

Package ‘Correplot’

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Type Package

Title A Collection of Functions for Graphing Correlation Matrices

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Author Jan Graffelman [aut, cre],
Jan De Leeuw [aut]

Maintainer Jan Graffelman <jan.graffelman@upc.edu>

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Imports corrplot, xtable, MASS, lsei, ggplot2

Description Routines for the graphical representation of correlation matrices by means of correlograms, MDS maps and biplots obtained by PCA, PFA or WALS (weighted alternating least squares); See Graffelman & De Leeuw (2023) <[doi:10.1080/00031305.2023.2186952](https://doi.org/10.1080/00031305.2023.2186952)>.

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aircraft

Characteristics of aircraft

Description

Four variables registered for 21 types of aircraft.

Usage

```
data("aircraft")
```

Format

A data frame with 21 observations on the following 4 variables.

SPR specific power

RGF flight range factor

PLF payload

SLF sustained load factor

Source

Gower and Hand, Table 2.1

References

Gower, J.C. and Hand, D.J. (1996) *Biplots*, Chapman & Hall, London

Examples

```
data(aircraft)
str(aircraft)
```

aircraftR

Correlations between characteristics of aircraft

Description

Correlations between SPR (specific power), RGF (flight range factor), PLF (payload) and SLF (sustained load factor) for 21 types of aircraft.

Usage

```
data(aircraftR)
```

Format

a matrix containing the correlations

Source

Gower and Hand, Table 2.1

References

Gower, J.C. and Hand, D.J. (1996) *Biplots*, Chapman & Hall, London

angleToR *Convert angles to correlations.*

Description

Function `angleToR` converts a vector of angles (in radians) to an estimate of the correlation matrix, given an interpretation function.

Usage

```
angleToR(x, ifun = "cos")
```

Arguments

`x` a vector of angles (in radians)
`ifun` the interpretation function ("cos" or "lincos")

Value

A correlation matrix

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

References

Graffelman, J. (2012) Linear-angle correlation plots: new graphs for revealing correlation structure. *Journal of Computational and Graphical Statistics*. 22(1): 92-106.

See Also

[cos](#), [lincos](#)

Examples

```
angles <- c(0, pi/3)
R <- angleToR(angles)
print(R)
```

artificialR *Correlations for 10 generated variables*

Description

A 10 by 10 artificial correlation matrix

Usage

```
data(artificialR)
```

Format

A matrix of correlations

Source

Trosset (2005), Table 1.

References

Trosset, M.W. (2005) Visualizing correlation. *Journal of Computational and Graphical Statistics*, 14(1), pp. 1–19.

athletesR *Correlation matrix of characteristics of Australian athletes*

Description

Correlation matrix of 12 characteristics of Austration athletes (Sex, Height, Weight, Lean Body Mass, RCC, WCC, Hc, Hg, Ferr, BMI, SSF, Bfat)

Usage

```
data(athletesR)
```

Format

A matrix of correlations

Source

Weisberg (2005), file ais.txt

References

Weisberg, S. (2005) *Applied Linear Regression*. Third edition, John Wiley & Sons, New Jersey.

banknotes

Swiss banknote data

Description

The Swiss banknote data consist of six measures taken on 200 banknotes, of which 100 are counterfeits, and 100 are normal.

Usage

```
data("banknotes")
```

Format

A data frame with 200 observations on the following 7 variables.

Length Banknote length

Left Left width

Right Right width

Bottom Bottom margin

Top Top margin

Diagonal Length of the diagonal of the image

Counterfeit 0 = normal, 1 = counterfeit

References

Weisberg, S. (2005) Applied Linear Regression. Third edition. John Wiley & Sons, New Jersey.

Examples

```
data(banknotes)
```

berkeleyR

Correlation matrix for boys of the Berkeley Guidance Study

Description

Correlation matrix for sex, height and weight at age 2, 9 and 18 and somatotype

Usage

```
data(berkeleyR)
```

Format

A matrix of correlations

Source

Weisberg (2005), file BGSBoys.txt

References

Weisberg, S. (2005) *Applied Linear Regression*. Third edition, John Wiley & Sons, New Jersey.

cathedralsR

Correlation matrix for height and length

Description

Correlation between nave height and total length

Usage

`data(cathedralsR)`

Format

A matrix of correlations

Source

Weisberg (2005), file cathedral.txt

References

Weisberg, S. (2005) *Applied Linear Regression*. Third edition, John Wiley & Sons, New Jersey.

correlogram	<i>Plot a correlogram</i>
-------------	---------------------------

Description

correlogram plots a correlogram for a correlation matrix.

