Package 'dae'

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Title Functions Useful in the Design and ANOVA of Experiments

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Depends R (>= 2.10.0), ggplot2

Imports graphics, methods, stats

Description The content falls into the following groupings: (i) Data, (ii) Factor manipulation functions, (iii) Design functions, (iv) ANOVA functions, (v) Matrix functions, (vi) Projector and canonical efficiency functions, and (vii) Miscellaneous functions. A document 'daeDesignRandomization.pdf', available in the doc subdirectory of the installation directory for 'dae', describes the use of the package for generating randomized layouts for experiments. The ANOVA functions facilitate the extraction of information when the 'Error' function has been used in the call to 'aov'.

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URL http://chris.brien.name

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dae-package	*Functions Useful in the Design and ANOVA of Experiments*

Description

The content falls into the following groupings: (i) Data, (ii) Factor manipulation functions, (iii) Design functions, (iv) ANOVA functions, (v) Matrix functions, (vi) Projector and canonical efficiency functions, and (vii) Miscellaneous functions. A document 'daeDesignRandomization.pdf', available in the doc subdirectory of the installation directory for 'dae', describes the use of the package for generating randomized layouts for experiments. The ANOVA functions facilitate the extraction of information when the 'Error' function has been used in the call to 'aov'.

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(i) Data

ABC.Interact.dat	Randomly generated set of values indexed by three factors
Fac4Proc.dat	Data for a 2^4 factorial experiment
Sensory3Phase.dat	Data for the three-pahse sensory evaluation experiment in Brien, C.J. and Payne, R.W. (1999)
SPLGrass.dat	Data for an experiment to investigate the effects of grazing patterns on pasture composition

(ii) Factor manipulation functions

as.numfac	Convert a factor to a numeric vector
fac.combine	Combines several factors into one
fac.divide	Divides a factor into several individual factors
fac.gen	Generate all combinations of several factors

fac.match	Match, for each combination of a set of columns in 'x', the row that has the same combination in 'table'
fac.nested	Creates a factor whose values are generated within those of the factor nesting.fac
fac.recode	Recodes factor 'levels' using possibly nonunique values in a vector.
mpone	Converts the first two levels of a factor into the numeric values -1 and +1

(iii) Design functions

blockboundaryPlot	This function plots a block boundary on a plot produced by 'designPlot'. It supersedes blockboundary.plot.
designLatinSqrSys	Generate a systematic plan for a Latin Square design.
designPlot	This function plots treatments within a matrix. It superseded design.plot.
fac.layout	Generate a randomized layout for an experiment
no.reps	Computes the number of replicates for an experiment
detect.diff	Computes the detectable difference for an experiment
power.exp	Computes the power for an experiment

(iv) ANOVA functions

fitted.aovlist	Extract the fitted values for a fitted model from an aovlist object
fitted.errors	Extract the fitted values for a fitted model
interaction.ABC.plot	Plots an interaction plot for three factors
qqyeffects	Half or full normal plot of Yates effects
resid.errors	Extract the residuals for a fitted model
residuals.aovlist	Extract the residuals from an aovlist object
strength	Generate paper strength values
tukey.1df	Performs Tukey's one-degree-of-freedom-test-for-nonadditivity
yates.effects	Extract Yates effects

(v) Matrix functions

elements	Extract the elements of an array specified by the subscripts
fac.ar1mat	Forms the ar1 correlation matrix for a (generalized) factor
fac.sumop	Computes the summation matrix that produces sums corresponding to a (generalized) factor
fac.vcmat	Forms the variance matrix for the variance component of a (generalized) factor
mat.I	Forms a unit matrix
mat.J	Forms a square matrix of ones
mat.ar1	Forms an ar1 correlation matrix
mat.ar2	Forms an ar2 correlation matrix
mat.ar3	Forms an ar3 correlation matrix

mat.sar2	Forms an sar2 correlation matrix
mat.exp	Forms an exponential correlation matrix
mat.banded	Forms a banded matrix
mat.dirprod	Forms the direct product of two matrices
mat.dirsum	Forms the direct sum of a list of matrices
mat.Vpred	Forms the variance matrix of predictions

(vi) Projector and canonical efficiency functions

Projector class:

projector	Create projectors
projector-class	Class projector
is.projector	Tests whether an object is a valid object of class projector
print.projector	Print projectors
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Accepts two or more formulae:

projs.canon	A canonical analysis of the relationships between two or more sets of projectors
summary.pcanon	A summary of the results of an analysis of the relationships between two or more sets of projectors
print.summary.pcanon	Prints the values in an 'summary.pcanon' object
efficiencies.pcanon	Extracts the canonical efficiency factors from a list of class 'pcanon'

Accepts exactly two formulae:

projs.2canon	A canonical analysis of the relationships between two sets of projectors
projs.combine.p2canon	Extract, from a p2canon object, the projectors
summary.p2canon	A summary of the results of an analysis of the relationships between two sets of projectors
print.summary.p2canon	Prints the values in an 'summary.p2canon' object that give the combined decomposition
efficiencies.p2canon	Extracts the canonical efficiency factors from a list of class 'p2canon'

Others:

Ameasures	Calculates the A-optimality measures from the variance matrix for predictions
decomp.relate	Examines the relationship between the eigenvectors for two decompositions
efficiency.criteria	Computes efficiency criteria from a set of efficiency factors

fac.meanop	Computes the projection matrix that produces means
proj2.eigen	Canonical efficiency factors and eigenvectors in joint decomposition of two projectors
proj2.efficiency	Computes the canonical efficiency factors for the joint decomposition of two projectors
proj2.combine	Compute the projection and Residual operators for two, possibly nonorthogonal, projectors
projs.structure	Orthogonalised projectors for the terms in a formula
show-methods	Methods for Function 'show' in Package dae

(vii) Miscellaneous functions

extab	Expands the values in table to a vector
get.daeTolerance	Gets the value of daeTolerance for the package dae
harmonic.mean	Calcuates the harmonic mean.
is.allzero	Tests whether all elements are approximately zero
rmvnorm	Generates a vector of random values from a multivariate normal distribution
set.daeTolerance	Sets the value of daeTolerance for the package dae

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ABC.Interact.dat   *Randomly generated set of values indexed by three factors*

Description

This data set has randomly generated values of the response variable MOE (Measure Of Effectiveness) which is indexed by the two-level factors A, B and C.

Usage

```
data(ABC.Interact.dat)
```

Format

A data.frame containing 8 observations of 4 variables.

Source

Generated by Chris Brien

Ameasures	*Calculates the A-optimality measures from the variance matrix for predictions*

Description

Calculates the A-optimality measures, possibly for different subgroups of the predictions, from the variance matrix for the predictions. The predictions will be sequentially broken into groups of the size specified by the elements of groupsizes. The A-optimality measures are calculated for the differences between predictions within each group and for those between predictions from different groups.

Usage

```
Ameasures(Vpred, groupsizes = NULL)
```

Arguments

Vpred   The variance matrix of the predictions.

groupsizes   A numeric containing group sizes. The sum of the elements of groupsizes must equal the order of Vpred. If NULL, the average for all pairwise differences is obtained.

Value

A matrix containing the within and between group A-optimality measures.

References

Smith, A. B., D. G. Butler, C. R. Cavanagh and B. R. Cullis (2015). Multi-phase variety trials using both composite and individual replicate samples: a model-based design approach. *Journal of Agricultural Science*, **153**, 1017-1029.

See Also

mat.Vpred, projs.canon.

Examples

```
## Reduced example from Smith et al. (2015)
## Generate two-phase design
mill.fac <- fac.gen(list(Mrep = 2, Mday = 2, Mord = 3))
field.lay <- fac.gen(list(Frep = 2, Fplot = 4))
field.lay$Variety <- factor(c("D","E","Y","W","G","D","E","M"),
                            levels = c("Y","W","G","M","D","E"))
start.design <- cbind(mill.fac, field.lay[c(3,4,5,8,1,7,3,4,5,8,6,2),])
rownames(start.design) <- NULL

## Set up matrices
W <- model.matrix(~ -1+ Variety, start.design)
ng <- ncol(W)
Gg<- diag(1, ng)
Vu <- with(start.design, fac.vcmat(Mrep, 0.3) +
```

```
                           fac.vcmat(fac.combine(list(Mrep, Mday)), 0.2) +
                           fac.vcmat(Frep, 0.1) +
                           fac.vcmat(fac.combine(list(Frep, Fplot)), 0.2))
R <- diag(1, nrow(start.design))

## Caluclate information matrix
Vp <- mat.Vpred(W = W, Gg = Gg, Vu = Vu, R = R)

## Calculate A-optimality measure
Ameasures(Vp)
```

---

as.numfac   *Convert a factor to a numeric vector*

Description

Converts a factor to a numeric vector with approximately the numeric values of its levels.
Hence, the levels of the factor must be numeric values, stored as characters. It uses the method
described in factor. Use as.numeric to convert a factor to a numeric vector with integers 1, 2,
... corresponding to the positions in the list of levels. You can also use fac.recode to recode the
levels to numeric values. If a numeric is supplied, it is left unchanged.

Usage

```
as.numfac(factor)
```

Arguments

factor   The factor to be converted.

Value

A numeric vector. An NA will be stored for any value of the factor whose level is not a number.

Author(s)

Chris Brien

See Also

as.numeric, fac.recode in package **dae**, factor.

Examples

```
## set up a factor and convert it to a numeric vector
a <- factor(rep(1:3, 4))
x <- as.numfac(a)
```

blockboundary.plot	*This function plots a block boundary on a plot produced by* design.plot. *It is being superseded by* blockboundaryPlot.

Description

This function plots a block boundary on a plot produced by design.plot. It allows control of the starting unit, through rstart and cstart, and the number of rows (nr) and columns (nc) from the starting unit that the blocks to be plotted are to cover. However, it is being supersed by blockboundaryPlot, which has more descriptive argument names.

Usage

```
blockboundary.plot(bdef = NULL, bseq = FALSE, rstart= 0, cstart = 0,
                   nr, nc, bcol = 1, bwd = 2)
```

Arguments

bdef
A matrix of block sizes:

- if there is only one row, then the first element is interpreted as the no. rows in each block and blocks with this number of rows are to be repeated across the rows of the design.
- if there is more than one row, then each row of the matrix specifies a block, with the sequence of rows in the matrix specifying a corresponding sequence of blocks down the rows of the design.

Similarly, a single value for a column specifies a repetition of blocks of that size across the columns of the design, while several column values specifies a sequence of blocks across the columns of the size specified.

bseq
A logical thatt determines whether block numbers are repetitions or sequences of block numbers.

rstart
A numeric speccifying the row after which the plotting of block boundaries is to start.

cstart
A numeric speccifying the column after which the plotting of block boundaries is to start.

nr
A numeric the number of rows (nr), from the starting unit, that the blocks to be plotted are to cover.

nc
A numeric the number of columns (nc), from the starting unit, that the blocks to be plotted are to cover.

bcol
A character string specifying the colour of the block boundary.
See Colour specification under the par function.

bwd
A numeric giving the width of the block boundary to be plotted.

Value

no values are returned, but modifications are made to the currently active plot.

See Also

design.plot, par, DiGGer

Examples

```
## Not run:
    SPL.Lines.mat <- matrix(as.numfac(Lines), ncol=16, byrow=T)
    colnames(SPL.Lines.mat) <- 1:16
    rownames(SPL.Lines.mat) <- 1:10
    SPL.Lines.mat <- SPL.Lines.mat[10:1, 1:16]
    windows()
    design.plot(SPL.Lines.mat,trts=1:10,new=TRUE,
                rstr="Rows",cstr="Columns", chtdiv=3, rprop = 1,cprop=1,
                plotbndry = TRUE)
    #Plot Mainplot boundaries
    blockboundary.plot(bdef = cbind(4,16), rstart = 1, bwd = 3, bcol = "green",
                       nr = 9, nc = 16)
    blockboundary.plot(bdef = cbind(1,4), bwd = 3, bcol = "green", nr = 1, nc = 16)
    blockboundary.plot(bdef = cbind(1,4), rstart= 9, bwd = 3, bcol = "green",
                       nr = 10, nc = 16)
    #Plot all 4 block boundaries
    blockboundary.plot(bdef = cbind(8,5,5,4), bseq=T, cstart = 1, rstart= 1,
                       bwd = 3,bcol = "blue", nr = 9, nc = 15)
    blockboundary.plot(bdef = cbind(10,16), bwd=3,bcol="blue", nr=10, nc=16)
    #Plot border and internal block boundaries only
    blockboundary.plot(bdef = cbind(8,14), cstart = 1, rstart= 1,
                       bwd = 3, bcol = "blue", nr = 9, nc =  15)
    blockboundary.plot(bdef = cbind(10,16), bwd = 3, bcol = "blue",
                       nr = 10, nc = 16)
## End(Not run)
```

---

blockboundaryPlot	*This function plots a block boundary on a plot produced by* [designPlot]*.*

Description

This function plots a block boundary on a plot produced by [designPlot]. It allows control of the starting unit, through rstart and cstart, and the number of rows (nrows) and columns (ncolumns) from the starting unit that the blocks to be plotted are to cover.

Usage

```
blockboundaryPlot(blockdefinition = NULL, blocksequence = FALSE,
                  rstart= 0, cstart = 0, nrows, ncolumns,
                  blocklinecolour = 1, blocklinewidth = 2)
```

Arguments

blockdefinition

A [matrix] of block sizes:

- if there is only one row, then the first element is interpreted as the no. rows in each block and blocks with this number of rows are to be repeated across the rows of the design.
- if there is more than one row, then each row of the matrix specifies a block, with the sequence of rows in the matrix specifying a corresponding sequence of blocks down the rows of the design.

Similarly, a single value for a column specifies a repetition of blocks of that size across the columns of the design, while several column values specifies a sequence of blocks across the columns of the size specified.

blocksequence   A [logical](#) thatt determines whether block numbers are repetitions or sequences of block numbers.

rstart   A [numeric](#) speccifying the row after which the plotting of block boundaries is to start.

cstart   A [numeric](#) speccifying the column after which the plotting of block boundaries is to start.

nrows   A [numeric](#) the number of rows (nrows), from the starting unit, that the blocks to be plotted are to cover.

ncolumns   A [numeric](#) the number of columns (ncolumns), from the starting unit, that the blocks to be plotted are to cover.

blocklinecolour

A [character](#) string specifying the colour of the block boundary.

See Colour specification under the [par](#) function.

blocklinewidth   A [numeric](#) giving the width of the block boundary to be plotted.

Value

no values are returned, but modifications are made to the currently active plot.

See Also

[designPlot](#), [par](#), DiGGer

Examples

```
## Not run:
    SPL.Lines.mat <- matrix(as.numfac(Lines), ncol=16, byrow=T)
    colnames(SPL.Lines.mat) <- 1:16
    rownames(SPL.Lines.mat) <- 1:10
    SPL.Lines.mat <- SPL.Lines.mat[10:1, 1:16]
    designPlot(SPL.Lines.mat, labels=1:10, new=TRUE,
               rtitle="Rows",ctitle="Columns",
               chardivisor=3, rcellpropn = 1, ccellpropn=1,
               plotcellboundary = TRUE)
    #Plot Mainplot boundaries
    blockboundaryPlot(blockdefinition = cbind(4,16), rstart = 1,
                      blocklinewidth = 3, blockcolour = "green",
                      nrows = 9, ncolumns = 16)
    blockboundaryPlot(blockdefinition = cbind(1,4),
                      blocklinewidth = 3, blockcolour = "green",
                      nrows = 1, ncolumns = 16)
  blockboundaryPlot(blockdefinition = cbind(1,4), rstart= 9, nrows = 10, ncolumns = 16,
                      blocklinewidth = 3, blockcolour = "green")
    #Plot all 4 block boundaries
    blockboundaryPlot(blockdefinition = cbind(8,5,5,4), blocksequence=T,
                      cstart = 1, rstart= 1, nrows = 9, ncolumns = 15,
                      blocklinewidth = 3,blockcolour = "blue")
  blockboundaryPlot(blockdefinition = cbind(10,16), blocklinewidth=3, blockcolour="blue",
                      nrows=10, ncolumns=16)
    #Plot border and internal block boundaries only
```

```
    blockboundaryPlot(blockdefinition = cbind(8,14), cstart = 1, rstart= 1,
                      nrows = 9, ncolumns =  15,
                      blocklinewidth = 3, blockcolour = "blue")
    blockboundaryPlot(blockdefinition = cbind(10,16),
                      blocklinewidth = 3, blockcolour = "blue",
                      nrows = 10, ncolumns = 16)
## End(Not run)
```

---

correct.degfree   *Check the degrees of freedom in an object of class projector*

Description

Check the degrees of freedom in an object of class "projector".

Usage

```
correct.degfree(object)
```

Arguments

object   An object of class "projector" whose degrees of freedom are to be checked.

Details

The degrees of freedom of the projector are obtained as its number of nonzero eigenvalues. An eigenvalue is regarded as zero if it is less than daeTolerance, which is initially set to.Machine$double.eps ^ 0.5 (about 1.5E-08). The function set.daeTolerance can be used to change daeTolerance.

Value

TRUE or FALSE depending on whether the correct degrees of freedom have been stored in the object of class "projector".

Author(s)

Chris Brien

See Also

degfree, projector in package **dae**.

projector for further information about this class.

Examples

```
## set up a 2 x 2 mean operator that takes the mean of a vector of 2 values
m <- matrix(rep(0.5,4), nrow=2)

## create a projector based on the matrix m
proj.m <- new("projector", data=m)

## add its degrees of freedom
degfree(proj.m) <- 1
```

```
## check degrees of freedom are correct
correct.degfree(proj.m)
```

---

dae-deprecated   *Deprecated Functions in Package dae*

Description

These functions have been renamed and deprecated in dae.

