
   R = NULL,
   VarComp = NULL,
   TestGroup = NULL,
   verbose = TRUE
)
```

Arguments

У	The response variable.
Χ	A matrix of fixed effects.
FixedEff	The effects of the fixed effects, if previously estimated or provided.
Z	The design matrix (or list of matrices) of the random effects. If all design matri-
	ces are an identity, this argument does not need to be provided.

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VarMat The variance-covariance matrix (or list of matrices) of the random effects.

R A variance-covariance matrix for the residuals, if they are not independent and identically distributed (i.e. if R != In).

VarComp

The variance components of each random effect, including the residuals. If VarComp=NULL, the function will estimate these variances using REML.

TestGroup

The index of observations from a test population. This will make the function perform prediction over these observations.

verbose

Logical: should steps of the co,putation be printed on screen?

Value

FixedEff The estimated fixed effects.

VarComp The estimated variance components for all random effects, and for the residual.

RandomEff The predicted random effects and their reliabilities.

modelFit A summary of the fit from the linear mixed model.

modelPred A summary of the prediction from the linear mixed model (when TestGroup is provided).

See Also

REMLsolve

```
## SINGLE COMPONENT WITH FIXED EFFECTS ##
# simulate genotypes
M \leftarrow simGeno(n=500,AlleleFreq=runif(100,0.05,0.5))
# simulate phenotypes
pheno <- simPheno(M,0.6,Vy=10)</pre>
# add fixed effects
X <- cbind(rep(1,nrow(M)),runif(nrow(M)),rbinom(nrow(M),1,0.5))</pre>
b < -c(5,-1,2)
y <- pheno$y + as.numeric(X%*%b)</pre>
# REML
G \leftarrow mkGRM(M)
tmp <- REMLsolve(y=y,X=X,VarMat=G)</pre>
tmp
# BLUP
aux <- BLUPsolve(y=y,X=X,VarMat=G,VarComp=tmp$VarComp)</pre>
str(aux)
# BLUP estimating variance components
aux <- BLUPsolve(y=y,X=X,VarMat=G)</pre>
str(aux)
## TWO VARIANCE COMPONENTS ##
# simulate genotypes
M <- simGeno(n=500,AlleleFreq=runif(100,0.05,0.5))
# simulate phenotypes
pheno <- simPheno(list(M[,1:70],M[,71:100]),c(0.4,0.2),Vy=10)
# BLUP using GRMs
G <- list(mkGRM(M[,1:70]),mkGRM(M[,71:100]))</pre>
aux <- BLUPsolve(y=pheno$y,VarMat=G)</pre>
str(aux)
```

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```
# prediction
aux <- BLUPsolve(y=pheno$y,VarMat=G,TestGroup=401:500)
str(aux)
# BLUP using genotype matrices
aux <- BLUPsolve(y=pheno$y,VarMat=list(diag(1,70),diag(1,30)),Z=list(M[,1:70],M[,71:100]))
str(aux)
# prediction
Vlist <- list(diag(1,70),diag(1,30))
Zlist <- list(M[,1:70],M[,71:100])
aux <- BLUPsolve(y=pheno$y,VarMat=Vlist,Z=Zlist,TestGroup=401:500)
str(aux)</pre>
```

ConfIntMeanBoot

Bootstrap confidence interval

Description

This function calculates the bootstrap confidence interval for the mean of a set of data.

Usage

```
ConfIntMeanBoot(x, levels = NULL, conf = 0.95, n.rep = 10^4)
```

The confidence level of the interval.

Arguments

A vector containing values to calculate confidence interval.
 levels A vector containing the levels (factor) from the data, to calculate separate confidence intervals.
 n.rep The number of bootstrap replicates.

Value

р

A data frame with the confidence interals.

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ConfIntMeanNorm	Confidence	interval for noi	rmal distributed dat	a

Description

This function calculates the confidence interval for the mean of a set of normal distributed data.

```
ConfIntMeanNorm(x, levels = NULL, p = 0.95)
```

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Arguments

x A vector containing values to calculate confidence interval.

levels A vector containing the levels (factor) from the data, to calculate separate confi-

dence intervals.

p The confidence level of the interval.