Usage

```
correlogram(R, labs=colnames(R), ifun="cos", cex=1, main="", ntrials=50,  
            xlim=c(-1.2, 1.2), ylim=c(-1.2, 1.2), pos=NULL, ...)
```

Arguments

R	a correlation matrix.
labs	a vector of labels for the variables.
ifun	the interpretation function ("cos" or "lincos")
cex	character expansion factor for the variable labels
main	a title for the correlogram
ntrials	number of starting points for the optimization routine
xlim	limits for the x axis (e.g. c(-1.2,1.2))
ylim	limits for the y axis (e.g. c(-1.2,1.2))
pos	if specified, overrules the calculated label positions for the variables.
...	additional arguments for the plot function.

Details

correlogram makes a correlogram on the basis of a set of angles. All angles are given w.r.t the positive x-axis. Variables are represented by unit vectors emanating from the origin.

Value

A vector of angles

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

References

Trosset, M.W. (2005) Visualizing correlation. *Journal of Computational and Graphical Statistics* 14(1), pp. 1–19

See Also

[fit_angles](#), [nlminb](#)

Examples

```
X <- matrix(rnorm(90),ncol=3)
R <- cor(X)
angles <- correlogram(R)
```

countriesR

Correlations between educational and demographic variables

Description

Correlations between infant mortality, educational and demographic variables (infd, phys, dens, agds, lit, hied, gnp)

Usage

```
data(countriesR)
```

Format

A matrix of correlations

Source

Chatterjee and Hadi (1988)

References

Chatterjee, S. and Hadi, A.S. (1988), *Sensitivity Analysis in Regression*. Wiley, New York.

FitRDeltaQSym

Approximation of a correlation matrix with column adjustment and symmetric low rank factorization

Description

Program FitRDeltaQSym calculates a low rank factorization for a correlation matrix. It adjusts for column effects, and the approximation is therefore asymmetric.

Usage

```
FitRDeltaQSym(R, W = NULL, nd = 2, eps = 1e-10, delta = 0, q = colMeans(R),
              itmax.inner = 1000, itmax.outer = 1000, verbose = FALSE)
```

Arguments

R	A correlation matrix
W	A weight matrix (optional)
nd	The rank of the low rank approximation
eps	The convergence criterion
delta	Initial value for the scalar adjustment (zero by default)
q	Initial values for the column adjustments (random by default)
itmax.inner	Maximum number of iterations for the inner loop of the algorithm
itmax.outer	Maximum number of iterations for the outer loop of the algorithm
verbose	Print information or not

Details

Program `FitRDeltaQSym` implements an iterative algorithm for the low rank factorization of the correlation matrix. It decomposes the correlation matrix as $R = \delta J + 1 q' + G G' + E$. The approximation of R is ultimately asymmetric, but the low rank factorization used for biplotting ($G G'$) is symmetric.

Value

A list object with fields:

delta	The final scalar adjustment
Rhat	The final approximation to the correlation matrix
C	The matrix of biplot vectors
rmse	The root mean squared error
q	The final column adjustments

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

References

Graffelman, J. and De Leeuw, J. (2023) Improved approximation and visualization of the correlation matrix. *The American Statistician* pp. 1–20. Available online as latest article [doi:10.1080/00031305.2023.2186952](https://doi.org/10.1080/00031305.2023.2186952)

See Also

[wAddPCA](#), [ipSymLS](#), [Keller](#)

Examples

```

data(HeartAttack)
X <- HeartAttack[,1:7]
X[,7] <- log(X[,7])
colnames(X)[7] <- "logPR"
R <- cor(X)
W <- matrix(1, 7, 7)
diag(W) <- 0
out.sym <- FitRDeltaQSym(R, W, eps=1e-6)
Rhat <- out.sym$Rhat

```

FitRwithPCAandWALS	<i>Calculate a low-rank approximation to the correlation matrix with four methods</i>
--------------------	---------------------------------------------------------------------------------------

Description

Function `FitRwithPCAandWALS` uses principal component analysis (PCA) and weighted alternating least squares (WALS) to calculate different low-rank approximations to the correlation matrix.

Usage

```
FitRwithPCAandWALS(R, nd = 2, itmaxout = 10000, itmaxin = 10000, eps = 1e-08)
```

Arguments

R	The correlation matrix
nd	The dimensionality of the low-rank solution (2 by default)
itmaxout	Maximum number of iterations for the outer loop of the algorithm
itmaxin	Maximum number of iterations for the inner loop of the algorithm
eps	Numerical criterion for convergence of the outer loop

Details

Four methods are run successively: standard PCA; PCA with an additive adjustment; WALS avoiding the fit of the diagonal; WALS avoiding the fit of the diagonal and with an additive adjustment.