Usage

```
proj2.decomp(...)
proj2.ops(...)
```

Arguments

...   absorbs arguments passed from the old functions of the style foo.bar().

Author(s)

Chris Brien

---

decomp.relate   *Examines the relationship between the eigenvectors for two decompositions*

Description

Two decompositions produced by [proj2.eigen](#) are compared by computing all pairs of crossproduct sums of eigenvectors from the two decompositions. It is most useful when the calls to [proj2.eigen](#) have the same Q1.

Usage

```
decomp.relate(decomp1, decomp2)
```

Arguments

decomp1   A [list](#) containing components efficiencies and eigenvectors such as is produced by [proj2.eigen](#).

decomp2   Another [list](#) containing components efficiencies and eigenvectors such as is produced by [proj2.eigen](#).

Details

Each element of the r1 x r2 [matrix](#) is the sum of crossproducts of a pair of eigenvectors, one from each of the two decompositions. A sum is regarded as zero if it is less than daeTolerance, which is initially set to .Machine\$double.eps ^ 0.5 (about 1.5E-08). The function [set.daeTolerance](#) can be used to change daeTolerance.

Value

A [matrix](#) that is r1 x r2 where r1 and r2 are the numbers of efficiencies of decomp1 and decomp2, respectively. The rownames and columnnames of the [matrix](#) are the values of the efficiency factors from decomp1 and decomp2, respectively.

Author(s)

Chris Brien

See Also

[proj2.eigen](#), [proj2.combine](#) in package **dae**, [eigen](#).

Examples

```
## PBIBD(2) from p. 379 of Cochran and Cox (1957) Experimental Designs.
## 2nd edn Wiley, New York
PBIBD2.unit <- list(Block = 6, Unit = 4)
PBIBD2.nest <- list(Unit = "Block")
trt <- factor(c(1,4,2,5, 2,5,3,6, 3,6,1,4, 4,1,5,2, 5,2,6,3, 6,3,4,1))
PBIBD2.lay <- fac.layout(unrandomized = PBIBD2.unit,
                         nested.factors=PBIBD2.nest,
                         randomized = trt)

##obtain sets of projectors
Q.unit <- projs.structure(~ Block/Unit, data = PBIBD2.lay)
Q.trt <- projs.structure(~ trt, data = PBIBD2.lay)

## obtain intra- and inter-block decompositions
decomp.inter <- proj2.eigen(Q.unit[["Block"]], Q.trt[["trt"]])
decomp.intra <- proj2.eigen(Q.unit[["Block:Unit"]], Q.trt[["trt"]])

## check that intra- and inter-block decompositions are orthogonal
decomp.relate(decomp.intra, decomp.inter)
```

---

degfree   *Degrees of freedom extraction and replacement*

Description

Extracts the degrees of freedom from or replaces them in an object of class "[projector](#)".

Usage

```
degfree(object)

degfree(object) <- value
```

Arguments

object	An object of class "[projector](#)" whose degrees of freedom are to be extracted or replaced.
value	An integer to which the degrees of freedom are to be set or an object of class "[projector](#)" or "matrix" from which the degrees of freedom are to be calulated.

Details

There is no checking of the correctness of the degrees of freedom, either already stored or as a supplied integer value. This can be done using `correct.degfree`.

When the degrees of freedom of the projector are to be calculated, they are obtained as the number of nonzero eigenvalues. An eigenvalue is regarded as zero if it is less than daeTolerance, which is initially set to .Machine$double.eps ^ 0.5 (about 1.5E-08). The function `set.daeTolerance` can be used to change daeTolerance.

Value

An object of class "`projector`" that consists of a square, summetric, idempotent matrix and degrees of freedom (rank) of the matrix.

Author(s)

Chris Brien

See Also

`correct.degfree`, `projector` in package **dae**.

`projector` for further information about this class.

Examples

```
## set up a 2 x 2 mean operator that takes the mean of a vector of 2 values
m <- matrix(rep(0.5,4), nrow=2)

## coerce to a projector
proj.m <- projector(m)

## extract its degrees of freedom
degfree(proj.m)

## create a projector based on the matrix m
proj.m <- new("projector", data=m)

## add its degrees of freedom and print the projector
degfree(proj.m) <- proj.m
print(proj.m)
```

---

design.plot   *This function plots treatments within a matrix. It is being superseded by* `design.plot`.

Description

This function plots treatments within a matrix and may be used to build a graphical representation of a matrix, highlighting the position of certain treatments and the blocking factors used in the design. It is a modified version of the function supplied with DiGGer. It includes more control over the labelling of the rows and columns of the design and allows for more flexible plotting of designs with unequal block size. However, it is being supersed by `design.plot`, which has more descriptive argument names.

Usage

```
design.plot(dsgn, trts = NULL, rprop = 1, cprop = 1, label = TRUE,
                plotchar = NULL, plotbndry = TRUE,
                chtdiv = 2, rchtdiv = 1, cchtdiv = 1,
                bseq = FALSE, bdef = NULL, bcol = 1, bwd = 2,
                rotate = FALSE, new = TRUE,
                cstr = "Range",rstr = "Row", rlab = TRUE, clab = TRUE,
                font = 1, rdecrease = FALSE, cdecrease = FALSE, ...)
```

Arguments

dsgn	a `matrix` containing a set of integers or characters representing the treatments.
trts	A `integer` or `character` vector giving specific treatment labels to be plotted.
rprop	a value giving the proportion of the row boundary of cell to plot.
cprop	a value giving the proportion of the column boundary of cell to plot.
label	a `logical` to indicate whether treatment labels are to be plotted in the cells. If TRUE, print a label for all treatments or specific treatments listed in `trts`. If FALSE, no labels are not printed in the cells.
plotchar	Either a `character` vector containing labels for the whole set of treatments or a single `integer` specifying a symbol to be used in plotting treatments.
plotbndry	A `logical` indicting whether a boundary is to plotted around a cell.
chtdiv	A `numeric` specifying the amount by which plotting text and symbols in the cells should be magnified/reduced relative to the default.
rchtdiv	A `numeric` specifying the amount by which the labels of the rows of the design should be magnified/reduced relative to the default.
cchtdiv	A `numeric` specifying the amount by which the labels of the columns of the design should be magnified/reduced relative to the default.
bseq	A `logical` that determines whether block numbers are repetitions or sequences of block numbers.
bdef	A `matrix` of block sizes: • if there is only one row, then the first element is interpreted as the no. rows in each block and blocks with this number of rows are to be repeated across the rows of the design. • if there is more than one row, then each row of the matrix specifies a block, with the sequence of rows in the matrix specifying a corresponding sequence of blocks down the rows of the design. Similarly, a single value for a column specifies a repetition of blocks of that size across the columns of the design, while several column values specifies a sequence of blocks across the columns of the size specified.
bcol	A `character` `string` specifying the colour of the block boundary. See Colour specification under the `par` function.
bwd	A `numeric` giving the width of the block boundary to be plotted.
rotate	A `logical` which, if TRUE, results in the matrix being rotated 90 degrees for plotting.
new	A `logical` indicating if a new plot is to be produced or the current plot is added to.

cstr	A `character` `string` to use as a label for columns of the [matrix]().
rstr	A `character` `string` to use as a label for rows of the [matrix]().
rlab	A `logical` indicating each row of the design is labelled. If the rows of the matrix are labelled, these are used; otherwise 1:nrow is used.
clab	A `logical` indicating each column of the design is labelled. If the columns of the matrix are labelled, these are used; otherwise 1:ncol is used.
font	An `integer` specifying the font family to be used for row and column labelling.
rdecrease	A `logical` indicating whether to reverse the row labels.
cdecrease	A `logical` indicating whether to reverse the column labels.
...	further arguments passed to [polygon]() in plotting the cell.

Value

no values are returned, but a plot is produced.

References

Coombes, N. E. (2009). *DiGGer design search tool in R*. http://www.austatgen.org/files/software/downloads/

See Also

[blockboundary.plot](), [par](), [polygon](), `DiGGer`

Examples

```
## Not run:
   design.plot(des.mat, trts=1:4, col="lightblue", new=TRUE,
               rstr="Lanes", cstr="Positions", chtdiv=3, rprop = 1,cprop=1,
               plotbndry = TRUE)
   design.plot(des.mat, trts=5:87, label=T, col="grey", chtdiv=3, new=FALSE,
               plotbndry = TRUE)
   design.plot(des.mat, trts=88:434, label=T, col="lightgreen", chtdiv=3,
               new=FALSE, plotbndry = TRUE,
               bseq=TRUE, bdef=cbind(4,10,12), bwd=3, bcol="blue")
## End(Not run)
```

---

designLatinSqrSys	*Generate a systematic plan for a Latin Square design*

Description

Generates a systematic plan for a Latin Square design using the method of cycling the integers 1 to the number of treatments. The start of the cycle for each row, or the first column, can be specified as a vector of integers.

Usage

```
designLatinSqrSys(order, start = NULL)
```

Arguments

order   The number of treatments.

start   A [numeric](#) containing order unique values between one and order. These are
               interpreted as the value for the fist column for each row. If NULL, 1:order is
               used.

Value

A [numeric](#) containing order x order integers between 1 and order such that, when the [numeric](#)
is considered as a square matrix of size order, each integer occurs once and only once in each row
and column of the matrix.

Examples

```
matrix(designLatinSqrSys(5, start = c(seq(1, 5, 2), seq(2, 5, 2))), nrow=5)
designLatinSqrSys(3)
```

---

designPlot   *This function is used to give a graphical representation of an experi-*
                              mental design using labels stored in a matrix.

Description

This function uses labels, usually derived from treatment and blocking factors from an experimental
design and stored in a matrix, to build a graphical representation of the matrix, highlighting the
position of certain labels . It is a modified version of the function supplied with DiGGer. It includes
more control over the labelling of the rows and columns of the design and allows for more flexible
plotting of designs with unequal block size.

Usage

```
designPlot(designMatrix, labels = NULL, altlabels = NULL, plotlabels = TRUE,
           rtitle = NULL, ctitle = NULL,
           rlabelsreverse = FALSE, clabelsreverse = FALSE,
           font = 1, chardivisor = 2, rchardivisor = 1, cchardivisor = 1,
           cellfillcolour = NA, plotcellboundary = TRUE,
           rcellpropn = 1, ccellpropn = 1,
           blocksequence = FALSE, blockdefinition = NULL,
           blocklinecolour = 1, blocklinewidth = 2,
           rotate = FALSE, new = TRUE, ...)
```

Arguments

designMatrix   A [matrix](#) containing a set of numerics or characters representing the labels.

labels   A [numeric](#) or [character](#) vector giving the labels in the matrix for which cells
               are to be plotted. If plotlabels is TRUE and altlabels is NULL then these
               labels are plotted in the cells. See [text](#) for more information on specifying the
               labels.

altlabels	Either a [character](#) vector of the same length as labels containing an alternative set of labels for the labels currently being plotted or a single [integer](#) specifying an alternative symbol to be used in plotting cells when plotlabels is TRUE. If altlabels is NULL, the labels specified in labels is plotted when plotlabels is TRUE. See [text](#) for more information on specifying the labels.
plotlabels	A [logical](#) to indicate whether labels are to be plotted in the cells. If TRUE, print all labels or the specific labels listed in labels. If FALSE, no labels are printed in the cells.
rtitle	A [character](#) string to use as a title for rows of the plot. If rtitle is NULL then no title is plotted.
ctitle	A [character](#) string to use as a title for columns of the plot. If ctitle is NULL then no title is plotted.
rlabelsreverse	A [logical](#) indicating whether to reverse the row labels.
clabelsreverse	A [logical](#) indicating whether to reverse the column labels.
font	An [integer](#) specifying the font to be used for row and column labelling. See [par](#) for further details.
chardivisor	A [numeric](#) that changes the size of text and symbols in the cells by dividing the default size by it.
rchardivisor	A [numeric](#) that changes the size of the labels of the rows of the design by dividing the default size by it.
cchardivisor	A [numeric](#) that changes the size of the labels of the columns of the design by dividing the default size by it.
cellfillcolour	A [character](#) string specifying the colour of the fill for the cells to be plotted in this call. If there is only one colour then all cells being plotted with that colour. If there is more than one colour then the number of colours must at least equal the number of labels and then the fill colours will be matched, one for one from the first colour, with the labels. The default, NA, is to leave ther cells unfilled. See also Colour specification under the [par](#) function.
plotcellboundary	
	A [logical](#) indicting whether a boundary is to plotted around a cell.
rcellpropn	a value between 0 and 1 giving the proportion of the standard row size of a cell size to be plotted as a cell.
ccellpropn	a value between 0 and 1 giving the proportion of the standard column size of a cell size to be plotted as a cell.
blocksequence	A [logical](#) that determines whether block numbers are repetitions or sequences of block numbers.
blockdefinition	
	A [matrix](#) of block sizes: • if there is only one row, then the first element is interpreted as the no. rows in each block and blocks with this number of rows are to be repeated across the rows of the design. • if there is more than one row, then each row of the matrix specifies a block, with the sequence of rows in the matrix specifying a corresponding sequence of blocks down the rows of the design. Similarly, a single value for a column specifies a repetition of blocks of that size across the columns of the design, while several column values specifies a sequence of blocks across the columns of the size specified.

blocklinecolour

>   A character string specifying the colour of the block boundary.
>
>   See also `Colour specification` under the [par](par) function.

blocklinewidth A numeric giving the width of the block boundary to be plotted.

rotate   A logical which, if TRUE, results in the matrix being rotated 90 degrees for plotting.

new   A logical indicating if a new plot is to be produced or the current plot is added to.

...   further arguments passed to polygon in plotting the cell.

Value

no values are returned, but a plot is produced.

References

Coombes, N. E. (2009). *DiGGer design search tool in R.* http://www.austatgen.org/files/software/downloads/

See Also

blockboundaryPlot, [par](par), polygon, DiGGer

Examples

```
## Not run:
  designPlot(des.mat, labels=1:4, col="lightblue", new=TRUE, plotcellboundary = TRUE,
             rtitle="Lanes", ctitle="Positions", chardivisor=3,
             rcellpropn = 1, ccellpropn=1)
  designPlot(des.mat, labels=5:87, plotlabels=TRUE, col="grey", chardivisor=3, new=FALSE,
             plotcellboundary = TRUE)
  designPlot(des.mat, labels=88:434, plotlabels=TRUE, col="lightgreen", chardivisor=3,
             new=FALSE, plotcellboundary = TRUE,
             blocksequence=TRUE, blockdefinition=cbind(4,10,12),
             blocklinewidth=3, blockcolour="blue")
## End(Not run)
```

---

detect.diff   *Computes the detectable difference for an experiment*

Description

Computes the delta that is detectable for specified replication, power, alpha.

Usage

```
detect.diff(rm=5, df.num=1, df.denom=10, sigma=1, alpha=0.05, power=0.8,
            tol = 0.001, print=FALSE)
```

Arguments

rm	The number of observations used in computing a mean.
df.num	The degrees of freedom of the numerator of the F for testing the term involving the means.
df.denom	The degrees of freedom of the denominator of the F for testing the term involving the means.
sigma	The population standard deviation.
alpha	The significance level to be used.
power	The minimum power to be achieved.
tol	The maximum difference tolerated between the power required and the power computed in determining the detectable difference.
print	TRUE or FALSE to have or not have a table of power calculation details printed out.

Value

A single numeric value containing the computed detectable difference.

Author(s)

Chris Brien

See Also

[power.exp](), [no.reps]() in package **dae**.

Examples

```
## Compute the detectable difference for a randomized complete block design
## with four treatments given power is 0.8 and alpha is 0.05.
rm <- 5
detect.diff(rm = rm, df.num = 3, df.denom = 3 * (rm - 1),sigma = sqrt(20))
```

---

efficiencies.p2canon   *Extracts the canonical efficiency factors from a list of class* p2canon.

Description

Produces a list containing the canonical efficiency factors for the joint decomposition of two sets of projectors (Brien and Bailey, 2009) obtained using [projs.2canon]().

Usage

```
efficiencies.p2canon(object, which = "adjusted")
```

Arguments

object   A list of class p2canon produced by projs.2canon.

which   A character string, either adjusted or pairwise. For adjusted, the canoni-
                cal efficiency factor are adjusted for other projectors from Q2. For pairwise,
                they are the unadjusted canonical efficiency factors between pairs of projectors
                consisting of one projector from each of two sets.

Value

A list with a component for each element of the Q1 argument from projs.2canon. Each compo-
nent is list, each its components corresponding to an element of the Q2 argument from projs.2canon

Author(s)

Chris Brien

References

Brien, C. J. and R. A. Bailey (2009). Decomposition tables for multitiered experiments. I. A chain
of randomizations. *The Annals of Statistics*, **36**, 4184 - 4213.

See Also

projs.2canon, summary.p2canon, proj2.efficiency, proj2.combine, proj2.eigen,
projs.structure in package **dae**, eigen.

projector for further information about this class.

Examples

```
## PBIBD(2) from p. 379 of Cochran and Cox (1957) Experimental Designs.
## 2nd edn Wiley, New York
PBIBD2.unit <- list(Block = 6, Unit = 4)
PBIBD2.nest <- list(Unit = "Block")
trt <- factor(c(1,4,2,5, 2,5,3,6, 3,6,1,4, 4,1,5,2, 5,2,6,3, 6,3,4,1))
PBIBD2.lay <- fac.layout(unrandomized = PBIBD2.unit,
                         nested.factors=PBIBD2.nest,
                         randomized = trt)

##obtain projectors using projs.structure
Q.unit <- projs.structure(~ Block/Unit, data = PBIBD2.lay)
Q.trt <- projs.structure(~ trt, data = PBIBD2.lay)

##obtain combined decomposition and summarize
unit.trt.p2canon <- projs.2canon(Q.unit, Q.trt)
efficiencies.p2canon(unit.trt.p2canon)
```

---

efficiencies.pcanon   *Extracts the canonical efficiency factors from a list of class* pcanon.