Value

A data frame with the confidence interals.

dMP

The Marchenko-Pastur density

Description

Density function of the Marchenko-Pastur distribution with rate parameter equal to a and variance equal to s2.

Usage

```
dMP(x, a, s2)
```

Arguments

x A vectors of quantiles.

a Rate parameter (n/m = #observations/#variables).

s2 Variance parameter.

Value

The density.

Gkappa	Kappa statistics
--------	------------------

Description

This function calculates the kappa statistics, to compare two variance-covariance matrices (G1 and G2) defined for the same group of individuals. Kappa is computed using the eigen-decomposition of the matrices, G=ULU' in which U and L are the eigen-vectors and eigen-values, respectively. \code{Kmat = U2'(G1)U2, and kappa=diag(Kmat). Finally, to compare G1 and G2, it is enough to plot lambda against kappa. The more linear their relationship, the more equivalent matrices G1 and G2 are, in terms of variance structure.

```
Gkappa(G1, G2 = NULL, U2 = NULL, L2 = NULL, Kmat = FALSE)
```

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Arguments

G1	The Reference matrix.
G2	The matrix which we want to compare with G1.
U2	The eigen-vectors of $G1$. If not provided, the function will obtain them internally.
L2	The eigen-values of G2. If not provided, the function will obtain them internally.
Kmat	Should the entire Kappa matrix be returned? Default is Kmat=FALSE.

Value

lambda The eigen-values of G2.

kappa The kappa statistics that compare G2 to G1.

Kmat The whole Kappa matrix.

```
# simulate uncorrelated genotypes
M <- simGeno(n=100,AlleleFreq=rep(0.5,200))
# three GRMs from the simulated genotypes
# the first with 100 SNPs
G1 \leftarrow mkGRM(M[,1:100])
# the second with 100 SNPs that don't overlap those from the first
G2 <- mkGRM(M[,101:200])
# the third with all SNPs (100 SNPs overlap with those from the first matrix)
G3 <- mkGRM(M)
#-----#
# compare GRMs using kappa
# verify how that is different from directly comparing the values
# verify how that is also different from comparing the eigen-values
##########
# G1 x G2 #
##########
par(mar=c(4.5,4.5,1,1))
# plot G1 x G2
plot(c(diag(G1),G1[upper.tri(G1)]),c(diag(G2),G2[upper.tri(G2)]),xlab="G1",ylab="G2")
abline(0,1,lty=2,col=2)
# plot eigen-values(G1) x eigen-values(G2)
\verb|plot(eigen(G1)$values,eigen(G2)$values,xlab="eigen-values(G1)",ylab="eigen-values(G2)")|
abline(0,1,lty=2,col=2)
# plot eigen-values(G2)xkappa
tmp <- Gkappa(G1,G2)</pre>
plot(tmp$lambda,tmp$kappa,xlim=range(tmp),ylim=range(tmp),xlab="lambda",ylab="kappa")
abline(0,1,lty=2,col=2)
###########
# G1 x G3 #
###########
par(mar=c(4.5,4.5,1,1))
# plot G1xG3
plot(c(diag(G1),G1[upper.tri(G1)]),c(diag(G3),G3[upper.tri(G3)]),xlab="G1",ylab="G3")
abline(0,1,lty=2,col=2)
# plot eigen-values(G1)xeigen-values(G3)
plot(eigen(G1)$values,eigen(G3)$values,xlab="eigen-values(G1)",ylab="eigen-values(G3)")
abline(0,1,lty=2,col=2)
```

