Package 'randPedPCA'

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Title Fast PCA for Large Pedigrees
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Description Carry out principal component analysis (PCA) of very large pedigrees such as found in breeding populations! This package, 'randPedPCA', exploits sparse matrices and randomised linear algebra to deliver a gazillion-times speedup compared to naive singular value decoposition (SVD) (and eigen decomposition).
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dspc

Add downsampling index to rppca object

Description

This index is used by plot.rppca to downsample the col (colour) values. It is stored in the rppca object's ds slot.

Usage

```
dspc(pc, to = 10000)
```

Arguments

pc an object of class rppca

to The down-sampling parameter. A numeric > 0 or a vector or NA. Interpreted as

a proportion or integer or a index vector, see details.

Details

The parameter to is used to specify and possibly which individuals are sampled. If NA, all individuals are retained. If to is of length one and is between 0 and 1, then it is interpreted as a proportion. If it is greater than 1, it is taken to be the number of individuals to be sampled (possibly rounded by sample.int). If to is a logical or an integer vector, it is used for logical or integer indexing, respectively. The integer indices of the sample individuals are written to the ds slot. If ds exists, it is overwritten with a warning.

Value

An (invisible) object of class rppca with a slot ds added.

getNumVectorsHutchinson

Compute the number of vectors to use for Hutchinson trace estimation

Description

Follows Skorski, M. (2021). Modern Analysis of Hutchinson's Trace Estimator. 2021 55th Annual Conference on Information Sciences and Systems (CISS), 1–5. https://doi.org/10.1109/CISS50987.2021.9400306

Usage

```
getNumVectorsHutchinson(e, d)
```

Arguments

e A numeric denoting the relative error margin

d A numeric. 1-d is the probability the the relative error is bounded by e.

Value

a scalar

hutchpp

Hutch++ trace estimation

Description

Hutch++ trace estimation

Usage

```
hutchpp(
   B,
   num_queries = 10,
   sketch_frac = 2/3,
   center = FALSE,
   oraculum = oraculumLi
)
```

Arguments

B An object related to the matrix A for which the trace is to be estimated

sketch_frac Hutch++ detail

center Whether or not to implicitly centre

oraculum The oracle function to use

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Details

The Hutch++ algorithm (Meyer et al. 2021, https://doi.org/10.48550/arXiv.2010.09649) estimates the trace of a matrix A by evaluating matrix vector products of A and (sub-gaussian) random vectors. This is used on a matrix B which is related to A through some function. The oracle function has to be chosen so that oracle(B, G) returns the product A the oracle function is set to work on a pedigree's L inverse matrix. But this implementation is general and should work - given a custom oracle function - on other input too.

In the context of pedigree PCA, this is used to estimate the trace of an (implicitly) centred additive relationship matrix.

There logical parameter center allows for a pedigree's L matrix to be (implicitly) centred. This is important because centring changes the total variance of the data and thus the trace of A.

Value

An estimate of A's trace - numeric

Examples

```
hutchpp(pedLInv)
hutchpp(pedLInv, center=TRUE)
```

importLinv

Generate spam object from L inverse file

Description

RandPedPCA relies in the spam onject format. But matrices are commonly stored in other formats.

Usage

```
importLinv(pth)
```

Arguments

pth

path to matrix market file for L inverse matrix in dgTMatrix format

Value

A spam sparse matrix

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pedLInv

Example pedigree L inverse matrix

Description

An L inverse matrix generated from an AlphaSimR simulation of 20 generations. An original population splits into sub-populations A and B. After a number of generations, crossbreeding starts.

Usage

pedLInv

Format

'pedLInv' Matrix object of class 'spam' of dimension 2650x2650, with 7750 (row-wise) nonzero elements. Density of the matrix is 0.11 Class 'spam' (32-bit)

Source

Simulation

pedLInv2

Example pedigree L inverse matrix

Description

An L inverse matrix generated from an AlphaSimR simulation of 20 generations. Two diverged populations A and B. After a number of generations, crossbreeding starts.

Usage

pedLInv2

Format

'pedLInv2' Matrix object of class 'spam' of dimension 2650x2650, with 7750 (row-wise) nonzero elements. Density of the matrix is 0.11 Class 'spam' (32-bit)

Source

Simulation

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pedLInv4

Example pedigree L inverse matrix

Description

An L inverse matrix generated from an AlphaSimR simulation of 20 generations. One population, ABCD, is split into four, A, B, C, D.