Description

Produces a list containing the canonical efficiency factors for the joint decomposition of two or more sets of projectors (Brien and Bailey, 2009) obtained using `projs.canon`.

Usage

```
efficiencies.pcanon(object, which = "adjusted")
```

Arguments

object   A list of class pcanon produced by `projs.canon`.

which   A character string, either adjusted or pairwise. For adjusted, the canonical efficiency factor are adjusted for other projectors from from the same set. For pairwise, they are the unadjusted canonical efficiency factors between pairs of projectors consisting of one projector from each of two sets.

Value

A list with a component for each component of object, except for the last component – `projs.canon` for more information about the components of a pcanon object.

Author(s)

Chris Brien

References

Brien, C. J. and R. A. Bailey (2009). Decomposition tables for multitiered experiments. I. A chain of randomizations. *The Annals of Statistics*, **36**, 4184 - 4213.

See Also

`projs.canon`, `summary.pcanon`, `proj2.efficiency`, `proj2.combine`, `proj2.eigen`, `projs.structure` in package **dae**, `eigen`.

`projector` for further information about this class.

Examples

```
## PBIBD(2) from p. 379 of Cochran and Cox (1957) Experimental Designs.
## 2nd edn Wiley, New York
PBIBD2.unit <- list(Block = 6, Unit = 4)
PBIBD2.nest <- list(Unit = "Block")
trt <- factor(c(1,4,2,5, 2,5,3,6, 3,6,1,4, 4,1,5,2, 5,2,6,3, 6,3,4,1))
PBIBD2.lay <- fac.layout(unrandomized = PBIBD2.unit,
                         nested.factors=PBIBD2.nest,
                         randomized = trt)

##obtain combined decomposition and summarize
```

```
unit.trt.canon <- projs.canon(list(unit=~ Block/Unit, trt=~ trt), data = PBIBD2.lay)
efficiencies.pcanon(unit.trt.canon)
```

---

efficiency.criteria   *Computes efficiency criteria from a set of efficiency factors*

Description

Computes efficiency criteria from a set of efficiency factors.

Usage

```
efficiency.criteria(efficiencies)
```

Arguments

efficiencies   A numeric containing a set of efficiency factors.

Details

The `aefficiency` criterion is the harmonic mean of the nonzero efficiency factors. The `mefficiency` criterion is the mean of the nonzero efficiency factors. The `eefficiency` criterion is the minimum of the nonzero efficiency factors. The `sefficiency` criterion is the variance of the nonzero efficiency factors. The `xefficiency` is the maximum of the efficiency factors. The `order` is the order of balance and is the number of unique nonzero efficiency factors. The `dforthog` is the number of efficiency factors that are equal to one.

Value

A `list` whose components are `aefficiency`, `mefficiency`, `sefficiency`, `eefficiency`, `xefficiency`, `order` and `dforthog`.

Author(s)

Chris Brien

See Also

[proj2.efficiency](), [proj2.eigen](), [proj2.combine]() in package **dae**, [eigen]().

[projector]() for further information about this class.

Examples

```
## PBIBD(2) from p. 379 of Cochran and Cox (1957) Experimental Designs.
## 2nd edn Wiley, New York
PBIBD2.unit <- list(Block = 6, Unit = 4)
PBIBD2.nest <- list(Unit = "Block")
trt <- factor(c(1,4,2,5, 2,5,3,6, 3,6,1,4, 4,1,5,2, 5,2,6,3, 6,3,4,1))
PBIBD2.lay <- fac.layout(unrandomized = PBIBD2.unit,
                         nested.factors=PBIBD2.nest,
                         randomized = trt)

## obtain sets of projectors
```

```
Q.unit <- projs.structure(~ Block/Unit, data = PBIBD2.lay)
Q.trt <- projs.structure(~ trt, data = PBIBD2.lay)

## save intrablock efficiencies
eff.inter <- proj2.efficiency(Q.unit[["Block:Unit"]], Q.trt[["trt"]])

## calculate efficiency criteria
efficiency.criteria(eff.inter)
```

---

elements   *Extract the elements of an array specified by the subscripts*

Description

Elements of the `array x` corresponding to the rows of the two dimensional object `subscripts` are extracted. The number of columns of `subscripts` corresponds to the number of dimensions of `x`. The effect of supplying less columns in `subscripts` than the number of dimensions in `x` is the same as for `"["`.

Usage

```
elements(x, subscripts)
```

Arguments

x	An array with at least two dimensions whose elements are to be extracted.
subscripts	A two dimensional object interpreted as elements by dimensions.

Value

A `vector` containing the extracted elements and whose length equals the number of rows in the `subscripts` object.

See Also

`Extract`

Examples

```
## Form a table of the means for all combinations of Row and Line.
## Then obtain the values corresponding to the combinations in the data frame x,
## excluding Row 3.
x <- fac.gen(list(Row = 2, Line = 4), each =2)
x$y <- rnorm(16)
RowLine.tab <- tapply(x$y, list(x$Row, x$Line), mean)
xs <- elements(RowLine.tab, subscripts=x[x$"Line" != 3, c("Row", "Line")])
```

---

extab   *Expands the values in table to a vector*

Description

Expands the values in `table` to a `vector` according to the `index.factors` that are considered to index the `table`, either in standard or Yates order. The order of the values in the `vector` is determined by the order of the values of the `index.factors`.

Usage

```
extab(table, index.factors, order="standard")
```

Arguments

table   A numeric `vector` containing the values to be expanded. Its length must equal the product of the number of used levels for the `factor`s in `index.factors` and the values in it correspond to all levels combinations of these `factor`s. That is, the values of the `index.factors` are irrelevant to `table`.

index.factors   A list of `factor`s that index the `table`. All the `factor`s must be the same length.

order   The order in which the levels combinations of the `index.factors` are to be considered as numbered in indexing `table`; `standard` numbers them as if they are arranged in standard order, that is with the first factor moving slowest and the last factor moving fastest; `yates` numbers them as if they are arranged in Yates order, that is with the first factor moving fastest and last factor moving slowest.

Value

A `vector` of length equal to the `factor`s in `index.factor` whose values are taken from `table`.

Author(s)

Chris Brien

Examples

```
## generate a small completely randomized design with the two-level
## factors A and B
n <- 12
CRD.unit <- list(Unit = n)
CRD.treat <- fac.gen(list(A = 2, B = 2), each = 3)
CRD.lay <- fac.layout(unrandomized=CRD.unit, randomized=CRD.treat,
                      seed=956)

## set up a 2 x 2 table of A x B effects
AB.tab <- c(12, -12, -12, 12)

## add a unit-length vector of expanded effects to CRD.lay
attach(CRD.lay)
CRD.lay$AB.effects <- extab(table=AB.tab, index.factors=list(A, B))
```

---

fac.ar1mat *forms the ar1 correlation matrix for a (generalized) factor*

Description

Form the correlation matrix for a (generalized) factor where the correlation between the levels follows an autocorrelation of order 1 (ar1) pattern.

Usage

```
fac.ar1mat(factor, rho)
```

Arguments

factor   The (generalized) factor for which the correlation between its levels displays
                an ar1 pattern.

rho   The correlation parameter for the ar1 process.

Details

The method is: a) form an n x n matrix of all pairwise differences in the numeric values corresponding to the observed levels of the factor by taking the difference between the following two n x n matrices are equal: 1) each row contains the numeric values corresponding to the observed levels of the factor, and 2) each column contains the numeric values corresponding to the observed levels of the factor, b) replace each element of the pairwise difference matrix with rho raised to the absolute value of the difference.

Value

An n x n matrix, where n is the length of the factor.

Author(s)

Chris Brien

See Also

fac.vcmat, fac.meanop, fac.sumop in package **dae**.

Examples

```
## set up a two-level factor and a three-level factor, both of length 12
A <- factor(rep(1:2, each=6))
B <- factor(rep(1:3, each=2, times=2))

## create a 12 x 12 ar1 matrix corrresponding to B
ar1.B <- fac.ar1mat(B, 0.6)
```

---

fac.combine   *Combines several factors into one*

Description

Combines several [factors](#) into one whose levels are the combinations of the used levels of the individual [factors](#).

Usage

```
fac.combine(factors, order="standard", combine.levels=FALSE, sep=",", ...)
```

Arguments

factors	A [list](#) of [factors](#) all of the same length.
order	Either standard or yates. The order in which the levels combinations of the [factors](#) are to be considered as numbered when forming the levels of the combined [factor](#); standard numbers them as if they are arranged in standard order, that is with the levels of the first factor moving slowest and those of the last factor moving fastest; yates numbers them as if they are arranged in Yates order, that is with the levels of the first factor moving fastest and those of the last factor moving slowest.
combine.levels	A logical specifying whether the levels labels of the new [factor](#) are to be combined from those of the [factors](#) being combined. The default is to use the integers from 1 to the product of the numbers of combinations of used levels of the individual [factors](#), numbering the levels according to order.
sep	A character string to separate the levels when combine.levels = TRUE.
...	Further arguments passed to the [factor](#) call creating the new [factor](#).

Value

A [factor](#) whose levels are formed form the observed combinations of the levels of the individual [factors](#).

Author(s)

Chris Brien

See Also

[fac.divide](#) in package **dae**.

Examples

```
## set up two factors
A <- factor(rep(1:2, each=6))
B <- factor(rep(1:3, each=2, times=2))

## obtain six-level factor corresponding to the combinations of A and B
AB <- fac.combine(list(A,B))
```

---

fac.divide *Divides a factor into several individual factors*

Description

Takes a factor and divides it into several individual factors as if the levels in the original combined.factor are numbered from one to its number of levels and correspond to the numbering of the levels combinations of the individual factors when these are arranged in standard or Yates order.

Usage

```
fac.divide(combined.factor, factor.names, order="standard")
```

Arguments

combined.factor

A factor that is to be divided into the individual factors listed in factor.names.

factor.names   A list of factors to be formed. The names in the list are the names of the factors and the component of a name is either a) a single numeric value that is the number of levels, b) a numeric vector that contains the levels of the factor, or c) a character vector that contains the labels of the levels of the factor.

order   Either standard or yates. The order in which the levels combinations of the factors in factor.names are to be considered as numbered; standard numbers them as if they are arranged in standard order, that is with the first factor moving slowest and the last factor moving fastest; yates numbers them as if they are arranged in Yates order, that is with the first factor moving fastest and last factor moving slowest.

Value

A data.frame whose columns consist of the factors listed in factor.names and whose values have been computed from the combined factor. All the factors will be of the same length.

Note

A single factor name may be supplied in the list in which case a data.frame is produced that contains the single factor computed from the numeric vector. This may be useful when calling this function from others.

Author(s)

Chris Brien

See Also

fac.combine in package **dae**.

Examples

```
## generate a small completely randomized design for 6 treatments
n <- 12
CRD.unit <- list(Unit = n)
treat <- factor(rep(1:4, each = 3))
CRD.lay <- fac.layout(unrandomized=CRD.unit, randomized=treat, seed=956)

## divide the treatments into two two-level factor A nd B
CRD.facs <- fac.divide(CRD.lay$treat, factor.names = list(A = 2, B = 2))
```

---

fac.gen   *Generate all combinations of several factors*

Description

Generate all combinations of several factors.

Usage

```
fac.gen(generate, each=1, times=1, order="standard")
```

Arguments

generate	A [list] of named objects and numbers that specify the [factor]s whose levels are to be generated and the pattern in these levels.If a component of the [list] is named, then the component should be either a) a single numeric value that is the number of levels, b) a numeric vector that contains the levels of the [factor], or c) a character vector that contains the labels of the levels of the [factor].
each	The number of times to replicate consecutively the elements of the levels generated according to pattern specified by the generate argument.
times	The number of times to repeat the whole generated pattern of levels generated according to pattern specified by the generate argument.
order	Either standard or yates. The order in which the speed of cycling through the levels is to move; combinations of the [factor]s are to be considered as numbered; standard cycles through the levels of the first factor slowest and the last factor moving fastest; yates cycles through the levels of the first factor fastest and last factor moving slowest.

Details

The levels of each [factor] are generated in a hierarchical pattern where the levels of one [factor] are held constant while those of the adjacent [factor] are cycled through the complete set once. If a number is supplied instead of a name, the pattern is generated as if a [factor] with that number of levels had been supplied in the same position as the number. However, no levels are stored for this unamed [factor].

Value

A [data.frame] of generated levels with columns corresponding to the codefactors in the generate list.

Warning

Avoid using factor names F and T as these might be confused with FALSE and TRUE.

Author(s)

Chris Brien

See Also

[fac.combine](#) in package **dae**

Examples

```
## generate a 2^3 factorial experiment with levels - and +, and
## in Yates order
mp <- c("-", "+")
fnames <- list(Catal = mp, Temp = mp, Press = mp, Conc = mp)
Fac4Proc.Treats <- fac.gen(generate = fnames, order="yates")

## Generate the factors A, B and D. The basic pattern has 4 repetitions
## of the levels of D for each A and B combination and 3 repetitions of
## the pattern of the B and D combinations for each level of A. This basic
## pattern has each combination repeated twice, and the whole of this
## is repeated twice. It generates 864 A, B and D combinations.
gen <- list(A = 3, 3, B = c(0,100,200), 4, D = c("0","1"))
fac.gen(gen, times=2, each=2)
```

---

fac.layout   *Generate a randomized layout for an experiment*

Description

Given a systematic design, generate a layout for an experiment consisting of randomized [factor](#)s that are randomized to the unrandomized [factor](#)s, taking into account the nesting between the unrandomized factors for the design.

Usage

```
fac.layout(unrandomized, nested.factors=NULL, except=NULL,
              randomized, seed=NULL, unit.permutation=TRUE)
```

Arguments

unrandomized   A [data.frame](#) or a [list](#) of [factor](#)s, along with their levels. If a [list](#), the name of each component of the [list](#) is a [factor](#) name and the component is either a single numeric value that is the number of levels, a numeric vector that contains the levels of the [factor](#) or a character codevector that contains the labels of the levels of the [factor](#).

nested.factors A [list](#) of the unrandomized [factor](#)s that are nested in other [factor](#)s in unrandomized. The name of each component is the name of a [factor](#) that is nested and the component is a character vector containing the [factor](#)s within which it is nested. It is emphasized that the nesting is a property of the design that is being employed (it is only partly based on the intrinsic nesting.

except	A character vector containing the names unrandomized factors that are to be excepted from the randomization.
randomized	A factor or a data.frame containing the values of the factor(s) to be randomized.
seed	A single value, interpreted as an integer, that specifies the starting value of the random number generator.
unit.permutation	
	A logical indicating whether to include the .Unit and .Permutation columns in the data.frame.

Details

This functon uses the method of randomization described by Bailey (1981). That is, a permutation of the units that respects the nesting for the design, but does not permute any of the factors in the except vector, is obtained. This permutation is applied to the unrandomized factors and then a data.frame containing both the permuted unrandomized and unpermuted randomized factors is formed. To produce the randomized layout, the rows of the joint data.frame are reordered so that its unrandomized factors are in either standard order or, if a data.frame was suppled to unrandomized, data frame order.

The .Units and .Permutation vectors enable one to swap between this permutation and the randomized layout. The *i*th value in .Permutation gives the unit to which unit *i* was assigned in the randomization.

Value

A data.frame consisting of the values for .Units and .Permutation vectors, provided unit.permutation is TRUE, along with the values for the unrandomized and randomized factors that specify the randomized layout for the experiment.

Author(s)

Chris Brien

References

Bailey, R.A. (1981) A unified approach to design of experiments. *Journal of the Royal Statistical Society, Series A*, **144**, 214–223.

See Also

fac.gen in package **dae**.

Examples

```
## generate a randomized layout for a 4 x 4 Latin square
## (the nested.factors agument is not needed here as none of the
## factors are nested)
LS.unit <- data.frame(row = ordered(rep(c("I","II","III","IV"), times=4)),
                      col = factor(rep(c(0,2,4,6), each=4)))
LS.ran <- data.frame(treat = factor(c(1:4, 2,3,4,1, 3,4,1,2, 4,1,2,3)))
data.frame(LS.unit, LS.ran)
LS.lay <- fac.layout(unrandomized=LS.unit, randomized=LS.ran, seed=7197132)
LS.lay[LS.lay$.Permutation,]
```

```
## generate a randomized layout for a replicated randomized complete
## block design, with the block factors arranged in standard order for
## rep then plot and then block
RCBD.unit <- list(rep = 2, plot=1:3, block = c("I","II"))
## specify that plot is nested in block and rep and that block is nested
## in rep
RCBD.nest <- list(plot = c("block","rep"), block="rep")
## generate treatment factor in systematic order so that they correspond
## to plot
tr <- factor(rep(1:3, each=2, times=2))
## obtain randomized layout
RCBD.lay <- fac.layout(unrandomized=RCBD.unit,
                       nested.factors=RCBD.nest,
                       randomized=tr, seed=9719532)
#sort into the original standard order
RCBD.perm <- RCBD.lay[RCBD.lay$.Permutation,]
#resort into randomized order
RCBD.lay <- RCBD.perm[order(RCBD.perm$.Units),]

## generate a layout for a split-unit experiment in which:
## - the main-unit factor is A with 4 levels arranged in
##    a randomized complete block design with 2 blocks;
## - the split-unit factor is B with 3 levels.
SPL.lay <- fac.layout(unrandomized=list(block = 2, main.unit = 4, split.unit = 3),
                       nested.factors=list(main.unit = "block",
                                           split.unit = c("block", "main.unit")),
                       randomized=fac.gen(list(A = 4, B = 3), times = 2),
                       seed=155251978, unit.permutation=FALSE)
```

---

fac.match	*Match, for each combination of a set of columns in* x, *the row that has the same combination in* table

---

Description

Match, for each combination of a set of columns in x, the rows that has the same combination in table. The argument multiples.allow controls what happens when there are multple matches in table of a combination in x.