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```
# plot eigen-values(G3)xkappa
tmp <- Gkappa(G1,G3)</pre>
plot(tmp$lambda,tmp$kappa,xlim=range(tmp),ylim=range(tmp),xlab="lambda",ylab="kappa")
abline(0,1,lty=2,col=2)
# simulate correlated genotypes
M <- simGeno(n=100,AlleleFreq=rep(0.5,1000),LD=TRUE,phased=FALSE)
# three GRMs from the simulated genotypes
# the first with 500 SNPs
G1 <- mkGRM(M[,1:500])
# the second with 500 SNPs that don't overlap those from the first
G2 <- mkGRM(M[,501:1000])
# the third with all SNPs (500 SNPs overlap with those from the first matrix)
G3 \leftarrow mkGRM(M)
            _____#
# compare GRMs using kappa
# verify how that is different from directly comparing the values
# verify how that is also different from comparing the eigen-values
##########
# G1 x G2 #
###########
par(mfrow=c(2,2), mar=c(4.5,4.5,1,1))
# plot G1xG2
plot(c(diag(G1),G1[upper.tri(G1)]),c(diag(G2),G2[upper.tri(G2)]),xlab="G1",ylab="G2")
abline(0,1,lty=2,col=2)
# plot eigen-values(G1)xeigen-values(G2)
plot(eigen(G1)$values,eigen(G2)$values,xlab="lambda(G1)",ylab="lambda(G2)")
abline(0,1,1ty=2,col=2)
# plot eigen-values(G2)xkappa
tmp <- Gkappa(G1,G2)</pre>
plot(tmp$lambda,tmp$kappa,xlim=range(tmp),ylim=range(tmp),xlab="lambda",ylab="kappa")
abline(0,1,lty=2,col=2)
plot(NA,xlim=c(0,1),ylim=c(0,1),xlab="",ylab="",axes=FALSE)
##########
# G1 x G3 #
###########
par(mfrow=c(2,2), mar=c(4.5,4.5,1,1))
# plot G1xG3
plot(c(diag(G1),G1[upper.tri(G1)]),c(diag(G3),G3[upper.tri(G3)]),xlab="G1",ylab="G3")
abline(0,1,lty=2,col=2)
# plot eigen-values(G1)xeigen-values(G3)
plot(eigen(G1)$values,eigen(G3)$values,xlab="lambda(G1)",ylab="lambda(G3)")
abline(0,1,lty=2,col=2)
# plot eigen-values(G3)xkappa
tmp <- Gkappa(G1,G3)</pre>
plot(tmp$lambda,tmp$kappa,xlim=range(tmp),ylim=range(tmp),xlab="lambda",ylab="kappa")
abline(0,1,lty=2,col=2)
plot(NA,xlim=c(0,1),ylim=c(0,1),xlab="",ylab="",axes=FALSE)
```

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Description

This function calculates the expected limit of prediction accuracy, using the relationship between two populations based on the singular-value decomposition of their genotype matrices. This limit is obtained as in Cuyabano and Gondro, 2020 (in prep), and the two populations must have the SNP-genotypes on the same set of SNPs.

Usage

```
LimPredAcc(a, b, n, h2, p = 0.95)
```

Arguments

a	The intercept obtained with popCor.
b	The slope obtained with popCor.
n	The size of the test population (the population for which predictions are to be made).
h2	The heritability of the trait.
р	The confidence level of the interval for the expected limit of prediction accuracy.

Value

A data frame with the mean expected limit of prediction accuracy, as well as the boundaries of its confidence interval.

Examples

```
# simulate genotypes
M <- simGeno(n=500, AlleleFreq=rep(0.5, 100))
\# separate two populations of individuals
# note that the populations do not need to have the same size
M1 \leftarrow M[1:300,]
M2 \leftarrow M[301:500,]
tmp <- popCor(M1,M2)</pre>
LimPredAcc(a=tmp$a,b=tmp$b,n=200,h2=0.5)
LimPredAcc(a=tmp$a,b=tmp$b,n=200,h2=seg(0.1,0.9,0.1))
# compare the expected limits of prediction accuracy to the classic sqrt(h2)
par(mar=c(4.5,4.5,1,1))
curve(sqrt(x),xlim=c(0,1),ylim=c(0,1),xlab=expression(h^2),ylab="Limit of prediction accuracy")
aux <- LimPredAcc(a=tmpa,b=tmp$b,n=200,h2=seq(0.01,0.99,0.01))
lines(seq(0.01, 0.99, 0.01), aux$mean, lty=2)
lines(seq(0.01,0.99,0.01),aux$CI_lower,lty=3,col=2)
lines(seq(0.01,0.99,0.01),aux$CI_upper,lty=3,col=2)
legend("topleft", lty=c(1,2,3), col=c(1,1,2), legend=c("sqrt(h2)", "R_expected", "R_CI"))
```

MatrixInv

Inverse of a matrix

Description

This function inverts any matrix, using the best method for that specific matrix.

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Usage