Usage

pedLInv4

Format

'pedLInv4' Matrix object of class 'spam' of dimension 4200x4200, with 12200 (row-wise) nonzero elements. Density of the matrix is 0.0692 Class 'spam' (32-bit)

Source

Simulation

pedMeta

Metadata associated with the example pedigree

Description

A dataframe.

Usage

pedMeta

Format

```
## 'pedMeta' A 'data.frame' of 2650 individuals (rows) with 12 variables (cols):
```

id Integer individual ID

population Population code. A, B or AB

generation Generation of the individual

mid dam ID

fid sire ID

gv1 genetic value

pv1 phenotypic value

gv2 genetic value

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```
pv2 phenotypic valuegv genetic valuepv phenotypic valuegenerationPlotShift for plotting
```

Source

Simulation

pedMeta2

#' Example genotype data #' #' An genotype matrix (0/1/2) generated from an AlphaSimR simulation of 20 generations. #' An original population splits into sub-populations A and B. After a number of #' generations, crossbreeding starts. #' #' #' @format ## 'pedGeno' #' Matrix of 2650 rows (inidividuals) and 11000 columns (SNPs) #' @source Simulation "pedGeno" Metadata associated with the 2nd example pedigree

Description

A dataframe.

Usage

pedMeta2

Format

```
## 'pedMeta2' A 'data.frame' of 2650 individuals (rows) with 12 variables (cols):

id Integer individual ID

population Population code. A, B or AB

generation Generation of the individual

mid dam ID

fid sire ID

gv1 genetic value

pv1 phenotypic value

gv2 genetic value

pv2 phenotypic value

gv genetic value
```

Source

Simulation

pv phenotypic value

generationPlotShift for plotting

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pedMeta4

#' Example genotype data 2 #' #' An genotype matrix (0/1/2) generated from an AlphaSimR simulation of 20 generations. #' Two diverged populations A and B. After a number of #' generations, crossbreeding starts. #' #' #' @format ## 'pedGeno2' #' Matrix of 2650 rows (inidividuals) and 11000 columns (SNPs) #' @source Simulation "pedGeno2" Metadata associated with the 2nd example pedigree

Description

A dataframe.

Usage

pedMeta4

Format

```
## 'pedMeta4' A 'data.frame' of 4200 individuals (rows) with 9 variables (cols):
```

id Integer individual ID

population Population code. A, B, C, D, or ABCD

generation Generation of the individual

mid dam ID

fid sire ID

gv1 genetic value

pv1 phenotypic value

gv2 genetic value

pv2 phenotypic value

Source

Simulation

plot3D

_	
plot3D	3D plot using rgl

Description

A simple wrapper around rgl's pot3d function.

Usage

```
plot3D(x, dims = c(1, 2, 3), xlab = NULL, ylab = NULL, zlab = NULL, ...)
```

Arguments

X	an rppca object
dims	vector of length 3 - indices of the PCs to plot
xlab	(optional) x axis label
ylab	(optional) yaxis label
zlab	(optional) xz axis label
	additional arguments passed to rgl::plot3d

Details

Note, different to plot.rppca, which is relatively slow, plot3D does not down-sample the principal components and it ignores the ds slot of an rppca object if present.

Value

No return value, called for its side effects.

Examples

```
pc <- rppca(pedLInv)
plot3D(pc)

ped <- pedigree(sire=pedMeta$fid, dam=pedMeta$mid, label=pedMeta$id)
pc2 <- rppca(ped)
plot3D(pc2)</pre>
```

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plot3DWithProj

3D plot of PC scores with projections on coordinate planes

Description

3D plot of PC scores with projections on coordinate planes

Usage

```
plot3DWithProj(
   pc,
   dims = c(1, 2, 3),
   plotProj = TRUE,
   grid = TRUE,
   col = 1,
   ff = 0.5,
   theta = -45,
   phi = 25
)
```

Arguments

```
pc An rppca object

dims integer vector, which PCs to plot, defauts to 1:3

plotProj logical, whether to plot the projections

grid logical, wheter to plot grids

col the dot colours, integer or string, scalar or vector

ff numeric, offset for projection (proportion of the orthogonal axis's range)

theta, phi polar coordinates in degrees. theta rotates round the vertical axis. phi rotates round the horizontal axis.
```

Value

nothing

Examples