Usage

```
fac.match(x, table, col.names, nomatch = NA_integer_, multiples.allow = FALSE)
```

Arguments

x	an R object, normally a data.frame, possibly a matrix.
table	an R object, normally a data.frame, possibly a matrix.
col.names	A character vector giving the columns in x and table that are to be matched.
nomatch	The value to be returned in the case when no match is found. Note that it is coerced to integer.

multiples.allow

> A `logical` indicating whether multiple matches of a combination in `x` to those
> in `table` is allowed. If `multiples.allow` is `FALSE`, an error is generated. If
> `multiples.allow` is `TRUE`, the first occuence in `table` is matched. This func-
> tion can be viewed as a generalization to multiple vectors of the `match` function
> that applies to single vectors.

Value

A `vector` of length equal to `x` that gives the rows in `table` that match the combinations of `col.names`
in `x`. The order of the rows is the same as the order of the combintions in `x`. The value returned if a
combination is unmatched is specified in the `nomatch` argument.

Author(s)

Chris Brien

See Also

`match`

Examples

```
## Not run:
#A single unmatched combination
kdata <- data.frame(Expt="D197-5",
                    Row=8,
                    Column=20, stringsAsFactors=FALSE)
index <- fac.match(kdata, D197.dat, c("Expt", "Row", "Column"))

# A matched and an unmatched combination
kdata <- data.frame(Expt=c("D197-5", "D197-4"),
                    Row=c(8, 10),
                    Column=c(20, 8), stringsAsFactors=FALSE)
index <- fac.match(kdata, D197.dat, c("Expt", "Row", "Column"))

## End(Not run)
```

---

fac.meanop   *computes the projection matrix that produces means*

Description

Computes the symmetric projection matrix that produces the means corresponding to a (general-
ized) `factor`.

Usage

```
fac.meanop(factor)
```

Arguments

factor   The (generalized) `factor` whose means the projection matrix computes from an
                observation-length vector.

Details

The design matrix **X** for a (generalized) [factor](#) is formed with a column for each level of the (generalized) [factor](#), this column being its indicator variable. The projection matrix is formed as X %*% (1/diag(r) %*% t(X), where r is the vector of levels replications.

A generalized [factor](#) is a [factor](#) formed from the combinations of the levels of several original [factor](#)s. Generalized [factor](#)s can be formed using [fac.combine](#).

Value

A [projector](#) containing the symmetric, projection matrix and its degrees of freedom.

Author(s)

Chris Brien

See Also

[fac.combine](#), [projector](#), [degfree](#), [correct.degfree](#), [fac.sumop](#) in package **dae**.

[projector](#) for further information about this class.

Examples

```
## set up a two-level factor and a three-level factor, both of length 12
A <- factor(rep(1:2, each=6))
B <- factor(rep(1:3, each=2, times=2))

## create a generalized factor whose levels are the combinations of A and B
AB <- fac.combine(list(A,B))

## obtain the operator that computes the AB means from a vector of length 12
M.AB <- fac.meanop(AB)
```

---

fac.nested	*creates a factor whose values are generated within those of the factor nesting.fac*

Description

Creates a [factor](#) whose levels are generated within those of the factor nesting.fac. All elements of nesting.fac having the same level are numbered from 1 to the number of different elements having that level.

Usage

```
fac.nested(nesting.fac, levels=NA, labels=NA, ...)
```

Arguments

nesting.fac	The factor within each of whose levels the created factor is to be generated.
levels	Optional vector of levels for the factor. Any data value that does not match a value in levels will be NA in the factor. The default value of levels is the the list of numbers from 1 to the maximum replication of the levels of nesting.fac, represented as characters.
labels	Optional vector of values to use as labels for the levels of the factor. The default is as.character(levels).
...	Further arguments passed to the factor call dreating the new factor.

Value

A factor that is a character vector with class attribute "factor" and a levels attribute which determines what character strings may be included in the vector.

Note

The levels of nesting.fac do not have to be equally replicated.

Author(s)

Chris Brien

See Also

[fac.gen](fac.gen) in package **dae**, factor.

Examples

```
## set up factor A
A <- factor(c(1, 1, 1, 2, 2))

## create nested factor
B <- fac.nested(A)
```

---

fac.recode	*Recodes factor* levels *using values in a vector. The values in the vector do not have to be unique.*

Description

Recodes the levels and values of a factor using each value in the newlevels vector to replace the corresponding value in the vector of levels of the factor.

Usage

```
fac.recode(factor, newlevels, ...)
```

Arguments

factor	The factor to be recoded.
newlevels	A vector of length levels(factor) containing values to use in the recoding.
...	Further arguments passed to the factor call creating the new factor.

Value

A factor.

Author(s)

Chris Brien

See Also

as.numfac and mpone in package **dae**, factor, relevel.

Examples

```
## set up a factor with labels
a <- factor(rep(1:4, 4), labels=c("A","B","C","D"))

## recode "A" and "D" to 1 and "B" and "C" to 2
b <- fac.recode(a, c(1,2,2,1), labels = c("a","b"))
```

---

fac.sumop	*computes the summation matrix that produces sums corresponding to a (generalized) factor*

Description

Computes the matrix that produces the sums corresponding to a (generalized) factor.

Usage

```
fac.sumop(factor)
```

Arguments

factor	The (generalized) factor whose sums the summation matrix computes from an observation-length vector.

Details

The design matrix **X** for a (generalized) factor is formed with a column for each level of the (generalized) factor, this column being its indicator variable. The summation matrix is formed as X %*% t(X).

A generalized factor is a factor formed from the combinations of the levels of several original factors. Generalized factors can be formed using fac.combine.

Value

A symmetric matrix.

Author(s)

Chris Brien

See Also

[fac.combine](#), [fac.meanop](#) in package **dae**.

Examples

```
## set up a two-level factoir and a three-level factor, both of length 12
A <- factor(rep(1:2, each=6))
B <- factor(rep(1:3, each=2, times=2))

## create a generlaized factor whose levels are the combinations of A and B
AB <- fac.combine(list(A,B))

## obtain the operator that computes the AB means from a vector of length 12
S.AB <- fac.sumop(AB)
```

---

fac.vcmat	*forms the variance matrix for the variance component of a (generalized) factor*

Description

Form the variance matrix for a (generalized) factor whose effects for its different levels are independently and identically distributed, with their variance given by the variance component; elements of the matrix will equal either zero or sigma2 and displays compound symmetry.

Usage

```
fac.vcmat(factor, sigma2)
```

Arguments

factor	The (generalized) [factor](#) for which the variance matrix is required.
sigma2	The variance component, being the of the random effects for the factor.

Details

The method is: a) form the n x n summation or relationship matrix whose elements are equal to zero except for those elements whose corresponding elements in the following two n x n matrices are equal: 1) each row contains the numeric values corresponding to the observed levels of the factor, and 2) each column contains the numeric values corresponding to the observed levels of the factor, b) multiply the summation matrix by sigma2.

Value

An n x n [matrix](#), where n is the length of the [factor](#).

Author(s)

Chris Brien

See Also

`fac.ar1mat`, `fac.meanop`, `fac.sumop` in package **dae**.

Examples

```
## set up a two-level factor and a three-level factor, both of length 12
A <- factor(rep(1:2, each=6))
B <- factor(rep(1:3, each=2, times=2))

## create a 12 x 12 ar1 matrix corrresponding to B
vc.B <- fac.vcmat(B, 2)
```

Fac4Proc.dat   *Data for a 2^4 factorial experiment*

Description

The data set come from an unreplicated 2^4 factorial experiment to investigate a chemical process. The response variable is the Conversion percentage (Conv) and this is indexed by the 4 two-level factors Catal, Temp, Press and Conc, with levels "-" and "+". The data is aranged in Yates order. Also included is the 16-level factor Runs which gives the order in which the combinations of the two-level factors were run.

Usage

```
data(Fac4Proc.dat)
```

Format

A data.frame containing 16 observations of 6 variables.

Source

Table 10.6 of Box, Hunter and Hunter (1978) *Statistics for Experimenters*. New York, Wiley.

---

fitted.aovlist   *Extract the fitted values for a fitted model from an aovlist object*

Description

Extracts the fitted values as the sum of the effects for all the fitted terms in the model, stopping at
`error.term` if this is specified. It is a method for the generic function `fitted`.

Usage

```
## S3 method for class 'aovlist'
fitted(object, error.term=NULL, ...)
```

Arguments

`object`	An aovlist object created from a call to `aov`.
`error.term`	The term from the `Error` function down to which effects are extracted for adding to the fitted values. The order of terms is as given in the ANOVA table. If `error.term` is NULL effects are extracted from all `Error` terms.
`...`	Further arguments passed to or from other methods.

Value

A numeric `vector` of fitted values.

Note

Fitted values will be the sum of effects for terms from the model, but only for terms external to any
`Error` function. If you want effects for terms in the `Error` function to be included, put them both
inside and outside the `Error` function so they are occur twice.

Author(s)

Chris Brien

See Also

`fitted.errors`, `resid.errors`, `tukey.1df` in package **dae**.

Examples

```
## set up data frame for randomized complete block design in Table 4.4 from
## Box, Hunter and Hunter (2005) Statistics for Experimenters. 2nd edn
## New York, Wiley.
RCBDPen.dat <- fac.gen(list(Blend=5, Flask=4))
RCBDPen.dat$Treat <- factor(rep(c("A","B","C","D"), times=5))
RCBDPen.dat$Yield <- c(89,88,97,94,84,77,92,79,81,87,87,
                       85,87,92,89,84,79,81,80,88)

## perform the analysis of variance
RCBDPen.aov <- aov(Yield ~ Blend + Treat + Error(Blend/Flask), RCBDPen.dat)
summary(RCBDPen.aov)
```

```
## two equivalent ways of extracting the fitted values
fit  <- fitted.aovlist(RCBDPen.aov)
fit <- fitted(RCBDPen.aov, error.term = "Blend:Flask")
```

---

fitted.errors   *Extract the fitted values for a fitted model*

Description

An alias for the generic function `fitted`. When it is available, the method `fitted.aovlist` extracts the fitted values, which is provided in the **dae** package to cover aovlist objects.

Usage

```
## S3 method for class 'errors'
fitted(object, error.term=NULL, ...)
```

Arguments

object	An aovlist object created from a call to `aov`.
error.term	The term from the `Error` function down to which effects are extracted for adding to the fitted values. The order of terms is as given in the ANOVA table. If `error.term` is `NULL` effects are extracted from all `Error` terms.
...	Further arguments passed to or from other methods.

Value

A numeric vector of fitted values.

Warning

See `fitted.aovlist` for specific information about fitted values when an `Error` function is used in the call to the `aov` function.

Author(s)

Chris Brien

See Also

`fitted.aovlist`, `resid.errors`, `tukey.1df` in package **dae**.

Examples

```
## set up data frame for randomized complete block design in Table 4.4 from
## Box, Hunter and Hunter (2005) Statistics for Experimenters. 2nd edn
## New York, Wiley.
RCBDPen.dat <- fac.gen(list(Blend=5, Flask=4))
RCBDPen.dat$Treat <- factor(rep(c("A","B","C","D"), times=5))
RCBDPen.dat$Yield <- c(89,88,97,94,84,77,92,79,81,87,87,
                       85,87,92,89,84,79,81,80,88)
```

```
## perform the analysis of variance
RCBDPen.aov <- aov(Yield ~ Blend + Treat + Error(Blend/Flask), RCBDPen.dat)
summary(RCBDPen.aov)

## three equivalent ways of extracting the fitted values
fit  <- fitted.aovlist(RCBDPen.aov)
fit <- fitted(RCBDPen.aov, error.term = "Blend:Flask")
fit <- fitted.errors(RCBDPen.aov, error.term = "Blend:Flask")
```

---

get.daeTolerance   *Gets the value of daeTolerance for the package dae*

Description

A function that gets the `vector` of values such that, in **dae** functions, values less than it are considered to be zero.

Usage

```
get.daeTolerance()
```

Value

The `vector` of two values for daeTolerance, one named `element.tol` that is used for elements of matrices and a second named `element.eigen` that is used for eigenvalues and quantities based on them, such as efficiency factors.

Author(s)

Chris Brien

See Also

[set.daeTolerance](#).

Examples

```
## get daeTolerance.
get.daeTolerance()
```

---

harmonic.mean   *Calcuates the harmonic mean.*

Description

A function to calcuate the harmonic mean of a set of nonzero numbers.

Usage

```
harmonic.mean(x)
```

Arguments

x   An object from whose elements the harmonic mean is to be computed.

Details

All the elements of x are tested as being less than daeTolerance, which is initially set to .Machine$double.eps ^ 0.5 (about 1.5E-08). The function `set.daeTolerance` can be used to change daeTolerance.

Value

A numeric. Returns Inf if x contains a value close to zero

Examples

```
y <- c(seq(0.1,1,0.2))
harmonic.mean(y)
```

---

interaction.ABC.plot   *Plots an interaction plot for three factors*

Description

Plots a function (the mean by default) of the response for the combinations of the three factors specified as the x.factor (plotted on the x axis of each plot), the groups.factor (plotted as separate lines in each plot) and the trace.factor (its levels are plotted in different plots). Interaction plots for more than three factors can be produced by using fac.combine to combine all but two of them into a single factor that is specified as the trace.factor.

Usage

```
interaction.ABC.plot(response, x.factor, groups.factor,
      trace.factor,data, fun="mean", title="A:B:C Interaction Plot",
      xlab, ylab, key.title, lwd=4, columns=2, ggplotFuncs = NULL, ...)
```

Arguments

response	A numeric vector containing the response variable from which a function (the mean by default) is computed for plotting on the y-axis.
x.factor	The factor to be plotted on the x-axis of each plot.
groups.factor	The factor plotted as separate lines in each plot.
trace.factor	The factor for whose levels there are separate plots.
data	A data.frame containing the three factors and the response.
fun	The function to be computed from the response for each combination of the three factors x.factor, groups.factor and trace.factor. By default, the mean is computed for each combination.
title	Title for plot window. By default it is "A:B:C Interaction Plot".
xlab	Label for the x-axis. By default it is the name of the x.factor.
ylab	Label for the y-axis. By default it is the name of the response.

key.title   Label for the key (legend) to the lines in each plot. By default it is the name of
                 the groups.factor.

lwd   The width of the lines. By default it is 4.

columns   The number of columns for arranging the several plots for the levels of the
                 groups.factor. By default it is 2.

ggplotFuncs   A list, each element of which contains the results of evaluating a ggplot func-
                 tion. It is created by calling the list function with a ggplot function call for
                 each element.

...   Other arguments that are passed down to ggplot methods.

Value

An object of class "ggplot", which can be plotted using print.

Author(s)

Chris Brien

See Also

fac.combine in package **dae**, interaction.plot.

Examples

```
## Not run:
## plot for Example 14.1 from Mead, R. (1990). The Design of Experiments:
## Statistical Principles for Practical Application. Cambridge,
## Cambridge University Press.
## use ?SPLGrass.dat for details
data(SPLGrass.dat)
interaction.ABC.plot(Main.Grass, x.factor=Period,
                     groups.factor=Spring, trace.factor=Summer,
                     data=SPLGrass.dat,
                     title="Effect of Period, Spring and Summer on Main Grass")

## plot for generated data
## use ?ABC.Interact.dat for data set details
data(ABC.Interact.dat)
## Add standard errors for plotting
## - here data contains a single value for each combintion of A, B and C
## - need to supply name for data twice
ABC.Interact.dat$se <- rep(c(0.5,1), each=4)
interaction.ABC.plot(MOE, A, B, C, data=ABC.Interact.dat,
                     ggplotFunc=list(geom_errorbar(data=ABC.Interact.dat,
                                                   aes(ymax=MOE+se, ymin=MOE-se),
                                                   width=0.2)))

## End(Not run)
```

---

is.allzero | *Tests whether all elements are approximately zero*

Description

A single-line `function` that tests whether all elements are zero (approximately).

Usage

```
is.allzero(x)
```

Arguments

x   An object whose elements are to be tested.

Details

The mean of the absolute values of the elements of x is tested to determine if it is less than daeTolerance, which is initially set to `.Machine$double.eps ^ 0.5` (about 1.5E-08). The function `set.daeTolerance` can be used to change daeTolerance.

Value

A `logical`.

Author(s)

Chris Brien

Examples

```
## create a vector of 9 zeroes and a one
y <- c(rep(0,9), 1)

## check that vector is only zeroes is FALSE
is.allzero(y)
```

---

is.projector | *Tests whether an object is a valid object of class projector*

Description

Tests whether an `object` is a valid object of class "`projector`".

Usage

```
is.projector(object)
```

Arguments

object   The `matrix` to be made into a projector.

Details

The function is.projector tests whether the object consists of a matrix that is square, symmetric and idempotent. In checking symmetry and idempotency, the equality of the matrix with either its transpose or square is tested. In this, a difference in elements is considered to be zero if it is less than daeTolerance, which is initially set to .Machine$double.eps ^ 0.5 (about 1.5E-08). The function set.daeTolerance can be used to change daeTolerance.