```
MatrixInv(A, use.eigen = FALSE, U = NULL, L = NULL)
```

Arguments

A A matrix.

use.eigen A logical indicating if the eigen-decomposition should be used to obtain the

inverse. Valid only for symmetric matrices.

U A matrix of the eigenvectors (optional).

A vector with the eigenvalues (optional).

Value

The inverse of the matrix.

mkGRM Genomic relationship matrix

Description

This function builds the genomic relationship matrix based on SNP-genotypes.

Usage

```
mkGRM(M, method = "VanRaden")
```

Arguments

M A SNP-genotype matrix (unphased).

method The method to use for standardizing the SNP-genotypes. Deafult is method="VanRaden",

and alternatively can be set $\mathtt{method="Individual"}$, which will standardize each

SNP individually.

Value

The genomic relationship matrix.

```
M <- simGeno(5,rep(0.5,1000))
M <- M[,apply(M,2,var) != 0]
mkGRM(M)
mkGRM(M,method="Individual")</pre>
```

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mkPED Pedigree relationship matrix	
------------------------------------	--

Description

This function builds the pedigree relationship matrix based on a population info in the same format as the generated by function simPopInfo. If you are generating the population info using simPopInfo, the pedigree relationship matrix can be created directly from simPopInfo by setting the parameter pedigree=TRUE.

Usage

```
mkPED(pop.info)
```

Arguments

pop.info

The population info, in the same format generated by simPopInfo.

Value

The pedigree relationship matrix.

Examples

```
mkPED(simPopInfo(10,0))
```

plotMP

Plot eigen-values comparatively with The Marchenko-Pastur distribution

Description

Plot eigen-values comparatively with The Marchenko-Pastur distribution

Usage

```
plotMP(L, a, s2, f = "cumulative")
```

Arguments

L	Α	vectors	of	eigen-va	lues.
---	---	---------	----	----------	-------

- a Rate parameter (n/m = #observations/#variables).
- s2 Variance parameter.
- f The type of function to be plotted, "density" or "cumulative" (default).

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pMP

The Marchenko-Pastur cumulative distribution

Description

Cumulative distribution function of the Marchenko-Pastur distribution with rate parameter equal to a and variance equal to s2.

Usage

```
pMP(x, a, s2)
```

Arguments

x A vectors of quantiles.

a Rate parameter (n/m = #observations/#variables).

s2 Variance parameter.

Value

The cumulative probability distribution.

popBreed

Simulates genotypes for generations of individuals

Description

This function outputs SNP-genotypes for generations of individuals, based on the given SNPs for generation zero (founder population) and on the population info in the same format generated by simPopInfo.

Usage

```
popBreed(M, pop.info)
```

Arguments

M A phased SNP matrix for generation zero.

pop.info The population info, in the same format generated by simPopInfo.

Value

A matrix with the SNPs for all individuals.

See Also

```
simGeno simPopInfo
```

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Examples

```
M <- simGeno(10,rep(0.5,3),phased=TRUE)
pop.info <- simPopInfo(c(10,15),1)
popBreed(M,pop.info)</pre>
```

popCor

Calculates the relationship between two populations

Description

This function calculates the relationship between two populations based on the singular-value decomposition of their genotype matrices. The relationship is obtained as in Cuyabano and Gondro, 2020 (in prep.), and the two populations must have the SNP-genotypes on the same set of SNPs.

Usage

```
popCor(M1, M2)
```

Arguments

M1 The SNP-genotypes of population one.
M2 The SNP-genotypes of population two.

Value

The parameters of the linear regression (intercept and slope) on the components built from the SVDs.

Examples

```
# simulate genotypes
M <- simGeno(n=500,AlleleFreq=rep(0.5,100),LD=TRUE)
# separate two populations of individuals
# note that the populations do not need to have the same size
M1 <- M[1:300,]
M2 <- M[301:500,]
popCor(M1,M2)</pre>
```

REMLsolve

REML estimates of variance components

Description

This function estimates the variance components and the fixed effects in a mixed model, using the restricted maximum likelihood (REML).