Value

A list object with fields:

Rhat.pca	Low-rank approximation obtained by PCA
Rhat.pca.adj	Low-rank approximation obtained by PCA with adjustment
Rhat.wals	Low-rank approximation obtained by WALS without fitting the diagonal
Rhat.wals.adj	Low-rank approximation obtained by WALS without fitting the diagonal and with adjustment

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

References

Graffelman, J. and De Leeuw, J. (2023) Improved approximation and visualization of the correlation matrix. *The American Statistician* pp. 1–20. Available online as latest article [doi:10.1080/00031305.2023.2186952](https://doi.org/10.1080/00031305.2023.2186952)

See Also

[wAddPCA](#)

Examples

```
data(HeartAttack)
X <- HeartAttack[,1:7]
X[,7] <- log(X[,7])
colnames(X)[7] <- "logPR"
R <- cor(X)
## Not run:
out <- FitRwithPCAandWALS(R)

## End(Not run)
```

fit_angles

Fit angles to a correlation matrix

Description

fit_angles finds a set of optimal angles for representing a particular correlation matrix by angles between vectors

Usage

```
fit_angles(R, ifun = "cos", ntrials = 10, verbose = FALSE)
```

Arguments

R	a correlation matrix.
ifun	an angle interpretation function (cosine, by default).
ntrials	number of trials for optimization routine nlmnb
verbose	be silent (FALSE), or produce more output (TRUE)

Value

a vector of angles (in radians)

Author(s)

anonymous

References

Trosset, M.W. (2005) Visualizing correlation. *Journal of Computational and Graphical Statistics* 14(1), pp. 1–19

See Also[nlminb](#)**Examples**

```
X <- matrix(rnorm(90),ncol=3)
R <- cor(X)
angles <- fit_angles(R)
print(angles)
```

fysiologyR

Correlations between thirteen physiological variables

Description

Correlations of 13 physiological variables (sys, dia, p.p., pul, cort, u.v., tot/100, adr/100, nor/100, adr/tot, tot/hr, adr/hr, nor/hr) obtained from 48 medical students

Usage

```
data(fysiologyR)
```

Format

A matrix of correlations

Source

Hills (1969), Table 1.

References

Hills, M (1969) On looking at large correlation matrices *Biometrika* 56(2): pp. 249.

ggbplot

*Create a biplot with ggplot2***Description**

Function ggbiplot creates a biplot of a matrix with ggplot2 graphics.

Usage

```
ggbplot(A, B, main = "", circle = TRUE, xlab = "", ylab = "", main.size = 8,
        xlim = c(-1, 1), ylim = c(-1, 1), rowcolor = "red", rowch = 1, colcolor = "blue",
        colch = 1, rowarrow = FALSE, colarrow = TRUE)
```

Arguments

A	A dataframe with coordinates and names for the biplot row markers
B	A dataframe with coordinates and names for the biplot column markers
main	A title for the biplot
circle	Draw a unit circle (circle=TRUE) or not (circle=FALSE)
xlab	The label for the x axis
ylab	The label for the y axis
main.size	Size of the main title
xlim	Limits for the horizontal axis
ylim	Limits for the vertical axis
rowcolor	Color used for the row markers
rowch	Symbol used for the row markers
colcolor	Color used for the column markers
colch	Symbol used for the column markers
rowarrow	Draw arrows from the origin to the row markers (rowarrow=TRUE) or not
colarrow	Draw arrows from the origin to the column markers (colarrow=TRUE) or not

Details

Dataframes A and B must consists of three columns labeled "PA1", "PA2" (coordinates of the first and second principal axis) and a column "strings" with the labels for the coordinates.

Dataframe B is optional. If it is not specified, a biplot with a single set of markers is constructed, for which the row settings must be specified.

Value

A ggplot2 object

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

References

Graffelman, J. and De Leeuw, J. (2023) On the visualisation of the correlation matrix. Available online. [doi:10.48550/arXiv.2211.13150](https://doi.org/10.48550/arXiv.2211.13150)

See Also

[bplot](#), [ggtally](#), [biplot](#)

Examples

```
data("HeartAttack")
X <- as.matrix(HeartAttack[,1:7])
n <- nrow(X)
Xt <- scale(X)/sqrt(n-1)
res.svd <- svd(Xt)
Fs <- sqrt(n)*res.svd$u # standardized principal components
Gp <- crossprod(t(res.svd$v),diag(res.svd$d)) # biplot coordinates for variables
rows.df <- data.frame(Fs[,1:2],as.character(1:n))
colnames(rows.df) <- c("PA1","PA2","strings")
cols.df <- data.frame(Gp[,1:2],colnames(X))
colnames(cols.df) <- c("PA1","PA2","strings")
ggbplot(rows.df,cols.df,xlab="PA1",ylab="PA2",main="PCA")
```

ggcorrelogram

Create a correlogram as a ggplot object.

Description

Function ggcorrelogram creates a correlogram of a correlation matrix using ggplot graphics.