Value

TRUE or FALSE depending on whether the object is a valid object of class "projector".

Warning

The degrees of freedom are not checked. correct.degfree can be used to check them.

Author(s)

Chris Brien

See Also

projector, correct.degfree in package **dae**.

projector for further information about this class.

Examples

```
## set up a 2 x 2 mean operator that takes the mean of a vector of 2 values
m <- matrix(rep(0.5,4), nrow=2)

## create an object of class projector
proj.m <- projector(m)

## check that it is a valid projector
is.projector(proj.m)
```

---

mat.ar1   *Forms an ar1 correlation matrix*

Description

Form the correlation matrix of order order whose correlations follow the ar1 pattern. The matrix is banded and has diagonal elements equal to one and the off-diagonal element in the ith row and jth column equal to ρ^k where $k = |i - j|$.

Usage

```
mat.ar1(rho, order)
```

Arguments

rho	The correlation on the first off-diagonal.
order	The order of the matrix to be formed.

Value

A banded correlation [matrix](#) whose elements follow an ar1 pattern.

See Also

[mat.I](#), [mat.J](#), [mat.exp](#), [mat.banded](#), [mat.ar2](#), [mat.ar3](#), [mat.sar2](#), [mat.ma1](#), [mat.ma2](#), [mat.arma](#)

Examples

```
corr <- mat.ar1(rho=0.4, order=4)
```

---

mat.ar2 *Forms an ar2 correlation matrix*

Description

Form the correlation [matrix](#) of order order whose correlations follow the ar2 pattern. The resulting [matrix](#) is banded.

Usage

```
mat.ar2(ARparameters, order)
```

Arguments

ARparameters   A [numeric](#) containing the three autoregressive parameter values of the process, being the weights given to the lag 1 and lag 2 response values.

order   The order of the [matrix](#) to be formed.

Details

The correlations in the correlation matrix, corr say, are calculated from the autoregressive parameters, ARparameters. The values in

- the diagonal (k = 1) of corr are one;
- the first subdiagonal band (k = 2) of corr are equal to ARparameters[1]/(1-ARparameters[2]);
- in subsequent disgonal bands, (k = 3:order), of corr are ARparameters[1]*corr[k-1] + ARparameters[2]*corr[k-2].

Value

A banded correlation [matrix](#) whose elements follow an ar2 pattern.

See Also

[mat.I](#), [mat.J](#), [mat.exp](#), [mat.banded](#), [mat.ar1](#), [mat.ar3](#), [mat.sar2](#), [mat.ma1](#), [mat.ma2](#), [mat.arma](#)

Examples

```
corr <- mat.ar2(ARparameters = c(0.4, 0.2), order = 4)
```

mat.ar3   *Forms an ar3 correlation matrix*

Description

Form the correlation [matrix](#) of order order whose correlations follow the ar3 pattern. The resulting [matrix](#) is banded.

Usage

```
mat.ar3(ARparameters, order)
```

Arguments

ARparameters   A [numeric](#) containing the three autoregressive parameter values of the process,
                being the weights given to the lag 1, lag 2 and lag 3 response values.

order   The order of the [matrix](#) to be formed.

Details

The correlations in the correlation matrix, corr say, are calculated from the autoregressive parameters, ARparameters.
Let omega = 1 - ARparameters[2] - ARparameters[3] * (ARparameters[1] + ARparameters[3]).
Then the values in

- the diagonal of corr (k = 1) are one;

- the first subdiagonal band (k = 2) of corr are equal to
  (ARparameters[1] + ARparameters[2]*ARparameters[3]) / omega;

- the second subdiagonal band (k = 3) of corr are equal to
  (ARparameters[1] * (ARparameters[1] + ARparameters[3]) +
  ARparameters[2] * (1 - ARparameters[2])) / omega;

- the subsequent subdiagonal bands, (k = 4:order), of corr are equal to
  ARparameters[1]*corrs[k-1] + ARparameters[2]*corrs[k-2] + ARparameters[3]*corrs[k-3].

Value

A banded correlation [matrix](#) whose elements follow an ar3 pattern.

See Also

[mat.I](#), [mat.J](#), [mat.banded](#), [mat.exp](#), [mat.ar1](#), [mat.ar2](#), [mat.sar2](#), [mat.ma1](#), [mat.ma2](#), [mat.arma](#)

Examples

```
corr <- mat.ar3(ARparameters = c(0.4, 0.2, 0.1), order = 4)
```

mat.arma   *Forms an arma correlation matrix*

Description

Form the correlation matrix of order order whose correlations follow the arma pattern. The resulting matrix is banded.

Usage

```
mat.arma(ARparameter, MAparameter, order)
```

Arguments

ARparameter	A numeric value for the autoregressive parameter of the process, being the weight given to the lag 1 response values.
MAparameter	A numeric value for the moving average parameter of the process, being the weight given to the lag 1 random variable.
order	The order of the matrix to be formed.

Details

The correlations in the correlation matrix, corr say, are calculated from the correlation parameters, ARparameters. The values in

- the diagonal (k = 1) of corr are one;
- the first subdiagonal band (k = 2) of corr are equal to ARparameters[1]/(1-ARparameters[2]);
- in subsequent disgonal bands, (k = 3:order), of corr are ARparameters[1]*corr[k-1] + ARparameters[2]*corr[k-2].

Value

A banded correlation matrix whose elements follow an arma pattern.

See Also

mat.I, mat.J, mat.exp, mat.banded, mat.ar1, mat.ar3, mat.sar2, mat.ma1, mat.ma2

Examples

```
corr <- mat.arma(ARparameter = 0.4, MAparameter = -0.2, order = 4)
```

---

mat.banded   *Form a banded matrix from a vector of values*

Description

Takes the first value in x and places it down the diagonal of the matrix. Takes the second value in x and places it down the first subdiagonal, both below and above the diagonal of the matrix. The third value is placed in the second subdiagonal and so on, until the bands for which there are elements in x have been filled. All other elements in the matrix will be zero.

Usage

```
mat.banded(x, nrow, ncol)
```

Arguments

x	A numeric containing the values for each band from 1 to the length of x.
nrow	The number of rows in the banded matrix being formed.
ncol	The number of columns in the banded matrix being formed.

Value

An $nrow \times ncol$ matrix.

See Also

matmult, mat.ar1, mat.ar2, mat.ar3, mat.sar2, mat.exp, mat.ma1, mat.ma2, mat.arma mat.I, mat.J

Examples

```
        m <- mat.banded(c(1,0.6,0.5), 5,5)
        m <- mat.banded(c(1,0.6,0.5), 3,4)
        m <- mat.banded(c(1,0.6,0.5), 4,3)
```

---

mat.dirprod   *Forms the direct product of two matrices*

Description

Form the direct product of the $m \times n$ matrix **A** and the $p \times q$ matrix **B**. It is also called the Kroneker product and the right direct product. It is defined to be the result of replacing each element of **A**, a_{ij}, with $a_{ij}\boldsymbol{B}$. The result matrix is $mp \times nq$.

The method employed uses the rep function to form two $mp \times nq$ matrices: (i) the direct product of **A** and **J**, and (ii) the direct product of **J** and **B**, where each **J** is a matrix of ones whose dimensions are those required to produce an $mp \times nq$ matrix. Then the elementwise product of these two matrices is taken to yield the result.

Usage

```
mat.dirprod(A, B)
```

Arguments

A	The left-hand matrix in the product.
B	The right-hand matrix in the product.

Value

An $mp \times nq$ matrix.

See Also

matmult, mat.dirprod

Examples

```
col.I <- mat.I(order=4)
row.I <- mat.I(order=28)
V <- mat.dirprod(col.I, row.I)
```

---

mat.dirsum   *Forms the direct sum of a list of matrices*

Description

The direct sum is the partitioned matrices whose diagonal submatrices are the matrices from which the direct sum is to be formed and whose off-diagonal submatrices are conformable matrices of zeroes. The resulting matrix is $m \times n$, where m is the sum of the numbers of rows of the contributing matrices and n is the sum of their numbers of columns.

Usage

```
mat.dirsum(matrices)
```

Arguments

matrices	A list, each of whose component is a matrix.

Value

An $m \times n$ matrix.

See Also

mat.dirprod, matmult

Examples

```
m1 <- matrix(1:4, nrow=2)
m2 <- matrix(11:16, nrow=3)
m3 <- diag(1, nrow=2, ncol=2)
dsum <- mat.dirsum(list(m1, m2, m3))
```

---

mat.exp   *Forms an exponential correlation matrix*

Description

Form the correlation [matrix](#) of order equal to the length of `coordinates`. The [matrix](#) has diagonal elements equal to one and the off-diagonal element in the ith row and jth column equal to ρ^k where $k = |coordinate[i] - coordinate[j]|$.

Usage

```
mat.exp(rho, coordinates)
```

Arguments

rho	The correlation for points a distance of one apart.
coordinates	The coordinates of points whose correlation [matrix](#) is to be formed.

Value

A correlation [matrix](#) whose elements depend on the power of the absolute distance apart.

See Also

[mat.I](#), [mat.J](#), [mat.banded](#), [mat.ar1](#), [mat.ar2](#), [mat.ar3](#), [mat.sar2](#), [mat.ma1](#), [mat.ma2](#), [mat.arma](#)

Examples

```
corr <- mat.exp(coordinates=c(3:6, 9:12, 15:18), rho=0.1)
```

---

mat.I   *Forms a unit matrix*

Description

Form the unit or identity [matrix](#) of order `order`.

Usage

```
mat.I(order)
```

Arguments

order	The order of the [matrix](#) to be formed.

Value

A square [matrix](#) whose diagonal elements are one and its off-diagonal are zero.

See Also

mat.J, mat.ar1

Examples

```
col.I <- mat.I(order=4)
```

---

mat.J	*Forms a square matrix of ones*

Description

Form the square matrix of ones of order order.

Usage

```
mat.J(order)
```

Arguments

order   The order of the matrix to be formed.

Value

A square matrix all of whose elements are one.

See Also

mat.I, mat.ar1

Examples

```
col.J <- mat.J(order=4)
```

---

mat.ma1	*Forms an ma1 correlation matrix*

Description

Form the correlation matrix of order order whose correlations follow the ma1 pattern. The matrix is banded and has diagonal elements equal to one and subdiagonal element equal to -MAparameter / (1 + MAparameter*MAparameter).

Usage

```
mat.ma1(MAparameter, order)
```

Arguments

MAparameter	The moving average parameter, being the weight applied to the lag 1 random pertubation.
order	The order of the [matrix](#) to be formed.

Value

A banded correlation [matrix](#) whose elements follow an ma1 pattern.

See Also

[mat.I](#), [mat.J](#), [mat.exp](#), [mat.banded](#), [mat.ar2](#), [mat.ar3](#), [mat.sar2](#), [mat.ma2](#), [mat.arma](#)

Examples

```
corr <- mat.ma1(MAparameter=0.4, order=4)
```

---

mat.ma2   *Forms an ma2 correlation matrix*

Description

Form the correlation [matrix](#) of order order whose correlations follow the ma2 pattern. The resulting [matrix](#) is banded.

Usage

```
mat.ma2(MAparameters, order)
```

Arguments

MAparameters	A [numeric](#) containing the two moving average parameter values of the process, being the weights given to the lag 1 and lag 2 random pertubations.
order	The order of the [matrix](#) to be formed.

Details

The correlations in the correlation matrix, corr say, are calculated from the moving average parameters, MAparameters. The values in

- the diagonal (k = 1) of corr are one;
- the first subdiagonal band (k = 2) of corr are equal to
  -MAparameters[1]*(1 - MAparameters[2]) / div;
- the second subdiagonal bande (k = 3) of corr are equal to -MAparameters[2] / div;
- in subsequent disgonal bands, (k = 4:order), of corr are zero,

where div = 1 + MMAparameters[1]*MAparameter[1] + MAparameters[2]*MAparameters[2].

Value

A banded correlation [matrix](#) whose elements follow an ma2 pattern.

See Also

mat.I, mat.J, mat.exp, mat.banded, mat.ar1, mat.ar3, mat.sar2, mat.ma1, mat.arma

Examples

```
corr <- mat.ma2(MAparameters = c(0.4, -0.2), order = 4)
```

---

mat.sar2   *Forms an sar2 correlation matrix*

Description

Form the correlation matrix of order order whose correlations follow the sar2 pattern, a pattern used in crop competition models. The resulting matrix is banded and is a constrained AR3 matrix.

Usage

```
mat.sar2(gamma, order, print = NULL)
```

Arguments

gamma	A numeric containing the two values of gamma, being parameters linked with spatial dependence and competition.
order	The order of the matrix to be formed.
print	A character giving the object to be printed. Currently, only the claculated values of the ar3parameters can be printed. If NULL, nothing is printed.

Details

The values of the AR3 parameters, phi, are calculated from the gammas as follows:
phi[1] = gamma[1] + 2 * gamma[2]; phi[2] = -gamma[2] * (2*gamma[2] + gamma[1]);
phi[3] = gamma[1] * gamma[2] * gamma[2].

Then the correlations in the correlation matrix, corr say, are calculated from the correlation parameters, phi. Let omega = 1 - phi[2] - phi[3] * (phi[1] + phi[3]). Then the values in

- the diagonal of corr (k = 1) are one;
- the first subdiagonal band (k = 2) of corr are equal to (phi[1] + phi[2]*phi[3]) / omega;
- the second subdiagonal band (k = 3) of corr are equal to
  (phi[1] * (phi[1] + phi[3]) + phi[2] * (1 - phi[2])) / omega;
- the subsequent subdiagonal bands, (k = 4:order), of corr are equal to
  phi[1]*corrs[k-1] + phi[2]*corrs[k-2] + phi[3]*corrs[k-3].

Value

A banded correlation matrix whose elements follow an sar2 pattern.

See Also

mat.I, mat.J, mat.banded, mat.exp, mat.ar1, mat.ar2, mat.ar3, mat.ma1, mat.ma2, mat.arma

Examples

```
corr <- mat.sar2(gamma = c(-0.4, 0.2), order = 4)
corr <- mat.sar2(gamma = c(-0.4, 0.2), order = 4, print = "ar3")
```

---

mat.Vpred	*Calculates the variances of a set of predicted effects from a mixed model*

Description

Calculates the variances of a set of predicted effects, given the incidence matrix for the effects to be predicted and, optionally, a variance matrix of the effects, a incidence matrix for the nuisance fixed factors and covariates, the variance matrix of the nuisance random effects in the mixed model and the residual variance matrix.

Usage

```
mat.Vpred(W, Gg = 0, X = matrix(1, nrow = nrow(W), ncol = 1), Vu = 0, R)
```

Arguments

W	The incidence [matrix](#) for the effects to be predicted.
Gg	The variance [matrix](#) of the effects to be predicted. If the effects to be predicted are fixed, set to 0.
X	The incidence [matrix](#) for the nuisance fixed factors and covariates. The default is a column vector of ones.
Vu	The variance [matrix](#) of the nuisance random effects. If there are none, set to zero.
R	The residual variance [matrix](#).

Details

The matrix is calculated as
Vpred = t(W) %*% Vinv %*% W + solve(Gg) - A%*%solve(t(X)%*%Vinv%*%X)%*%t(A),
where Vinv = solve(Vu + R) and A = t(W) %*% Vinv %*% X.

Value

A [matrix](#) containing the varainces and covariances of the predictions.

Author(s)

Chris Brien

References

Smith, A. B., D. G. Butler, C. R. Cavanagh and B. R. Cullis (2015). Multi-phase variety trials using both composite and individual replicate samples: a model-based design approach. *Journal of Agricultural Science*, **153**, 1017-1029.

See Also

Ameasures.

Examples

```
## Reduced example from Smith et al. (2015)
## Generate two-phase design
mill.fac <- fac.gen(list(Mrep = 2, Mday = 2, Mord = 3))
field.lay <- fac.gen(list(Frep = 2, Fplot = 4))
field.lay$Variety <- factor(c("D","E","Y","W","G","D","E","M"),
                            levels = c("Y","W","G","M","D","E"))
start.design <- cbind(mill.fac, field.lay[c(3,4,5,8,1,7,3,4,5,8,6,2),])
rownames(start.design) <- NULL

## Set up matrices
W <- model.matrix(~ -1+ Variety, start.design)
ng <- ncol(W)
Gg<- diag(1, ng)
Vu <- with(start.design, fac.vcmat(Mrep, 0.3) +
                          fac.vcmat(fac.combine(list(Mrep, Mday)), 0.2) +
                          fac.vcmat(Frep, 0.1) +
                          fac.vcmat(fac.combine(list(Frep, Fplot)), 0.2))
R <- diag(1, nrow(start.design))

## Caluclate information matrix
Vp <- mat.Vpred(W = W, Gg = Gg, Vu = Vu, R = R)
```

---

meanop *computes the projection matrix that produces means*

Description

Replaced by fac.meanop.

---

mpone *Converts the first two levels of a factor into the numeric values -1 and +1*

Description

Converts the first two levels of a factor into the numeric values -1 and +1.

Usage

```
mpone(factor)
```

Arguments

factor   The factor to be converted.

Value

A numeric `vector`.

Warning

If the [factor](#) has more than two `levels` they will be coerced to numeric values.

Author(s)

Chris Brien

See Also

[mpone](#) in package **dae**, [factor](#), [relevel](#).