```
REMLsolve(y, X = NULL, Z = NULL, VarMat = NULL, R = NULL)
```

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Arguments

У	The response variable.
Χ	A matrix of fixed effects.
Z	The design matrix (or list of matrices) of the random effects. If all design matrices are an identity, this argument does not need to be provided.
VarMat	The variance-covariance matrix (or list of matrices) of the random effects.
R	A variance-covariance matrix for the residuals, if they are not independent and identically distributed (i.e. R != In)

Value

FixedEff The estimated fixed effects.

VarComp The estimated variance components for all random effects, and for the residual.

Examples

```
## SINGLE COMPONENT WITH FIXED EFFECTS ##
# simulate genotypes
M \leftarrow simGeno(n=500,AlleleFreq=runif(100,0.05,0.5))
# simulate phenotypes
pheno <- simPheno(M, 0.6, Vy=10)
# add fixed effects
X <- cbind(rep(1,nrow(M)),runif(nrow(M)),rbinom(nrow(M),1,0.5))</pre>
b < c(5,-1,2)
y <- pheno$y + as.numeric(X%*%b)</pre>
# REML
G \leftarrow mkGRM(M)
REMLsolve(y=y,X=X,VarMat=G)
## TWO VARIANCE COMPONENTS ##
# simulate genotypes
M \leftarrow simGeno(n=500,AlleleFreq=runif(100,0.05,0.5))
# simulate phenotypes
pheno <- simPheno(list(M[,1:70],M[,71:100]),c(0.4,0.2),Vy=10)\\
# add fixed effects
X <- cbind(rep(1,nrow(M)),runif(nrow(M)),rbinom(nrow(M),1,0.5))</pre>
b < c(5,-1,2)
y <- pheno$y + as.numeric(X%*%b)</pre>
# REML
G \leftarrow list(mkGRM(M[,1:70]),mkGRM(M[,71:100]))
REMLsolve(y=y,X=X,VarMat=G)
REMLsolve(y=y, X=X, VarMat=list(diag(1,70), diag(1,30)), Z=list(M[,1:70], M[,71:100]))
tmp < - REMLsolve(y=y, X=X, VarMat=list(diag(1,70), diag(1,30)), Z=list(M[,1:70], M[,71:100])) \\ VarComp(x=y, X=X, VarMat=list(diag(1,70), diag(1,30)), Z=list(M[,1:70], M[,1:70], M[,1:7
c(sum(apply(M[,1:70],2,var)),sum(apply(M[,71:100],2,var)),1)*tmp
```

simGeno

Simulate genotypes

Description

This function simulates numeric SNP-genotypes.

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Usage

```
simGeno(n, AlleleFreq, LD = FALSE, R = NULL, phased = FALSE)
```

Arguments

n	The number of individuals to simulate.
AlleleFreq	A vector with the required allele frequencies for each SNP, with length(AlleleFreq) being the number of SNPs.
LD	Logical indicating if SNP-genotypes should be simulated in linkage disequilibrium.
R	The correlation matrix for the linkage disequilibrium between SNPs. If R=NULL, a default matrix with 0.95^(lag between SNPs) will be generated.
phased	Lagical indicating if phased genotypes should be returned.

Value

The simulated SNPs, as a single matrix with allele counts (phased=FALSE), or as a list of matrices with t0,1 at each allele (phased=TRUE).

Examples