Usage

```
ggcorrelogram(R, labs = colnames(R), ifun = "cos", cex = 1, main = "", ntrials = 50,
              xlim = c(-1.2, 1.2), ylim = c(-1.2, 1.2), hjust = 1, vjust = 2, size = 2,
              main.size = 8)
```

Arguments

R	a correlation matrix
labs	a vector of labels for the variables
ifun	the interpretation function ("cos" or "lincos")
cex	character expansion factor for the variable labels
main	a title for the correlogram

<code>ntrials</code>	number of starting points for the optimization routine
<code>xlim</code>	limits for the x axis (e.g. <code>c(-1.2,1.2)</code>)
<code>ylim</code>	limits for the y axis (e.g. <code>c(-1.2,1.2)</code>)
<code>hjust</code>	horizontal adjustment of variable labels (by default 1 for all variables)
<code>vjust</code>	vertical adjustment of variable labels (by default 2 for all variables)
<code>size</code>	font size for the labels of the variables
<code>main.size</code>	font size of the main title of the correlogram

Details

`ggcorrelogram` makes a correlogram on the basis of a set of angles. All angles are given w.r.t the positive x-axis. Variables are represented by unit vectors emanating from the origin.

Value

A `ggplot` object. Field `theta` of the output contains the angles for the variables.

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

References

Trosset, M.W. (2005) Visualizing correlation. *Journal of Computational and Graphical Statistics* 14(1), pp. 1–19

See Also

[correlogram](#), [fit_angles](#), [nlminb](#)

Examples

```
set.seed(123)
X <- matrix(rnorm(90), ncol=3)
R <- cor(X)
angles <- ggcorrelogram(R)
```

`ggtally`

Create a correlation tally stick on a biplot vector

Description

Function `ggtally` puts a series of dots along a biplot vector of a correlation matrix, so marking the change in correlation along the vector with specified values.

Usage

```
ggtally(G, p1, adj = 0, values = seq(-1, 1, by = 0.2), dotsize = 0.1, dotcolour = "black")
```

Arguments

G	A matrix (or vector) of biplot markers
p1	A ggplot2 object with a biplot
adj	A scalar adjustment for the correlations
values	Values of the correlations to be marked off by dots
dotsize	Size of the dot
dotcolour	Colour of the dot

Details

Any set of values for the correlation to be marked off can be used, though a standard scale with 0.2 increments is recommended.

Value

A ggplot2 object with the updated biplot

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

References

Graffelman, J. and De Leeuw, J. (2023) On the visualisation of the correlation matrix. Available online. [doi:10.48550/arXiv.2211.13150](https://doi.org/10.48550/arXiv.2211.13150)

See Also

[ggbplot](#)

Examples

```
library(calibrate)
data(goblets)
R <- cor(goblets)
out.sd <- eigen(R)
V <- out.sd$vectors[,1:2]
D1 <- diag(out.sd$values[1:2])
Gp <- crossprod(t(V),sqrt(D1))
pca.df <- data.frame(Gp)
pca.df$strings <- colnames(R)
colnames(pca.df) <- c("PA1", "PA2", "strings")
p1 <- ggbplot(pca.df, pca.df, main="PCA correlation biplot", xlab="", ylab="", rowarrow=TRUE,
              rowcolor="blue", rowch="", colch="")
p1 <- ggtally(Gp, p1, values=seq(-0.2, 0.6, by=0.2), dotsize=0.1)
```

gobletsR

Correlations between size measurements of archeological goblets

Description

Correlations between 6 size measurements of archeological goblets

Usage

```
data(gobletsR)
```

Format

A matrix of correlations

Source

Manly (1989)

References

Manly, B.F.J. (1989) *Multivariate statistical methods: a primer*. Chapman and Hall, London.

HeartAttack

Myocardial infarction or Heart attack data

Description

Data set consisting of 101 observations of patients who suffered a heart attack.

Usage

```
data("HeartAttack")
```

Format

A data frame with 101 observations on the following 8 variables.

Pulse Pulse

CI Cardiac index

SI Systolic index

DBP Diastolic blood pressure

PA Pulmonary artery pressure

VP Ventricular pressure

PR Pulmonary resistance

Status Deceased or survived

Source

Table 18.1, (Saporta 1990, pp. 452–454)

References

Saporta, G. (1990) Probabilites analyse des donnees et statistique. Paris, Editions technip

Examples

```
data(HeartAttack)
str(HeartAttack)
```

ipSymLS	<i>Function for obtaining a weighted least squares low-rank approximation of a symmetric matrix</i>
---------	-----------------------------------------------------------------------------------------------------

Description

Function ipSymLS implements an alternating least squares algorithm that uses both decomposition and block relaxation to find the optimal positive semidefinite approximation of given rank p to a known symmetric matrix of order n .