```
ped <- pedigree(pedMeta$fid,
pedMeta$mid,
pedMeta$id
)
pc <- rppca(ped)
plot3DWithProj(pc, col=as.numeric(factor(pedMeta$population)))</pre>
```

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randRangeFinder	Generate range matrix for SVD

Description

Generate range matrix for SVD

Usage

```
randRangeFinder(L, rank, depth, numVectors, cent = FALSE)
```

Arguments

L a pedigree's L inverse matrix in sparse 'spam' format
rank An integer, how many principal components to return
depth integer, number of iterations for generating the range matrix
numVectors An integer > rank, to specify the oversampling for the
cent logical whether or not to (implicitly) 'centre' the additive relationship matrix,

or more precisely, its underlying 'data matrix' L

Value

The range matrix for randSVD

randSVD	Singular value decomposition in sparse triangular matrix

Description

Uses randomised linear algebra, see Halko et al. (2010). Singular value decomposition (SVD) decomposes a matrix $X=U\Sigma W^T$

Usage

```
randSVD(L, rank, depth, numVectors, cent = FALSE)
```

Arguments

L	a pedigree's L inverse matrix in sparse 'spam' format
rank	An integer, how many principal components to return

depth integer, the number of iterations for generating the range matrix

numVectors An integer > rank to specify the oversampling for the

cent logical, whether or not to (implicitly) centre the additive relationship matrix

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Value

```
A list of three: u = U, d = Sigma, and v = W^T
```

randTraceHutchinson

Trace estimation for sparse L inverse matrices

Description

Using Hutchinson's method

Usage

```
randTraceHutchinson(L, numVectors)
```

Arguments

L A pedigree's L inverse matrix

numVectors an integer specifying how many random vectors to use

If you do not have a good reason to do otherwise, use the function hutchpp

instead.

The higher numVectors, the higher the accuracy and the longer the running time. Accuracy can be estimated with the function getNumVectorsHutchinson.

Value

a scalar

rppca

Fast pedigree PCA using sparse matrices and randomised linear algebra

Description

Fast pedigree PCA using sparse matrices and randomised linear algebra

Usage

```
rppca(X, ...)
## S3 method for class 'spam'
rppca(
    X,
    method = "randSVD",
    rank = 10,
    depth = 3,
```

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```
numVectors = 15,
totVar = NULL,
center = FALSE,
...
)

## S3 method for class 'pedigree'
rppca(
    X,
    method = "randSVD",
    rank = 10,
    depth = 3,
    numVectors = 15,
    totVar = NULL,
    center = FALSE,
...
)
```

Arguments

Χ	A representation of a	pedigree, see Details.

... optional arguments passed to methods

method string only randSVD (the default) is implemented rank integer how many principal components to return

depth integer number of iterations for generating the range matrix numVectors integer > rank to specify the oversampling for the range matrix

totVar scalar (optional) the total variance, required for computation of variance pro-

portions when using an L-inverse matrix a input

center logical whether or not to (implicitly) centre the additive relationship matrix

Details

The output slots are named like those of R's built in prcomp function. Rotation is not returned by default as it is the transpose of the PC scores, which are returned in x. scale and center are set to FALSE.

Value

A list containing:

x the principal components

sdev the variance components of each PC. Note that the total variance is not known per se and this these components cannot be used to compute the proportion of the total variance accounted for by each PC. However, if nVecTraceEst is specified, rppca will estimate the total variance and return variance proportions.

vProp the estimated variance proportions accounted for by each PC. Only returned if totVar is set.

sparse2spam

```
scale always FALSE
```

center logical indicating whether or not the implicit data matrix was centred

rotation the right singular values of the relationship matrix. Only returned if returnRotation == TRUE

varProps proportion of the total variance explained by each PC. Only returned if starting from a pedigree object without centring, or if totVar is supplied.

Examples

sparse2spam

Convert generic sparse matrix to spam format

Description

Convert generic sparse matrix to spam format

Usage

```
sparse2spam(sprs)
```

Arguments

sprs

A sparse matrix.

Value

A spam sparse matrix

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