Examples

```
## generate all combinations of two two-level factors
mp <- c("-", "+")
Frf3.trt <- fac.gen(list(A = mp, B = mp))

## add factor C, whose levels are the products of the levles of A and B
Frf3.trt$C <- factor(mpone(Frf3.trt$A)*mpone(Frf3.trt$B), labels = mp)
```

---

no.reps   *Computes the number of replicates for an experiment*

Description

Computes the number of pure replicates required in an experiment to achieve a specified power.

Usage

```
no.reps(multiple=1., df.num=1.,
        df.denom=expression((df.num + 1.) * (r - 1.)), delta=1.,
        sigma=1., alpha=0.05, power=0.8, tol=0.1, print=FALSE)
```

Arguments

multiple	The multiplier, m, which when multiplied by the number of pure replicates of a treatment, r, gives the number of observations rm used in computing means for some, not necessarily proper, subset of the treatment factors; m is the replication arising from other treatment factors. However, for single treatment factor experiments the subset can only be the treatment factor and m = 1.
df.num	The degrees of freedom of the numerator of the F for testing the term involving the treatment factor subset.
df.denom	The degrees of freedom of the denominator of the F for testing the term involving the treatment factor subset.
delta	The true difference between a pair of means for some, not necessarily proper, subset of the treatment factors.
sigma	The population standard deviation.

alpha	The significance level to be used.
power	The minimum power to be achieved.
tol	The maximum difference tolerated between the power required and the power computed in determining the number of replicates.
print	TRUE or FALSE to have or not have a table of power calculation details printed out.

Value

A list containing nreps, a single numeric value containing the computed number of pure replicates, and power, a single numeric value containing the power for the computed number of pure replicates.

Author(s)

Chris Brien

See Also

power.exp, detect.diff in package **dae**.

Examples

```
## Compute the number of replicates (blocks) required for a randomized
## complete block design with four treatments.
no.reps(multiple = 1, df.num = 3,
        df.denom = expression(df.num * (r - 1)), delta = 5,
      sigma = sqrt(20), print = TRUE)
```

---

power.exp	*Computes the power for an experiment*

Description

Computes the power for an experiment.

Usage

```
power.exp(rm=5., df.num=1., df.denom=10., delta=1., sigma=1.,
             alpha=0.05, print=FALSE)
```

Arguments

rm	The number of observations used in computing a mean.
df.num	The degrees of freedom of the numerator of the F for testing the term involving the means.
df.denom	The degrees of freedom of the denominator of the F for testing the term involving the means.
delta	The true difference between a pair of means.
sigma	The population standard deviation.
alpha	The significance level to be used.
print	TRUE or FALSE to have or not have a table of power calculation details printed out.

Value

A single `numeric` value containing the computed power.

Author(s)

Chris Brien

See Also

[no.reps](), [detect.diff]() in package **dae**.

Examples

```
## Compute power for a randomized complete block design with four treatments
## and five blocks.
rm <- 5
power.exp(rm = rm, df.num = 3, df.denom = 3 * (rm - 1), delta = 5,
          sigma = sqrt(20),print = TRUE)
```

---

print.projector   *Print projectors*

Description

Print an object of class "[projector]()", displaying the matrix and its degrees of freedom (rank).

Usage

```
## S3 method for class 'projector'
print(x, ...)
```

Arguments

x	The object of class "[projector]()" to be printed.
...	Further arguments passed to or from other methods.

Author(s)

Chris Brien

See Also

[print](), [print.default](), [show]().

[projector]() for further information about this class.

Examples

```
## set up a 2 x 2 mean operator that takes the mean of a vector of 2 values
m <- matrix(rep(0.5,4), nrow=2)

## create an object of class projector
proj.m <- projector(m)

## print the object either using the Method function, the generic function or show
print.projector(proj.m)
print(proj.m)
proj.m
```

print.summary.p2canon   *Prints the values in an* summary.p2canon *object*

Description

Prints a `summary.p2canon` object, which is also a `data.frame`, in a pretty format.

Usage

```
## S3 method for class 'summary.p2canon'
print(x, ...)
```

Arguments

x	A summary.p2canon object.
...	further arguments passed to print.

Value

No value is returned.

See Also

summary.p2canon

Examples

```
## PBIBD(2) from p. 379 of Cochran and Cox (1957) Experimental Designs.
## 2nd edn Wiley, New York
PBIBD2.unit <- list(Block = 6, Unit = 4)
PBIBD2.nest <- list(Unit = "Block")
trt <- factor(c(1,4,2,5, 2,5,3,6, 3,6,1,4, 4,1,5,2, 5,2,6,3, 6,3,4,1))
PBIBD2.lay <- fac.layout(unrandomized = PBIBD2.unit,
                         nested.factors=PBIBD2.nest,
                         randomized = trt)

##obtain projectors using projs.structure
Q.unit <- projs.structure(~ Block/Unit, data = PBIBD2.lay)
Q.trt <- projs.structure(~ trt, data = PBIBD2.lay)

##obtain combined decomposition and print summary
```

```
unit.trt.p2canon <- projs.2canon(Q.unit, Q.trt)
summ <- summary(unit.trt.p2canon)
print(summ)
```

---

print.summary.pcanon   *Prints the values in an* summary.pcanon *object*

Description

Prints a summary.pcanon object, which is also a data.frame, in a pretty format.

Usage

```
## S3 method for class 'summary.pcanon'
print(x, ...)
```

Arguments

x   A summary.pcanon object.

...   further arguments passed to print.

Value

No value is returned.

See Also

summary.pcanon

Examples

```
## PBIBD(2) from p. 379 of Cochran and Cox (1957) Experimental Designs.
## 2nd edn Wiley, New York
PBIBD2.unit <- list(Block = 6, Unit = 4)
PBIBD2.nest <- list(Unit = "Block")
trt <- factor(c(1,4,2,5, 2,5,3,6, 3,6,1,4, 4,1,5,2, 5,2,6,3, 6,3,4,1))
PBIBD2.lay <- fac.layout(unrandomized = PBIBD2.unit,
                         nested.factors=PBIBD2.nest,
                         randomized = trt)

##obtain combined decomposition and summarize
unit.trt.canon <- projs.canon(list(unit=~ Block/Unit, trt=~ trt), data = PBIBD2.lay)
summ <- summary(unit.trt.canon, which = c("aeff","eeff","order"))
print(summ)
```

---

proj2.combine	*Compute the projection and Residual operators for two, possibly nonorthogonal, projectors*

Description

The canonical relationship between a pair of projectors is established by decomposing the range of Q1 into a part that pertains to Q2 and a part that is orthogonal to Q2. It also produces the nonzero canonical efficiency factors for the joint decomposition of Q1 and Q and the corresponding eigenvectors of Q1 (James and Wilkinson, 1971). Q1 and Q2 may be nonorthogonal.

Usage

```
proj2.combine(Q1, Q2)
```

Arguments

Q1   A symmetric `projector` whose range is to be decomposed.

Q2   A symmetric `projector` whose range in Q1 is required.

Details

The nonzero canonical efficiency factors are the nonzero eigenvalues of Q1 %*% Q2 %*% Q1 (James and Wilkinson, 1971). An eigenvalue is regarded as zero if it is less than daeTolerance, which is initially set to .Machine$double.eps ^ 0.5 (about 1.5E-08). The function `set.daeTolerance` can be used to change daeTolerance.

The eigenvectors are the eigenvectors of Q1 corresponding to the nonzero canonical efficiency factors. The eigenvectors for Q2 can be obtained by premultiplying those for Q1 by Q2.

Qres is computed using equation 4.10 from James and Wilkinson (1971), if the number of distinct canonical efficiency factors is less than 10. If this fails to produce a projector or the number of distinct canonical efficiency factors is 10 or more, equation 5.3 of Payne and Tobias (1992) is used to obtain Qres. In this latter case, Qres = Q1 - Q1 %*% ginv(Q2 %*% Q1 %*% Q2) %*% Q1. Qconf is obtained by subtracting Qres from Q1.

Value

A `list` with the following components:

1. **efficiencies:** a `vector` containing the nonzero canonical efficiency factors;
2. **eigenvectors:** an n x r `matrix`, where n is the order of the projectors and r is the number of nonzero canonical efficiency factors; it contains the eigenvectors of Q1 corresponding to the nonzero canonical efficiency factors.
3. **Qconf:** a `projector` onto the part of the range of Q1 with which Q2 is confounded;
4. **Qres:** a `projector` onto the part of the range of Q1 that is orthogonal to the range of Q2.

Author(s)

Chris Brien

References

James, A. T. and Wilkinson, G. N. (1971) Factorization of the residual operator and canonical decomposition of nonorthogonal factors in the analysis of variance. *Biometrika*, **58**, 279–294.

Payne, R. W. and R. D. Tobias (1992). General balance, combination of information and the analysis of covariance. *Scandinavian Journal of Statistics*, **19**, 3–23.

See Also

`proj2.eigen`, `proj2.efficiency`, `decomp.relate` in package **dae**.

`projector` for further information about this class.

Examples

```
## PBIBD(2) from p. 379 of Cochran and Cox (1957) Experimental Designs.
## 2nd edn Wiley, New York
PBIBD2.unit <- list(Block = 6, Unit = 4)
PBIBD2.nest <- list(Unit = "Block")
trt <- factor(c(1,4,2,5, 2,5,3,6, 3,6,1,4, 4,1,5,2, 5,2,6,3, 6,3,4,1))
PBIBD2.lay <- fac.layout(unrandomized = PBIBD2.unit,
                         nested.factors=PBIBD2.nest,
                         randomized = trt)

## obtain sets of projectors
Q.unit <- projs.structure(~ Block/Unit, data = PBIBD2.lay)
Q.trt <- projs.structure(~ trt, data = PBIBD2.lay)

## obtain the projection operators for the interblock analysis
PBIBD2.Bops <- proj2.combine(Q.unit[["Block:Unit"]], Q.trt[["trt"]])
Q.B.T <- PBIBD2.Bops$Qconf
Q.B.res <- PBIBD2.Bops$Qres

## demonstrate their orthogonality
is.allzero(Q.B.T %*% Q.B.res)
```

---

proj2.efficiency	*Computes the canonical efficiency factors for the joint decomposition of two projectors*

Description

Computes the canonical efficiency factors for the joint decomposition of two projectors (James and Wilkinson, 1971).

Usage

```
proj2.efficiency(Q1, Q2)
```

Arguments

Q1   An object of class "`projector`".

Q2   An object of class "`projector`".

Details

The nonzero canonical efficiency factors are the nonzero eigenvalues of Q1 *%*%* Q2 *%*%* Q1 (James and Wilkinson, 1971). An eigenvalue is regarded as zero if it is less than daeTolerance, which is initially set to .Machine$double.eps ^ 0.5 (about 1.5E-08). The function `set.daeTolerance` can be used to change daeTolerance.

Value

A vector containing the nonzero canonical efficiency factors.

Author(s)

Chris Brien

References

James, A. T. and Wilkinson, G. N. (1971) Factorization of the residual operator and canonical decomposition of nonorthogonal factors in the analysis of variance. *Biometrika*, **58**, 279-294.

See Also

`efficiency.criteria`, `proj2.eigen`, `proj2.combine` in package **dae**, `eigen`.

`projector` for further information about this class.

Examples

```
## PBIBD(2) from p. 379 of Cochran and Cox (1957) Experimental Designs.
## 2nd edn Wiley, New York
PBIBD2.unit <- list(Block = 6, Unit = 4)
PBIBD2.nest <- list(Unit = "Block")
trt <- factor(c(1,4,2,5, 2,5,3,6, 3,6,1,4, 4,1,5,2, 5,2,6,3, 6,3,4,1))
PBIBD2.lay <- fac.layout(unrandomized = PBIBD2.unit,
                         nested.factors=PBIBD2.nest,
                         randomized = trt)

## obtain sets of projectors
Q.unit <- projs.structure(~ Block/Unit, data = PBIBD2.lay)
Q.trt <- projs.structure(~ trt, data = PBIBD2.lay)

## save intrablock efficiencies
eff.intra <- proj2.efficiency(Q.unit[["Block"]], Q.trt[["trt"]])
```

---

proj2.eigen	*Canonical efficiency factors and eigenvectors in joint decomposition of two projectors*

Description

Computes the canonical efficiency factors for the joint decomposition of two projectors and the eigenvectors corresponding to the first projector (James and Wilkinson, 1971).

Usage

```
proj2.eigen(Q1, Q2)
```

Arguments

Q1   An object of class "`projector`".

Q2   An object of class "`projector`".

Details

The component efficiencies is a `vector` containing the nonzero canonical efficiency factors for the joint decomposition of the two projectors. The nonzero canonical efficiency factors are the nonzero eigenvalues of Q1 %*% Q2 %*% Q1 (James and Wilkinson, 1971). An eigenvalue is regarded as zero if it is less than daeTolerance, which is initially set to .Machine$double.eps ^ 0.5 (about 1.5E-08). The function `set.daeTolerance` can be used to change daeTolerance.

The component eigenvectors is an n x r `matrix`, where n is the order of the projectors and r is the number of nonzero canonical efficiency factors; it contains the eigenvectors of Q1 corresponding to the nonzero canonical efficiency factors. The eigenvectors for Q2 can be obtained by premultiplying those for Q1 by Q2.

Value

A `list` with components efficiencies and eigenvectors.

Author(s)

Chris Brien

References

James, A. T. and Wilkinson, G. N. (1971) Factorization of the residual operator and canonical decomposition of nonorthogonal factors in the analysis of variance. *Biometrika*, **58**, 279-294.

See Also

`proj2.efficiency`, `proj2.combine` in package **dae**, `eigen`.

`projector` for further information about this class.

Examples

```
## PBIBD(2) from p. 379 of Cochran and Cox (1957) Experimental Designs.
## 2nd edn Wiley, New York
PBIBD2.unit <- list(Block = 6, Unit = 4)
PBIBD2.nest <- list(Unit = "Block")
trt <- factor(c(1,4,2,5, 2,5,3,6, 3,6,1,4, 4,1,5,2, 5,2,6,3, 6,3,4,1))
PBIBD2.lay <- fac.layout(unrandomized = PBIBD2.unit,
                         nested.factors=PBIBD2.nest,
                         randomized = trt)

## obtain sets of projectors
Q.unit <- projs.structure(~ Block/Unit, data = PBIBD2.lay)
Q.trt <- projs.structure(~ trt, data = PBIBD2.lay)

## obtain intra- and inter-block decompositions
```

```
decomp.inter <- proj2.eigen(Q.unit[["Block"]], Q.trt[["trt"]])
decomp.intra <- proj2.eigen(Q.unit[["Block:Unit"]], Q.trt[["trt"]])

#extract intrablock efficiencies
decomp.intra$efficiencies
```

---

projector   *Create projectors*

Description

The class "projector" is the subclass of the class "matrix" in which matrices are square, symmetric and idempotent.

The function projector tests whether a matrix satisfies these criteria and if it does creates a "projector" object, computing the projector's degrees of freedom and adding them to the object.

Usage

```
projector(Q)
```

Arguments

Q   The matrix to be made into a projector.

Details

In checking that the matrix is square, symmetric and idempotent, the equality of the matrix with either its transpose or square is tested. In this, a difference in elements is considered to be zero if it is less than daeTolerance, which is initially set to .Machine$double.eps ^ 0.5 (about 1.5E-08). The function set.daeTolerance can be used to change daeTolerance.

Value

An object of Class "projector" that consists of a square, summetric, idempotent matrix and degrees of freedom (rank) of the matrix.

Author(s)

Chris Brien

See Also

degfree, correct.degfree in package **dae**.

projector for further information about this class.

Examples

```
## set up a 2 x 2 mean operator that takes the mean of a vector of 2 values
m <- matrix(rep(0.5,4), nrow=2)

## create an object of class projector
proj.m <- projector(m)

## check that it is a valid projector
is.projector(proj.m)
```

---

projector-class   *Class projector*

Description

The class "projector" is the subclass of matrices that are square, symmetric and idempotent.

is.projector is the membership function for this class.

degfree is the extractor function for the degrees of freedom and degfree<- is the replacement function.

correct.degfree checks whether the stored degrees of freedom are correct.

Objects from the Class

An object of class "projector" consists of a square, symmetric, idempotent matrix along with its degrees of freedom (rank).

Objects can be created by calls of the form new("projector", data, nrow, ncol, byrow, dimnames, ...). However, this does not add the degrees of freedom to the object. These can be added using the replacement function degfree<-. Alternatively, the function projector creates the new object from a matrix, adding its degrees of freedom at the same time.

Slots

.Data: Object of class "matrix"

degfree: Object of class "integer"

Extends

Class "matrix", from data part. Class "array", by class "matrix", distance 2. Class "structure", by class "matrix", distance 3. Class "vector", by class "matrix", distance 4, with explicit coerce.

Methods

coerce signature(from = "projector", to = "matrix")

print signature(x = "projector")

show signature(object = "projector")

Author(s)

Chris Brien

See Also

[projector](), [degfree](), [correct.degfree]() in package **dae**.

Examples

```
showClass("projector")

## set up a 2 x 2 mean operator that takes the mean of a vector of 2 values
m <- matrix(rep(0.5,4), nrow=2)

## create an object of class projector
proj.m <- projector(m)

## check that it is a valid projector
is.projector(proj.m)

## create a projector based on the matrix m
proj.m <- new("projector", data=m)

## add its degrees of freedom and print the projector
degfree(proj.m) <- proj.m
```

---

projs.2canon	*A canonical analysis of the relationships between two sets of projec-tors*

Description

Computes the canonical efficiency factors for the joint decomposition of two structures or sets of mutually orthogonally projectors (Brien and Bailey, 2009), orthogonalizing projectors in the Q2 list to those earlier in the list of projectors with which they are partially aliased. The results can be summarized in the form of a skeleton ANOVA table.