Usage

```
ipSymLS(target, w = matrix(1, dim(target)[1], dim(target)[2]), ndim = 2,
        init = FALSE, itmax = 100, eps = 1e-06, verbose = FALSE)
```

Arguments

target	Symmetric matrix to be approximated
w	Matrix of weights
ndim	Number of dimensions extracted (2 by default)
init	Initial value for the solution (optional; if supplied should be a matrix of dimensions <code>nrow(target)</code> by <code>ndim</code>)
itmax	Maximum number of iterations
eps	Tolerance criterion for convergence
verbose	Show the iteration history (<code>verbose=TRUE</code>) or not (<code>verbose=FALSE</code>)

Value

A matrix with the coordinates for the variables

Author(s)

deleeuw@stat.ucla.edu

References

- De Leeuw, J. (2006) A decomposition method for weighted least squares low-rank approximation of symmetric matrices. Department of Statistics, UCLA. Retrieved from <https://escholarship.org/uc/item/1wh197mh>
- Graffelman, J. and De Leeuw, J. (2023) Improved approximation and visualization of the correlation matrix. The American Statistician pp. 1–20. Available online as latest article [doi:10.1080/00031305.2023.2186952](https://doi.org/10.1080/00031305.2023.2186952)

Examples

```
data(banknotes)
R <- cor(banknotes)
W <- matrix(1,nrow(R),nrow(R))
diag(W) <- 0
Fp.als <- ipSymLS(R,w=W,verbose=TRUE,eps=1e-15)
Rhat.als <- Fp.als%*%t(Fp.als)
```

jointlim

Establish limits for x and y axis

Description

jointlim computes a sensible range for x and y axis if two sets of points are to be plotted simultaneously

Usage

```
jointlim(X, Y)
```

Arguments

X	Matrix of coordinates
Y	Matrix of coordinates

Value

xlim	minimum and maximum for x-range
ylim	minimum and maximum for y-range

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

Examples

```
X <- matrix(runif(20),ncol=2)
Y <- matrix(runif(20),ncol=2)
print(jointlim(X,Y)$xlim)
```

Keller	<i>Program Keller calculates a rank p approximation to a correlation matrix according to Keller's method.</i>
--------	----------------------------------------------------------------------------------------------------------------------------

Description

Keller's method is based on iterated eigenvalue decompositions that are used to adjust the diagonal of the correlation matrix.

Usage

```
Keller(R, eps = 1e-06, nd = 2, itmax = 10)
```

Arguments

R	A correlation matrix
eps	Numerical criterion for convergence (default eps=1e-06)
nd	Number of dimensions used in the spectral decomposition (default nd=2)
itmax	The maximum number of iterations

Value

A matrix containing the approximation to the correlation matrix-

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

References

Keller, J.B. (1962) Factorization of Matrices by Least-Squares. *Biometrika*, 49(1 and 2) pp. 239–242.

Graffelman, J. and De Leeuw, J. (2023) Improved approximation and visualization of the correlation matrix. *The American Statistician* pp. 1–20. Available online as latest article [doi:10.1080/00031305.2023.2186952](https://doi.org/10.1080/00031305.2023.2186952)

See Also

[ipSymLS](#)

Examples

```
data(Kernels)
R <- cor(Kernels)
Rhat <- Keller(R)
```

Kernels

Wheat kernel data

Description

Wheat kernel data set taken from the UCI Machine Learning Repository

Usage

```
data("Kernels")
```

Format

A data frame with 210 observations on the following 8 variables.

area Area of the kernel

perimeter Perimeter of the kernel

compactness Compactness ($C = 4 \cdot \pi \cdot A / P^2$)

length Length of the kernel

width Width of the kernel

asymmetry Asymmetry coefficient

groove Length of the groove of the kernel

variety Variety (1=Kama, 2=Rosa, 3=Canadian)

Source

<https://archive.ics.uci.edu/ml/datasets/seeds>

References

M. Charytanowicz, J. Niewczas, P. Kulczycki, P.A. Kowalski, S. Lukasik, S. Zak, A Complete Gradient Clustering Algorithm for Features Analysis of X-ray Images. in: Information Technologies in Biomedicine, Ewa Pietka, Jacek Kawa (eds.), Springer-Verlag, Berlin-Heidelberg, 2010, pp. 15-24.

Examples

```
data(Kernels)
```

linangplot	<i>Linang plot</i>
------------	--------------------

Description

linangplot produces a plot of two variables, such that the correlation between the two variables is linear in the angle.

Usage

```
linangplot(x, y, tmx = NULL, tmy = NULL, ...)
```

Arguments

x	x variable
y	y variable
tmx	vector of tickmarks for the x variable
tmy	vector of tickmarks for the y variable
...	additional arguments for the plot routine

Value

Xt	coordinates of the points
B	axes for the plot
r	correlation coefficient
angledegrees	angle between axes in degrees
angleradians	angle between axes in radians
r	correlation coefficient

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

See Also

[plotcorrelogram](#)

Examples

```
x <- runif(10)
y <- rnorm(10)
linangplot(x,y)
```

lincos	<i>Linearized cosine function</i>
--------	-----------------------------------

Description

Function `lincos` linearizes the cosine function over the interval $[0, 2\pi]$. The function returns $-2/\pi * x + 1$ over $[0, \pi]$ and $2/\pi * x - 3$ over $[\pi, 2\pi]$

Usage

```
lincos(x)
```

Arguments

`x` angle in radians

Value

a real number in $[-1, 1]$.

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

References

Graffelman, J. (2012) Linear-angle correlation plots: new graphs for revealing correlation structure. *Journal of Computational and Graphical Statistics*. 22(1): 92-106.