```
\label{eq:simGeno} \begin{split} & simGeno(n=10,AlleleFreq=rep(0.5,5)) \\ & simGeno(n=10,AlleleFreq=rep(0.5,5),phased=TRUE) \\ & simGeno(n=10,AlleleFreq=rep(0.5,5),LD=TRUE) \\ & R_LD <- diag(1,5) \\ & R_LD[abs(col(R_LD)-row(R_LD)) == 1] <- 0.95 \\ & R_LD[abs(col(R_LD)-row(R_LD)) == 2] <- 0.85 \\ & R_LD[abs(col(R_LD)-row(R_LD)) == 3] <- 0.75 \\ & R_LD[abs(col(R_LD)-row(R_LD)) == 4] <- 0.65 \\ & simGeno(n=10,AlleleFreq=rep(0.5,5),LD=TRUE,R=R_LD) \\ & geno <- simGeno(n=1000,AlleleFreq=rep(0.5,5),LD=TRUE,R=R_LD,phased=TRUE) \\ & str(geno) \\ & cor(rbind(geno[[1]],geno[[2]])) \end{split}
```

simPheno Simulate phenotypes

Description

This function simulates phenotypes (continuous or binary) based on the quantitative trait loci (QTL), only with the genetic values and residuals (y=g+e).

```
simPheno(M, h2, Vy = 1, binary = FALSE, p = NULL, sim.error = 1e-04)
```

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Arguments

М	A matrix or a list of matrices of the (unphased) SNP-genotypes set to be the QTL.
h2	The heritability.
Vy	The phenotypic variance (for continuous traits). Default is Vy=1.
binary	$\label{logical} \mbox{Logical indicating whether simulated phenotypes should be binary. Default is $$ \mbox{binary=FALSE}.$
p	If binary=TRUE, the probability of observing 1's in the phenotypes, with default p=0.5.
sim.error	An error threshold to be considered on the orthogonalization of genetic values and residuals.

Value

y The simulated phenotypes.

a The simulated QTL effects.

g The simulated genetic values.

e The simulated residuals.

h2 The simulated heritability.

See Also

simGeno

```
## general examples ##
AF <- runif(500,0.05,1)
geno <- simGeno(n=1000,AlleleFreq=AF,LD=TRUE)</pre>
pheno <- simPheno(M=geno,h2=0.6,Vy=10)</pre>
str(pheno)
pheno <- simPheno(M=geno,h2=0.6,binary=TRUE)</pre>
str(pheno)
mean(pheno$y) # this is the realized probability of observing y = 1
pheno <- simPheno(M=geno, h2=0.6, binary=TRUE, p=0.2)</pre>
str(pheno)
mean(phenoy) # this is the realized probability of observing y = 1
## example with multiple components with different heritabilities ##
AF <- runif(500,0.05,1)
geno <- simGeno(n=1000,AlleleFreq=AF,LD=TRUE)</pre>
pheno <- simPheno(M=list(geno[,1:350],geno[,351:500]),h2=c(0.2,0.4),Vy=10)
str(pheno)
pheno <- simPheno(M=list(geno[,1:350],geno[,351:500]),h2=c(0.2,0.4),Vy=10,binary=TRUE,p=0.3)
mean(phenoy) # this is the realized probability of observing y = 1
```

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simPopInfo Simulate a population info	
---------------------------------------	--

Description

This function simulates a population info, with individuals' ID, sex, generation and parent information.

Usage

```
simPopInfo(
   n,
   n.gen = 0,
   p.male = 0.5,
   pop.info = NULL,
   pedigree = FALSE,
   pre.id = ""
)
```

Arguments

n	The size of each generation simulated. If a single number is provided, all generations will be of the same size.
n.gen	The number of generations to simulate after generation zero. The generation counter starts at zero, being that the founder population. By default, n.gen=0, and if n.gen is set to any positive number, e.g. n.gen=1, that means that the info on the founder generation and on one generation after that will be generated.
p.male	The proportion of male individuals in the population.
pop.info	The information on a base population on which the code should build new generations.
pedigree	Should the pedigree be generated? Default is pedigree=FALSE.
pre.id	An ID label to paste at the begining of the default ID generated by the function.

Value

A data frame with the population info, and the pedigree when required.

```
info The population info.ped The pedigree matrix.
```

```
simPopInfo(10,0)
simPopInfo(10,2)
simPopInfo(3,1,pedigree=TRUE)
simPopInfo(10,2,0.25)
simPopInfo(c(5,10,15),2)
```

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