Usage

```
projs.2canon(Q1, Q2)
```

Arguments

Q1	A list whose components are objects of class "[projector]()".
Q2	A list whose components are objectsof class "[projector]()".

Details

Two loops, one nested within the other. are performed. The first cycles over the components of Q1 and the nested loop cycles over the components of Q2. The joint decomposition of the two projectors in each cycle, one from Q1 (say Q1[[i]]) and the other from Q2 (say Q2[[j]]) is obtained using [proj2.combine](). In particular, the nonzero canonical efficiency factors for the joint decomposition of the two projectors is obtained. The nonzero canonical efficiency factors are the nonzero eigenvalues of Q1[[i]] %*% Q2[[j]] %*% Q1[[i]] (James and Wilkinson, 1971). An eigenvalue is regarded as zero if it is less than daeTolerance, which is initially set to

.Machine$double.eps ^ 0.5 (about 1.5E-08). The function set.daeTolerance can be used to change daeTolerance.

However, a warning occurs if any pair of Q2 projectors (say Q2[[j]] and Q2[[k]]) do not have adjusted orthgonality with respect to any Q1 projector (say Q1[[i]]), because they are partially aliased. That is, if Q2[[j]] %*% Q1[[i]] %*% Q2[[k]] is nonzero for any pair of different Q2 projectors and any Q1 projector. When it is nonzero, the projector for the later term in the list of projectors is orthogonalized to the projector that is earlier in the list.

Value

A list of class p2canon. It has a component for each component of Q1. Each of the components for Q1 is a list; its components are one for each component of Q2 and a component Pres. Each of the Q2 components is a list of three components: pairwise, adjusted and Qproj. These components are based on an eigenalysis of the relationship between the projectors for the parent Q1 and Q2 components. Each pairwise component is based on the nonzero canonical efficiency factors for the joint decomposition of the two parent projectors (see proj2.eigen). An adjusted component is based on the nonzero canonical efficiency factors for the joint decomposition of the Q1 component and the Q2 component, the latter adjusted for all Q2 projectors that have occured previously in the list. The Qproj component is the adjusted projector for the parent Q2 component. The pairwise and adjusted components have the following components: efficiencies, aefficiency, mefficiency, sefficiency, eefficiency, xefficiency, order and dforthog – for details see efficiency.criteria.

Author(s)

Chris Brien

References

Brien, C. J. and R. A. Bailey (2009). Decomposition tables for multitiered experiments. I. A chain of randomizations. *The Annals of Statistics*, **36**, 4184 - 4213.

James, A. T. and Wilkinson, G. N. (1971) Factorization of the residual operator and canonical decomposition of nonorthogonal factors in the analysis of variance. *Biometrika*, **58**, 279-294.

See Also

summary.p2canon, efficiencies.p2canon, projs.combine.p2canon, projs.structure , proj2.efficiency, proj2.combine, proj2.eigen, efficiency.criteria in package **dae**, eigen. projector for further information about this class.

Examples

```
## PBIBD(2) from p. 379 of Cochran and Cox (1957) Experimental Designs.
## 2nd edn Wiley, New York
PBIBD2.unit <- list(Block = 6, Unit = 4)
PBIBD2.nest <- list(Unit = "Block")
trt <- factor(c(1,4,2,5, 2,5,3,6, 3,6,1,4, 4,1,5,2, 5,2,6,3, 6,3,4,1))
PBIBD2.lay <- fac.layout(unrandomized = PBIBD2.unit,
                         nested.factors=PBIBD2.nest,
                         randomized = trt)

##obtain projectors using projs.structure
Q.unit <- projs.structure(~ Block/Unit, data = PBIBD2.lay)
Q.trt <- projs.structure(~ trt, data = PBIBD2.lay)
```

```
##obtain combined decomposition and summarize
unit.trt.p2canon <- projs.2canon(Q.unit, Q.trt)
summary(unit.trt.p2canon)
```

---

projs.canon   *A canonical analysis of the relationships between sets of projectors*

Description

Computes the canonical efficiency factors for the joint decomposition of two or more structures or sets of mutually orthogonally projectors (Brien and Bailey, 2009), orthogonalizing projectors in a set to those earlier in the set of projectors with which they are partially aliased. The results can be summarized in the form of a skeleton ANOVA table.

Usage

```
projs.canon(formulae, orthogonalize = "differencing",
            which.criteria = c("aefficiency","eefficiency","order"),
            omit.projectors = c("p2canon", "combined"), data = NULL, ...)
```

Arguments

formulae	An object of class list whose compoents are of class formula.
orthogonalize	A string indicating the method for orthogonalizing a projector to those for terms that occurred previously in the formula. Two options are available: differencing; eigenmethods. If a single value is given, it is used for all formulae.
which.criteria	A character vector nominating the efficiency criteria to be included in the summary of aliasing between terms within a structure. It can be none, all or some combination of aefficiency, mefficiency, sefficiency, eefficiency, xefficiency, order and dforthog – for details see efficiency.criteria. If none, no summary is printed.
omit.projectors	A character vector of the types of projectors to omit from the returned pcanon object. If pcanon is included in the vector then the projectors in these objects will be replaced with a numeric containing their degrees of freedom. If combined is included in the vector then the projectors for the combined decomposition will be replaced with a numeric containing their degrees of freedom. If none is included in the vector then no projectors will be omitted.
data	A data frame contains the values of the factors and variables that occur in formulae.
...	further arguments passed to terms.

Details

For each formula supplied in formulae, the set of projectors is obtained using projs.structure; there is one projector for each term in a formula. Then projs.2canon is used to perform an analysis of the canonical relationships between two sets of projectors for the first two formulae. If there are further formulae, the relationships between its projectors and the already established decomposition is obtained using projs.2canon. The core of the analysis is the determination of eigenvalues of the product of pairs of projectors using the results of James and Wilkinson (1971).

Value

A `list` of class pcanon. It has as many components as there are formulae. The first component is
the joint decomposition of two structures derived from the first two formulae, being the p2canon
object produced by projs.2canon. Then there is a component for each further formulae that con-
tains the p2canon object obtained by applying projs.2canon to the structure for a formula and the
already established joint decomposition of the structures for the previous formulae in the `formulae`.
The last component contains the the `list` of the projectors that give the combined canonical de-
composition derived from the `formulae`.

Author(s)

Chris Brien

References

Brien, C. J. and R. A. Bailey (2009). Decomposition tables for multitiered experiments. I. A chain
of randomizations. *The Annals of Statistics*, **36**, 4184 - 4213.

James, A. T. and Wilkinson, G. N. (1971) Factorization of the residual operator and canonical
decomposition of nonorthogonal factors in the analysis of variance. *Biometrika*, **58**, 279-294.

See Also

summary.pcanon, efficiencies.pcanon, projs.2canon, projs.structure ,
proj2.efficiency, proj2.combine, proj2.eigen, efficiency.criteria, in package **dae**, eigen.

projector for further information about this class.

Examples

```
## PBIBD(2) from p. 379 of Cochran and Cox (1957) Experimental Designs.
## 2nd edn Wiley, New York
PBIBD2.unit <- list(Block = 6, Unit = 4)
PBIBD2.nest <- list(Unit = "Block")
trt <- factor(c(1,4,2,5, 2,5,3,6, 3,6,1,4, 4,1,5,2, 5,2,6,3, 6,3,4,1))
PBIBD2.lay <- fac.layout(unrandomized = PBIBD2.unit,
                         nested.factors=PBIBD2.nest,
                         randomized = trt)

##obtain combined decomposition and summarize
unit.trt.canon <- projs.canon(list(unit=~ Block/Unit, trt=~ trt), data = PBIBD2.lay)
summary(unit.trt.canon, which = c("aeff","eeff","order"))

## Three-phase sensory example from Brien and Payne (1999)
## Not run: data(Sensory3Phase.dat)
        Eval.Field.Treat.canon <- projs.canon(list(
                                eval=~ ((Occasions/Intervals/Sittings)*Judges)/Positions,
                                    field=~ (Rows*(Squares/Columns))/Halfplots,
                                    treats=~ Trellis*Method), data=Sensory3Phase.dat)
        summary(Eval.Field.Treat.canon, which.criteria =c("aefficiency", "order"))

## End(Not run)
```

projs.combine.p2canon   *Extract, from a p2canon object, the projectors that give the combined canonical decomposition*

Description

Extracts, from a p2canon object obtained using `projs.2canon`, the projectors that give the combined canonical decomposition of two sets of projectors (Brien and Bailey, 2009).

Usage

```
projs.combine.p2canon(object)
```

Arguments

object   A `list` of class p2canon produced by `projs.2canon`.

Value

A `list`, each of whose components is a projector in the decomposition.

Author(s)

Chris Brien

References

Brien, C. J. and R. A. Bailey (2009). Decomposition tables for multitiered experiments. I. A chain of randomizations. *The Annals of Statistics*, **36**, 4184 - 4213.

See Also

`projs.2canon`, `proj2.eigen`, `proj2.combine` in package **dae**.

`projector` for further information about this class.

Examples

```
## PBIBD(2) from p. 379 of Cochran and Cox (1957) Experimental Designs.
## 2nd edn Wiley, New York
PBIBD2.unit <- list(Block = 6, Unit = 4)
PBIBD2.nest <- list(Unit = "Block")
trt <- factor(c(1,4,2,5, 2,5,3,6, 3,6,1,4, 4,1,5,2, 5,2,6,3, 6,3,4,1))
PBIBD2.lay <- fac.layout(unrandomized = PBIBD2.unit,
                         nested.factors=PBIBD2.nest,
                         randomized = trt)

## obtain sets of projectors
Q.unit <- projs.structure(~ Block/Unit, data = PBIBD2.lay)
Q.trt <- projs.structure(~ trt, data = PBIBD2.lay)

##obtain combined decomposition
unit.trt.p2canon <- projs.2canon(Q.unit, Q.trt)
UcombineT <- projs.combine.p2canon(unit.trt.p2canon)
```

---

projs.structure   *orthogonalized projectors for the terms in a formula*

Description

Produces a set of mutually orthogonal projectors, one for each term in the formula. These are used to specify a structure, or an orthogonal decomposition of the data space. Firstly, the primary projector $\mathbf{X}(\mathbf{X}'\mathbf{X})^{-}\mathbf{X}'$, where X is the design matrix for the term, is calculated for each term. Then each projector is made orthogonal to terms aliased with it, either by `differencing` or `eigenmethods`. `Differencing` relies on comparing the factors involved in two terms to identify which of the terms previous to a term in the expanded formula may be partially aliased with it; `eigenmethods` forces the orthogonal projector for a term to be orthogonal to all terms previous to it in the expanded formula.

Usage

```
projs.structure(formula, orthogonalize = "differencing",
                which.criteria = c("aefficiency","eefficiency","order"),
                data = NULL, ...)
```

Arguments

formula	An object of class `formula` from which the terms will be obtained.
orthogonalize	A string indicating the method for orthogonalizing a projector to those for terms that occurred previously in the formula. Two options are available: `differencing`; `eigenmethods`.
which.criteria	A character `vector` nominating the efficiency criteria to be included in the summary of aliasing between terms. It can be `none`, `all` or some combination of `aefficiency`, `mefficiency`, `sefficiency`, `eefficiency`, `xefficiency`, `order` and `dforthog` – for details see `efficiency.criteria`. If `none`, no summary is printed.
data	A data frame contains the values of the factors and variables that occur in `formula`.
...	further arguments passed to `terms`.

Details

In orthogonalizing a projector to others in the set, the `differencing` method subtracts from a primary projector each orthogonalized projector for a term whose factors/variables are a subset of its factors/variables. This relies on ensuring that all projectors whose factors/variables are a subset of the current projector occur before it in the expanded formula. It is checked that the set of matrices are mutually orthogonal. If they are not then a warning is given. It may happen that differencing does not produce a projector, in which case `eigenmethods` must be used.

The `eigenmethods` uses equation 4.10 of James and Wilkinson (1971), which involves calculating the canonical efficiency factors for pairs of primary projectors. The latter method forces each projector to be orthogonal to all terms previous to it in the expanded formula and produces a table of efficiency criteria for partially aliased terms.

Value

A `list` with a component for each term in the `formula`, the component containing the orthogonalized projector for the term.

Author(s)

Chris Brien

References

James, A. T. and Wilkinson, G. N. (1971) Factorization of the residual operator and canonical decomposition of nonorthogonal factors in the analysis of variance. *Biometrika*, **58**, 279-294.

See Also

`proj2.efficiency`, `proj2.combine`, `proj2.eigen`, `projs.2canon` in package **dae**, `eigen`.

`projector` for further information about this class.

Examples

```
## PBIBD(2) from p. 379 of Cochran and Cox (1957) Experimental Designs.
## 2nd edn Wiley, New York
PBIBD2.unit <- list(Block = 6, Unit = 4)
PBIBD2.nest <- list(Unit = "Block")
trt <- factor(c(1,4,2,5, 2,5,3,6, 3,6,1,4, 4,1,5,2, 5,2,6,3, 6,3,4,1))
PBIBD2.lay <- fac.layout(unrandomized = PBIBD2.unit,
                         nested.factors=PBIBD2.nest,
                         randomized = trt)

## manually obtain projectors for units
Q.G <- projector(matrix(1, nrow=24, ncol=24)/24)
Q.B <- projector(fac.meanop(PBIBD2.lay$Block) - Q.G)
Q.BP <- projector(diag(1, nrow=24) - Q.B - Q.G)

## manually obtain projector for trt
Q.T <- projector(fac.meanop(PBIBD2.lay$trt) - Q.G)

##compute intrablock efficiency criteria
effic <- proj2.efficiency(Q.BP, Q.T)
effic
efficiency.criteria(effic)

##obtain projectors using projs.structure
Q.unit <- projs.structure(~ Block/Unit, data = PBIBD2.lay)
Q.trt <- projs.structure(~ trt, data = PBIBD2.lay)

##obtain combined decomposition and summarize
unit.trt.p2canon <- projs.2canon(Q.unit, Q.trt)
summary(unit.trt.p2canon, which = c("aeff","eeff","order"))
```

---

qqyeffects   *Half or full normal plot of Yates effects*

Description

Produces a half or full normal plot of the Yates effects from a 2^k factorial experiment.

Usage

```
qqyeffects(aov.obj, error.term="Within", data=NULL, pch=16,
           full=FALSE, ...)
```

Arguments

aov.obj	An aov object or aovlistobject created from a call to aov.
error.term	The term from the Error function from which the Yates effects are estimated. Only required when Error used in call to aov.
data	A data.frame in which the variables specified in the aov.obj will be found. If missing, the variables are searched for in the standard way.
pch	The number of a plotting symbol to be drawn when plotting points (use help(points) for details).
full	whether a full or half normal plot is to be produced. The default is for a half-normal plot; full=TRUE produces a full normal plot.
...	Further graphical parameters may be specified (use help(par) for possibilities).

Details

A half or full normal plot of the Yates effects is produced. You will be able to interactively select effects to be labelled (click reasonably close to the point and on the side where you want the label placed). **Right click on the graph and select Stop when you have finished labelling effects.** A regression line fitted to the unselected effects and constrained to go through the origin is plotted. Also, a list of the labelled effects, if any, are printed to standard ouptut.

Value

Returns, invisibly, a list with components x and y, giving coordinates of the plotted points.

Author(s)

Chris Brien

See Also

yates.effects in package **dae**, qqnorm.

Examples

```
## analysis of 2^4 factorial experiment from Table 10.6 of Box, Hunter and
## Hunter (1978) Statistics for Experimenters. New York, Wiley.
## use ?Fac4Proc.dat for data set details
data(Fac4Proc.dat)
Fac4Proc.aov <- aov(Conv ~ Catal * Temp * Press * Conc + Error(Runs),
                                                   Fac4Proc.dat)
qqyeffects(Fac4Proc.aov, error.term="Runs", data=Fac4Proc.dat)
```

---

resid.errors   *Extract the residuals for a fitted model*

Description

An alias for the generic function residuals. When it is available, the method residuals.aovlist extracts residuals, which is provided in the package **dae** to cover aovlist objects.

Usage

```
resid.errors(...)
```

Arguments

...   Arguments passed to residuals.aovlist.

Value

A numeric vector containing the residuals.

Note

See residuals.aovlist for specific information about the residuals when an Error function is used in the call to the aov function.

Author(s)

Chris Brien

See Also

fitted.errors, residuals.aovlist, tukey.1df in package **dae**.

Examples

```
## set up data frame for randomized complete block design in Table 4.4 from
## Box, Hunter and Hunter (2005) Statistics for Experimenters. 2nd edn
## New York, Wiley.
RCBDPen.dat <- fac.gen(list(Blend=5, Flask=4))
RCBDPen.dat$Treat <- factor(rep(c("A","B","C","D"), times=5))
RCBDPen.dat$Yield <- c(89,88,97,94,84,77,92,79,81,87,87,
                       85,87,92,89,84,79,81,80,88)

## perform the analysis of variance
```

```
RCBDPen.aov <- aov(Yield ~ Blend + Treat + Error(Blend/Flask), RCBDPen.dat)
summary(RCBDPen.aov)

## two equivalent ways of extracting the residuals
res  <- residuals.aovlist(RCBDPen.aov)
res <- residuals(RCBDPen.aov, error.term = "Blend:Flask")
res <- resid.errors(RCBDPen.aov)
```

---

residuals.aovlist   *Extract the residuals from an aovlist object*

Description

Extracts the residuals from `error.term` or, if `error.term` is not specified, the last `error.term` in the analysis. It is a method for the generic function `residuals`.