See Also

[cos](#)

Examples

```
angle <- pi
y <- lincos(angle)
print(y)
```

pco

Principal Coordinate Analysis

Description

pco is a program for Principal Coordinate Analysis.

Usage

pco(Dis)

Arguments

Dis A distance or dissimilarity matrix

Details

The program pco does a principal coordinates analysis of a dissimilarity (or distance) matrix (Dij) where the diagonal elements, Dii, are zero.

Note that when we dispose of a similarity matrix rather than a distance matrix, a transformation is needed before calling coorprincipal. For instance, if Sij is a similarity matrix, Dij might be obtained as $D_{ij} = 1 - S_{ij}/\text{diag}(S_{ij})$

Goodness of fit calculations need to be revised such as to deal (in different ways) with negative eigenvalues.

Value

PC	the principal coordinates
Dl	all eigenvalues of the solution
Dk	the positive eigenvalues of the solution
B	double centred matrix for the eigenvalue decomposition
decom	the goodness of fit table

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

See Also

[cmdscale](#)

Examples

```

citynames <- c("Aberystwyth", "Brighton", "Carlisle", "Dover", "Exeter", "Glasgow", "Hull",
              "Inverness", "Leeds", "London", "Newcastle", "Norwich")
A <-matrix(c(
0,244,218,284,197,312,215,469,166,212,253,270,
244,0,350,77,167,444,221,583,242,53,325,168,
218,350,0,369,347,94,150,251,116,298,57,284,
284,77,369,0,242,463,236,598,257,72,340,164,
197,167,347,242,0,441,279,598,269,170,359,277,
312,444,94,463,441,0,245,169,210,392,143,378,
215,221,150,236,279,245,0,380,55,168,117,143,
469,583,251,598,598,169,380,0,349,531,264,514,
166,242,116,257,269,210,55,349,0,190,91,173,
212,53,298,72,170,392,168,531,190,0,273,111,
253,325,57,340,359,143,117,264,91,273,0,256,
270,168,284,164,277,378,143,514,173,111,256,0),ncol=12)
rownames(A) <- citynames
colnames(A) <- citynames
out <- pco(A)
plot(out$PC[,2],-out$PC[,1],pch=19,asp=1)
textxy(out$PC[,2],-out$PC[,1],rownames(A))

```

 PearsonLee

Heights of mothers and daughters

Description

Heights of 1375 mothers and daughters (in cm) in the UK in 1893-1898.

Usage

```
data(PearsonLee)
```

Format

dataframe with Mheight and Dheight

Source

Weisberg, Chapter 1

References

Weisberg, S. (2005) *Applied Linear Regression*, John Wiley & Sons, New Jersey

pfa *Principal factor analysis*

Description

Program pfa performs (iterative) principal factor analysis, which is based on the computation of eigenvalues of the reduced correlation matrix.

Usage

```
pfa(X, option = "data", m = 2, initial.communality = "R2", crit = 0.001, verbose = FALSE)
```

Arguments

X	A data matrix or correlation matrix
option	Specifies the type of matrix supplied by argument X. Values for option are data, cor or cov. data is the default.
m	The number of factors to extract (2 by default)
initial.communality	Method for computing initial communalities. Possibilities are R2 or maxcor.
crit	The criterion for convergence. The default is 0.001. A smaller value will require more iterations before convergence is reached.
verbose	When set to TRUE, additional numerical output is shown.

Value

Res	Matrix of residuals
Psi	Diagonal matrix with specific variances
La	Matrix of loadings
Shat	Estimated correlation matrix
Fs	Factor scores

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

References

Mardia, K.V., Kent, J.T. and Bibby, J.M. (1979) Multivariate analysis.

Rencher, A.C. (1995) Methods of multivariate analysis.

Satorra, A. and Neudecker, H. (1998) Least-Squares Approximation of off-Diagonal Elements of a Variance Matrix in the Context of Factor Analysis. *Econometric Theory* 14(1) pp. 156–157.

See Also[princomp](#)**Examples**

```
X <- matrix(rnorm(100),ncol=2)
out.pfa <- pfa(X)
# based on a correlation matrix
R <- cor(X)
out.pfa <- pfa(R,option="cor")
```

proteinR

Correlations between sources of protein

Description

Correlations between sources of protein for a number of countries (Red meat, White meat, Eggs, Milk, Fish, Cereals, Starchy food, Nuts, Fruits and vegetables).

Usage

```
data(proteinR)
```

Format

A matrix of correlations

Source

Manly (1989)

References

Manly, B.F.J. (1989) *Multivariate statistical methods: a primer*. Chapman and Hall, London.

proteinsR	<i>Correlations between sources of protein</i>
-----------	------------------------------------------------

Description

Correlations between sources of protein for a number of countries (Red meat, White meat, Eggs, Milk, Fish, Cereals, Starchy food, Nuts, Fruits and vegetables).

Usage

```
data(proteinR)
```

Format

A matrix of correlations

Source

Manly (1989)

References

Manly, B.F.J. (1989) *Multivariate statistical methods: a primer*. Chapman and Hall, London.

recordsR	<i>Correlations between national track records for men</i>
----------	------------------------------------------------------------

Description

Correlations between national track records for men (100m,200m,400m,800m,1500m,5000m,10.000m and Marathon)

Usage

```
data(recordsR)
```

Format

A matrix of correlations

Source

Johnson and Wichern, Table 8.6

References

Johnson, R.A. and Wichern, D.W. (2002) *Applied Multivariate Statistical Analysis*. Fifth edition. New Jersey: Prentice Hall.

rmse *Calculate the root mean squared error*

Description

Program `rmse` calculates the RMSE for a matrix approximation.

Usage

```
rmse(R, Rhat, W = matrix(1, nrow(R), ncol(R)) - diag(nrow(R)),  
      verbose = FALSE, per.variable = FALSE)
```

Arguments

<code>R</code>	The original matrix
<code>Rhat</code>	The approximating matrix
<code>W</code>	A symmetric matrix of weights
<code>verbose</code>	Print output (<code>verbose=TRUE</code>) or not (<code>verbose=FALSE</code>)
<code>per.variable</code>	Calculate the RMSE for the whole matrix (<code>per.variable=FALSE</code>) or for each variable separately (<code>per.variable=TRUE</code>)

Details

By default, function `rmse` assumes a symmetric correlation matrix as input, together with its approximation. The approximation does not need to be symmetric. Weight matrix `W` has to be symmetric. By default, the diagonal is excluded from RMSE calculations ($W = J - I$). To include it, specify $W = J$, that is set `W = matrix(1, nrow(R), ncol(R))`

Value

the calculated `rmse`

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

References

Graffelman, J. and De Leeuw, J. (2023) Improved approximation and visualization of the correlation matrix. *The American Statistician* pp. 1–20. doi:[10.1080/00031305.2023.2186952](https://doi.org/10.1080/00031305.2023.2186952)

Examples

```

data(banknotes)
X <- as.matrix(banknotes[,1:6])
p <- ncol(X)
J <- matrix(1,p,p)
R <- cor(X)
out.sd <- eigen(R)
V <- out.sd$vectors
D1 <- diag(out.sd$values)
V2 <- V[,1:2]
D2 <- D1[1:2,1:2]
Rhat <- V2*%D2*%t(V2)
rmse(R,Rhat,W=J)

```

rmsePCAandWALS	<i>Generate a table of root mean square error (RMSE) statistics for principal component analysis (PCA) and weighted alternating least squares (WALS).</i>
----------------	-----------------------------------------------------------------------------------------------------------------------------------------------------------

Description

Function `rmsePCAandWALS` creates table with the RMSE for each variable, for a low-rank approximation to the correlation matrix obtained by PCA or WALS.

Usage

```
rmsePCAandWALS(R, output, digits = 4, omit.diagonals = c(FALSE,FALSE,TRUE,TRUE))
```

Arguments

R	The correlation matrix
output	A list object with four approximationst to the correlation matrix
digits	The number of digits used in the output
omit.diagonals	Vector of four logicals for omitting the diagonal of the correlation matrix for RMSE calculations. Defaults to <code>c(FALSE,FALSE,TRUE,TRUE)</code> , to include the diagonal for PCA and exclude it for WALS

Value

A matrix with one row per variable and four columns for RMSE statistics.

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

References

Graffelman, J. and De Leeuw, J. (2023) Improved approximation and visualization of the correlation matrix. *The American Statistician* pp. 1–20. doi:10.1080/00031305.2023.2186952

See Also

[FitRwithPCAandWALS](#)

Examples

```
data(HeartAttack)
X <- HeartAttack[,1:7]
X[,7] <- log(X[,7])
colnames(X)[7] <- "logPR"
R <- cor(X)
## Not run:
out <- FitRwithPCAandWALS(R)
Results <- rmsePCAandWALS(R,out)

## End(Not run)
```

storksR

Correlations between three variables

Description

Danish data from 1953-1977 giving the correlations between nesting storks, human birth rate and per capita electricity consumption.

Usage

```
data(storksR)
```

Format

A matrix of correlations

Source

Gabriel and Odoroff, Table 1.

References

Gabriel, K. R. and Odoroff, C. L. (1990) Biplots in biomedical research. *Statistics in Medicine* 9(5): pp. 469-485.

students	<i>Marks for 5 student exams</i>
----------	----------------------------------

Description

Matrix of marks for five exams, two with closed books and three with open books (Mechanics (C), Vectors (C), Algebra (O), Analysis (O) and Statistics (O)).

Usage

```
data(students)
```

Format

A data matrix

Source

Mardia et al., Table 1.2.1

References

Mardia, K.V., Kent, J.T. and Bibby, J.M. (1979) *Multivariate Analysis*, Academic Press London.

studentsR	<i>Correlations between marks for 5 exams</i>
-----------	-----------------------------------------------

Description

Correlation matrix of marks for five exams, two with closed books and three with open books (Mechanics (C), Vectors (C), Algebra (O), Analysis (O) and Statistics (O)).