Usage

```
## S3 method for class 'aovlist'
residuals(object, error.term=NULL, ...)
```

Arguments

`object`	An aovlist object created from a call to `aov`.
`error.term`	The term from the `Error` function for which the residuals are to be extracted. If `error.term` is NULL the residuals are extracted from the last `Error` term.
`...`	Further arguments passed to or from other methods.

Value

A numeric `vector` containing the residuals.

Author(s)

Chris Brien

See Also

`fitted.errors`, `resid.errors`, `tukey.1df` in package **dae**.

Examples

```
## set up data frame for randomized complete block design in Table 4.4 from
## Box, Hunter and Hunter (2005) Statistics for Experimenters. 2nd edn
## New York, Wiley.
RCBDPen.dat <- fac.gen(list(Blend=5, Flask=4))
RCBDPen.dat$Treat <- factor(rep(c("A","B","C","D"), times=5))
RCBDPen.dat$Yield <- c(89,88,97,94,84,77,92,79,81,87,87,
                       85,87,92,89,84,79,81,80,88)

## perform the analysis of variance
RCBDPen.aov <- aov(Yield ~ Blend + Treat + Error(Blend/Flask), RCBDPen.dat)
summary(RCBDPen.aov)
```

```
## two equivalent ways of extracting the residuals
res  <- residuals.aovlist(RCBDPen.aov)
res <- residuals(RCBDPen.aov, error.term = "Blend:Flask")
```

rmvnorm	*generates a vector of random values from a multivariate normal distribution*

Description

Generates a vector of random values from an n-dimensional multivariate normal distribution whose mean is given by the n-vector mean and variance by the n x n symmetric matrix V. It uses the method described by Ripley (1987, p.98)

Usage

```
rmvnorm(mean, V, method = 'eigenanalysis')
```

Arguments

mean	The mean vector of the multivariate normal distribution from which the random values are to be generated.
V	The variance matrix of the multivariate normal distribution from which the random values are to be generated.
method	The method used to decompose the variance matrix in producing a a matrix to transform the iid standard normal values. The two methods available are `'eigenanalysis'` and `'choleski'`, where only the first letter of each option is obligatory. The default method is eigenanalysis, which is slower but is likely to be more stable than Choleski decomposition.

Details

The method is: a) uses either the eigenvalue or Choleski decomposition of the variance matrix, V, to form the matrix that transforms an iid vector of values to a vector with variance V; b) generate a vector of length equal to mean of standard normal values; c) premultiply the vector of standard normal values by the transpose of the upper triangular factor and, to the result, add mean.

Value

A vector of length n, equal to the length of mean.

Author(s)

Chris Brien

References

Ripley, B. D. (1987) *Stochastic simulation*. Wiley, New York.

See Also

fac.ar1mat, fac.vcmat, in package **dae**, rnorm, and chol.

Examples

```
## set up a two-level factor and a three-level factor, both of length 12
A <- factor(rep(1:2, each=6))
B <- factor(rep(1:3, each=2, times=2))

## generate random values from a multivariate normal for which
#the mean is 20 for all variables and
#the variance matrix has random effects for factor A, ar1 pattern for B and
#residual random variation
mean <- rep(20, 12)
V <- fac.vcmat(A, 5) + fac.ar1mat(B, 0.6) + 2*mat.I(12)
y <- rmvnorm(mean, V)
```

Sensory3Phase.dat   *Data for the three-pahse sensory evaluation experiment in Brien, C.J.*
                           and Payne, R.W. (1999)

Description

The data is from an experiment involved two phases. In the field phase a viticultural experiment was conducted to investigate the differences between 4 types of trellising and 2 methods of pruning. The design was a split-plot design in which the trellis types were assigned to the main plots using two adjacent Youden squares of 3 rows and 4 columns. Each main plot was split into two subplots (or halfplots) and the methods of pruning assigned at random independently to the two halfplots in each main plot. The produce of each halfplot was made into a wine so that there were 24 wines altogether.

The second phase was an evaluation phase in which the produce from the halplots was evaluated by 6 judges all of whom took part in 24 sittings. In the first 12 sittings the judges evaluated the wines made from the halfplots of one square; the final 12 sittings were to evaluate the wines from the other square. At each sitting, each judge assessed two glasses of wine from each of the halplots of one of the main plots. The main plots allocated to the judges at each sitting were determined as follows. For the allocation of rows, each occasion was subdivided into 3 intervals of 4 consecutive sittings. During each interval, each judge examined plots from one particular row, these being determined using two 3x3 Latin squares for each occasion, one for judges 1-3 and the other for judges 4-6. At each sitting judges 1-3 examined wines from one particular column and judges 4-6 examined wines from another column. The columns were randomized to the 2 sets of judges x 3 intervals x 4 sittings using duplicates of a balanced incomplete block design for v=4 and k=2 that were latinized. This balanced incomplete block design consists of three sets of 2 blocks, each set containing the 4 "treatments". For each interval, a different set of 2 blocks was taken and each block assigned to two sittings, but with the columns within the block placed in reverse order in one sitting compared to the other sitting. Thus, in each interval, a judge would evaluate a wine from each of the 4 columns.

The `data.frame` contains the following factors, in the order give: Occasion, Judges, Interval, Sittings, Position, Squares, Rows, Columns, Halfplot, Trellis, Method. They are followed by the simulated response variable Score.

The scores are ordered so that the factors Occasion, Judges, Interval, Sittings and Position are in standard order; the remaining factors are in randomized order.

Usage

```
data(Sensory3Phase.dat)
```

Format

A data.frame containing 576 observations of 12 variables.

References

Brien, C.J. and Payne, R.W. (1999) Tiers, structure formulae and the analysis of complicated experiments. The Statistician, 48, 41-52.

---

set.daeTolerance   *Sets the values of daeTolerance for the package dae*

Description

A function that sets the values such that, in **dae** functions, values less than it are considered to be zero. The values are stored in a `vector` named daeTolerance in the daeEnv environment. The `vector` is of length two and, initially, both values are set to `.Machine$double.eps ^ 0.5` (about 1.5E-08). One value is named `element.tol` and is used for elements of matrices; the second is named `element.eigen` and is used for eigenvalues and quantities based on them, such as efficiency factors.

Usage

```
set.daeTolerance(element.tol=NULL, eigen.tol=NULL)
```

Arguments

element.tol	The value to to which the first element of the `daeTolerance` `vector` is to be set. If more than one value is supplied, only the first value is used.
eigen.tol	The value to to which the second element of the `daeTolerance` `vector` is to be set. If more than one value is supplied, only the first value is used.

Value

The `vector` daeTolerance is returned invisibly.

Author(s)

Chris Brien

See Also

get.daeTolerance.

Examples

```
## set daeTolerance.
set.daeTolerance(1E-04, 1E-08)
```

---

show-methods   *Methods for Function show in Package dae*

Description

Methods for function show in Package **dae**

Methods

signature(object = "projector") Prints the matrix and its degrees of freedom.

See Also

projector for further information about this class.

SPLGrass.dat   *Data for an experiment to investigate the effects of grazing patterns on pasture composition*

Description

The response variable is the percentage area covered by the principal grass (Main.Grass). The design for the experiment is a split-unit design. The main units are arranged in 3 Rows x 3 Columns. Each main unit is split into 2 SubRows x 2 SubColumns.

The factor Period, with levels 3, 9 and 18 days, is assigned to the main units using a 3 x 3 Latin square. The two-level factors Spring and Summer are assigned to split-units using a criss-cross design within each main unit. The levels of each of Spring and Summer are two different grazing patterns in its season.

Usage

data(SPLGrass.dat)

Format

A data.frame containing 36 observations of 8 variables.

Source

Example 14.1 from Mead, R. (1990). *The Design of Experiments: Statistical Principles for Practical Application.* Cambridge, Cambridge University Press.

---

strength *Generate paper strength values*

Description

Generates paper strength values for an experiment with different temperatures.

Usage

```
strength(nodays, noruns, temperature, ident)
```

Arguments

nodays	The number of days over which the experiment is to be run.
noruns	The number of runs to be performed on each day of the experiment.
temperature	A [factor](#) that encapsulates the layout by giving the temperature to be investigated for each run on each day. These must be ordered so that the temperatures for the first day are given in the order in which they are to be investigated on that day. These must be followed by the noruns temperatures for the second day and so on. Consequently, the factor temperature will have nodays*noruns values.
ident	The digits of your student identity number. That is, leave out any letters.

Value

A data.frame object containing the factors day, run and temperature and a vector of the generated strengths.

Author(s)

Chris Brien

Examples

```
## Here temperature is a factor with 4*3 = 12 values whose
## first 3 values specify the temperatures to be applied in
## the 3 runs on the first day, values 4 to 6 specify the
## temperatures for the 3 runs on day 2, and so on.
temperature <- factor(rep(c(80,85,90), 4))
exp.strength <- strength(nodays = 4, noruns = 3,
                         temperature = temperature, ident = 0123456)

## In this second example, a completely randomized design is generated
## for the same 3 temperatures replicated 4 times. The layout is stored
## in the data.frame called Design.
Design <- fac.layout(unrandomized=list(runs = 12),
                     randomized = temperature,
                     seed = 5847123)
## eradicate the unrandomized version of temperature
remove("temperature")

## The 12 temperatures in Design are to be regarded as being assigned to
## days and runs in the same manner as for the first example.
```

```
exp.strength <- strength(nodays = 4, noruns = 3,
                         temperature = Design$temperature, ident = 0123456)
```

---

summary.p2canon   *Summarize a canonical analysis of the relationships between two sets of projectors*

Description

Produces a summary of the efficiency criteria computed from the canonical efficiency factors for the joint decomposition of two sets of projectors (Brien and Bailey, 2009) obtained using `projs.2canon`. It takes the form of a decomposition or skeleton ANOVA table.

Usage

```
## S3 method for class 'p2canon'
summary(object, which.criteria = c("aefficiency", "eefficiency", "order"), ...)
```

Arguments

object   A `list` of class p2canon produced by `projs.2canon`.

which.criteria A character vector nominating the efficiency criteria to be included in the summary. It can be none, all or some combination of `aefficiency`, `mefficiency`, `sefficiency`, `eefficiency`, `xefficiency`, `order` and `dforthog` – for details see `efficiency.criteria`.

...   further arguments affecting the summary produced.

Value

An object of classes summary.p2canon and data.frame, whose rows correspond to the pairs of projectors, one from the Q1 argument and the other from the Q2 argument from `projs.2canon`; only pairs with non-zero efficiency factors are included. In addition, a line is included for each nonzero Residual Q1 projector.

Author(s)

Chris Brien

References

Brien, C. J. and R. A. Bailey (2009). Decomposition tables for multitiered experiments. I. A chain of randomizations. *The Annals of Statistics*, **36**, 4184 - 4213.

See Also

`projs.2canon`, `proj2.efficiency`, `efficiency.criteria`, `proj2.combine`, `proj2.eigen`, `projs.structure`, `print.summary.p2canon` in package **dae**, `eigen`.

`projector` for further information about this class.

Examples

```
## PBIBD(2) from p. 379 of Cochran and Cox (1957) Experimental Designs.
## 2nd edn Wiley, New York
PBIBD2.unit <- list(Block = 6, Unit = 4)
PBIBD2.nest <- list(Unit = "Block")
trt <- factor(c(1,4,2,5, 2,5,3,6, 3,6,1,4, 4,1,5,2, 5,2,6,3, 6,3,4,1))
PBIBD2.lay <- fac.layout(unrandomized = PBIBD2.unit,
                         nested.factors=PBIBD2.nest,
                         randomized = trt)

##obtain projectors using projs.structure
Q.unit <- projs.structure(~ Block/Unit, data = PBIBD2.lay)
Q.trt <- projs.structure(~ trt, data = PBIBD2.lay)

##obtain combined decomposition and summarize
unit.trt.p2canon <- projs.2canon(Q.unit, Q.trt)
summary(unit.trt.p2canon)
```

---

summary.pcanon	*Summarize a canonical analysis of the relationships between sets of projectors*

Description

Produces a summary of the efficiency criteria computed from the canonical efficiency factors for the joint decomposition of two or more sets of projectors (Brien and Bailey, 2009) obtained using `projs.canon`. It takes the form of a decomposition or skeleton ANOVA table.

Usage

```
## S3 method for class 'pcanon'
summary(object, which.criteria = c("aefficiency", "eefficiency", "order"), ...)
```

Arguments

`object`	A `list` of class pcanon produced by `projs.canon`.
`which.criteria`	A character `vector` nominating the efficiency criteria to be included in the summary. It can be none, all or some combination of aefficiency, mefficiency, sefficiency, eefficiency, xefficiency, order and dforthog – for details see `efficiency.criteria`. If there is only one formula, this argument is ignored.
`...`	further arguments affecting the summary produced.

Value

An object of classes `summary.pcanon` and `data.frame`, whose rows correspond to subspaces in the decomposition; it is of the nature of a skeleton analysis of variance table. It has an `attribute` named `ntiers` that is equal to the number of tiers.

Author(s)

Chris Brien

References

Brien, C. J. and R. A. Bailey (2009). Decomposition tables for multitiered experiments. I. A chain of randomizations. *The Annals of Statistics*, **36**, 4184 - 4213.

See Also

[projs.canon](), [proj2.efficiency](), [efficiency.criteria](), [proj2.combine](), [proj2.eigen](), [projs.structure](), [print.summary.pcanon]()in package **dae**, [eigen]().

[projector]() for further information about this class.

Examples

```
## PBIBD(2) from p. 379 of Cochran and Cox (1957) Experimental Designs.
## 2nd edn Wiley, New York
PBIBD2.unit <- list(Block = 6, Unit = 4)
PBIBD2.nest <- list(Unit = "Block")
trt <- factor(c(1,4,2,5, 2,5,3,6, 3,6,1,4, 4,1,5,2, 5,2,6,3, 6,3,4,1))
PBIBD2.lay <- fac.layout(unrandomized = PBIBD2.unit,
                         nested.factors=PBIBD2.nest,
                         randomized = trt)

##obtain combined decomposition and summarize
unit.trt.canon <- projs.canon(list(unit=~ Block/Unit, trt=~ trt), data = PBIBD2.lay)
summary(unit.trt.canon, which = c("aeff","eeff","order"))
```

---

tukey.1df	*Performs Tukey's one-degree-of-freedom-test-for-nonadditivity*

Description

Performs Tukey's one-degree-of-freedom-test-for-nonadditivity on a set of residuals from an analysis of variance.

Usage

```
tukey.1df(aov.obj, data, error.term="Within")
```

Arguments

aov.obj	An aov object or aovlist object created from a call to [aov]().
error.term	The term from the Error function whose residuals are to be tested for nonadditivity. Only required when the Error function used in call to aov, so that an aovlist object is created.
data	A data.frame containing the original response variable and factors used in the call to [aov]().

Value

A list containing Tukey.SS, Tukey.F, Tukey.p, Devn.SSq being the SSq for the 1df test, F value for test and the p-value for the test.

Note

In computing the test quantities fitted values must be obtained. If error.term is specified, fitted values will be the sum of effects extracted from terms from the Error function, but only down to that specified by error.term.The order of terms is as given in the ANOVA table. If error.term is unspecified, all effects for terms external to any Error terms are extracted and summed.

Extracted effects will only be for terms external to any Error function. If you want effects for terms in the Error function to be included, put them both inside and outside the Error function so they are occur twice.

Author(s)

Chris Brien

See Also

[fitted.errors](), [resid.errors]() in package **dae**.

Examples

```
## set up data frame for randomized complete block design in Table 4.4 from
## Box, Hunter and Hunter (2005) Statistics for Experimenters. 2nd edn
## New York, Wiley.
RCBDPen.dat <- fac.gen(list(Blend=5, Flask=4))
RCBDPen.dat$Treat <- factor(rep(c("A","B","C","D"), times=5))
RCBDPen.dat$Yield <- c(89,88,97,94,84,77,92,79,81,87,87,
                       85,87,92,89,84,79,81,80,88)

## perform the analysis of variance
RCBDPen.aov <- aov(Yield ~ Blend + Treat + Error(Blend/Flask), RCBDPen.dat)
summary(RCBDPen.aov)

## Obtain the quantities for Tukey's test
tukey.1df(RCBDPen.aov, RCBDPen.dat, error.term = "Blend:Flask")
```

---

yates.effects	*Extract Yates effects*

Description

Extracts Yates effects from an aov object or aovlist object.

Usage

```
yates.effects(aov.obj, error.term="Within", data=NULL)
```

Arguments

aov.obj	An aov object or aovlist object created from a call to [aov]().
error.term	The term from the Error function from which the Yates effects are estimated. Only required when Error used in call to aov.
data	A data.frame in which the variables specified in the aov.obj will be found. If missing, the variables are searched for in the standard way.

Details

Yates effects are specific to 2^k experiments, where Yates effects are conventionally defined as the difference between the upper and lower levels of a factor. We follow the convention used in Box, Hunter and Hunter (1978) for scaling of higher order interactions: all the Yates effects are on the same scale, and represent the average difference due to the interaction between two different levels. Effects are estimated only from the error term supplied to the error.term argument.

Value

A vector of the Yates effects.

Author(s)

Chris Brien

See Also

qqyeffects in package **dae**, aov.

Examples

```
## analysis of 2^4 factorial experiment from Table 10.6 of Box, Hunter and
## Hunter (1978) Statistics for Experimenters. New York, Wiley.
## use ?Fac4Proc.dat for data set details
data(Fac4Proc.dat)
Fac4Proc.aov <- aov(Conv ~ Catal * Temp * Press * Conc + Error(Runs),
                                                   Fac4Proc.dat)
round(yates.effects(Fac4Proc.aov, error.term="Runs", data=Fac4Proc.dat), 2)
```

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