Usage

```
data(studentsR)
```

Format

A matrix of correlations

Source

Mardia et al., Table 1.2.1

References

Mardia, K.V., Kent, J.T. and Bibby, J.M. (1979) *Multivariate Analysis*, Academic Press London.

tally *Create a tally on a biplot vector*

Description

Function `tally` marks of a set of dots on a biplot vector. It is thought for biplot vectors representing correlations, such that their correlation scale becomes visible, without doing a full calibration with tick marks and tick mark labels.

Usage

```
tally(G, adj = 0, values = seq(-1, 1, by = 0.2), pch = 19, dotcolor = "black", cex = 0.5,
      color.negative = "red", color.positive = "blue")
```

Arguments

<code>G</code>	Matrix with biplot coordinates of the variables
<code>adj</code>	A scalar adjustment for the correlations
<code>values</code>	The values of the correlations to be marked off by dots
<code>pch</code>	The character code used for marking off correlations
<code>dotcolor</code>	The colour of the dots that are marked off
<code>cex</code>	The character expansion factor for a dot.
<code>color.negative</code>	The colour of the segments of the negative part of the correlation scale
<code>color.positive</code>	The colour of the segments of the positive part of the correlation scale

Value

NULL

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

References

Graffelman, J. and De Leeuw, J. (2023) Improved approximation and visualization of the correlation matrix. *The American Statistician* pp. 1–20. doi:10.1080/00031305.2023.2186952

See Also

[bplot](#), [calibrate](#)

Examples

```
data(goblets)
R <- cor(goblets)
results <- eigen(R)
V <- results$vectors
Dl <- diag(results$values)
#
# Calculate correlation biplot coordinates
#
G <- crossprod(t(V[,1:2]),sqrt(Dl[1:2,1:2]))
#
# Make the biplot
#
bplot(G,G,rowch=NA,colch=NA,collab=colnames(R),
      xl=c(-1.1,1.1),yl=c(-1.1,1.1))
#
# Create a correlation tally stick for variable X1
#
tally(G[1,])
```

tr

Compute the trace of a matrix

Description

tr computes the trace of a matrix.

Usage

```
tr(X)
```

Arguments

X a (square) matrix

Value

the trace (a scalar)

Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

Examples

```
X <- matrix(runif(25),ncol=5)
print(X)
print(tr(X))
```

wAddPCA *Low-rank matrix approximation by weighted alternating least squares*

Description

Function wAddPCA calculates a weighted least squares approximation of low rank to a given matrix.

Usage

```
wAddPCA(x, w = matrix(1, nrow(x), ncol(x)), p = 2, add = "all", bnd = "opt",
        itmaxout = 1000, itmaxin = 1000, epsout = 1e-06, epsin = 1e-06,
        verboseout = TRUE, verbosein = FALSE)
```

Arguments

x	The data matrix to be approximated
w	The weight matrix
p	The dimensionality of the low-rank solution (2 by default)
add	The additive adjustment to be employed. Can be "all" (default), "nul" (no adjustment), "one" (adjustment by a single scalar), "row" (adjustment by a row) or "col" (adjustment by a column).
bnd	Can be "opt" (default), "all", "row" or "col".
itmaxout	Maximum number of iterations for the outer loop of the algorithm
itmaxin	Maximum number of iterations for the inner loop of the algorithm
epsout	Numerical criterion for convergence of the outer loop
epsin	Numerical criterion for convergence of the inner loop
verboseout	Be verbose on the outer loop iterations
verbosein	Be verbose on the inner loop iterations

Value

A list object with fields:

a	The left matrix (A) of the factorization $X = AB'$
b	The right matrix (B) of the factorization $X = AB'$
z	The product AB'
f	The final value of the loss function
u	Vector for rows used to construct rank 1 weights
v	Vector for columns used to construct rank 1 weights
p	The vector with row adjustments
q	The vector with column adjustments
itel	Iterations needed for convergence
delta	The additive adjustment
y	The low-rank approximation to x

Author(s)

jan@deleeuwpx.net

References

Graffelman, J. and De Leeuw, J. (2023) Improved approximation and visualization of the correlation matrix. *The American Statistician* pp. 1–20. Available online as latest article [doi:10.1080/00031305.2023.2186952](https://doi.org/10.1080/00031305.2023.2186952)

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See Also

[ipSymLS](#)

Examples

```
data(HeartAttack)
X <- HeartAttack[,1:7]
X[,7] <- log(X[,7])
colnames(X)[7] <- "logPR"
R <- cor(X)
W <- matrix(1, 7, 7)
diag(W) <- 0
Wals.out <- wAddPCA(R, W, add = "nul", verboseout = FALSE)
Rhat <- Wals.out$y
```

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