# Package 'ffmanova' 

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## Type Package

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Author Øyvind Langsrud [aut, cre],
Bjørn-Helge Mevik [aut]
Maintainer Øyvind Langsrud [oyl@ssb.no](mailto:oyl@ssb.no)

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Description General linear modeling with multiple responses (MANCOVA). An overall pvalue for each model term is calculated by the 50-
50 MANOVA method by Langsrud (2002) [doi:10.1111/1467-9884.00320](doi:10.1111/1467-9884.00320), which handles collinear responses. Rotation testing, described by Langsrud (2005) [doi:10.1007/s11222-005-4789-5](doi:10.1007/s11222-005-4789-5), is used to compute adjusted single response p-values according to familywise error rates and false discovery rates (FDR). The approach to FDR is described in the appendix of Moen et al. (2005) [doi:10.1128/AEM.71.4.2086-2094.2005](doi:10.1128/AEM.71.4.2086-2094.2005). Unbalanced designs are handled by Type II sums of squares as argued in Langsrud (2003) [doi:10.1023/A:1023260610025](doi:10.1023/A:1023260610025). Furthermore, the Type II philosophy is extended to continuous design variables as described in Langsrud et al. (2007) [doi:10.1080/02664760701594246](doi:10.1080/02664760701594246). This means that the method is invariant to scale changes and that common pitfalls are avoided.

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dressing Dressing data

## Description

A dataset from an experiment studying structural and rheological properties of a full fat dressing.

## Usage

data(dressing)

## Format

A data frame with 29 observations on the following 7 variables.
press a numeric vector with values 75,125 and 225 . The homogenisation pressure.
stab a numeric vector with values $0.3,0.4$ and 0.5 . Amount of stabiliser.
emul a numeric vector with values $0.1,0.2$ and 0.35 . Amount of emulsifier.
day a factor with levels $1, \ldots, 5$. The day the experimental run was performed on.
visc a numeric vector. The measured viscosity of the dressing.
rheo a matrix with 9 columns. Nine different response-parameters derived from rheological measuring. These parameters contain information about the physics of the dression (more general that viscosity).
pvol a matrix with 241 columns. Particle-volume curves. Using a coulter-counter instrument fat particles were counted and their volumes were registered. These data are presented as smoothed histograms (equally spaced bins on log-scale). The total area under the curve represents the total volume of the counted fat particles. The shape of the curve reflects how the total fat volume is distributed among the different particle sizes.

## Details

The data comes from an experiment in which full fat dressings were produced with different amount of stabiliser and emulsifier, and with different homogenisation pressure (se above).
A full factorial $3^{3}$ design with two additional center points was used. The experiment was run over five days. It was unknown up front how many experimental runs could be performed each day, so the order of the runs was randomised.

For each dressing, viscosity, rheology and particle volume measurements were taken (se above).
The day is stored as a factor. The other design variables are stored as numerical variables. If one wants to use them as factors, one can use e.g. factor (press) in the model formula, or dressing\$press <- factor(dressing\$press) prior to calling the modelling function.

## Source

The data is taken from a research project financed by a grant (131472/112) from the Norwegian Research Council. The project was managed by Stabburet, which is a major manufacturer of dressing in Norway.
ffAnova Type II* Anova

## Description

Analysis of variance table for a linear model using type II* sums of squares, which are described in Langsrud et al. (2007). Type II* extends the type II philosophy to continuous variables. The results are invariant to scale changes and pitfalls are avoided.

## Usage

ffAnova(formula, data $=$ NULL)

## Arguments

formula A model formula or an R object (preferably an lm object).
data An optional data frame, list or environment.

## Details

This function is a variant of ffmanova for the univariate special case. The two input parameters will be interpreted by model. frame.

## Value

An object of class "anova" (see anova).

## Author(s)

Øyvind Langsrud and Bjørn-Helge Mevik

## References

Langsrud, Ø., Jørgensen, K., Ofstad, R. and Næs, T. (2007): "Analyzing Designed Experiments with Multiple Responses", Journal of Applied Statistics, 34, 1275-1296.

## Examples

```
# Generate example data
set.seed(123)
a <- c(0, 0, 0, 10, 10, 10, 1, 1, 1)
A <- as.character(a) # A is categorical
b <- 1:9
y <- rnorm(9)/10 + a # y depends strongly on a (and A)
a100 <- a + 100 # change of scale (origin)
b100 <- b + 100 # change of scale (origin)
# Four ways of obtaining the same results
ffAnova(y ~ A * b)
ffAnova(y ~ A * b100)
ffAnova(lm(y ~ A * b))
ffAnova(y ~ A * b, data.frame(A = A, y = y, b = 1:9))
# Second order continuous variable
ffAnova(y ~ a + I(a^2))
# Model equivalent to 'y ~ A * b'
ffAnova(y ~ (a + I(a^2)) * b)
# Demonstrating similarities and differences using package car
if (!require(car)) # Package car is loaded if available
    Anova <- function(x) { # Replacement function if car not available
        warning("No results since package car is not available")}
lm_Ab <- lm(y ~ A * b)
lm_Ab100 <- lm(y ~ A * b100)
# Type II same as type II* in this case
Anova(lm_Ab) # Type II
Anova(lm_Ab100) # Type II
ffAnova(lm_Ab) # Type II*
ffAnova(lm_Ab100) # Type II*
# Type III depends on scale
Anova(lm_Ab, type = 3)
Anova(lm_Ab100, type = 3)
lm_a <- lm(y ~ a + I(a^2))
lm_a100 <- lm(y ~ a100 + I(a100^2))
# Now Type II depends on scale
Anova(lm_a) # Type II
Anova(lm_a100) # Type II
ffAnova(lm_a) # Type II*
```


## Description

General linear modeling of fixed-effects models with multiple responses is performed. The function calculates 50-50 MANOVA $p$-values, ordinary univariate $p$-values and adjusted $p$-values using rotation testing.

## Usage

ffmanova(formula, data $=$ NULL, stand $=$ TRUE, nSim $=0$, verbose $=$ TRUE, returnModel $=$ TRUE, returnY $=$ FALSE, returnYhat $=$ FALSE, returnYhatStd $=$ FALSE, newdata $=$ NULL, linComb = NULL, nonEstimableAsNA = TRUE, outputClass = "ffmanova")

## Arguments

\(\left.$$
\begin{array}{ll}\text { formula } & \text { Model formula. See "Note" below. } \\
\text { data } & \text { An optional data frame or list. } \\
\text { stand } & \begin{array}{l}\text { Logical. Standardization of responses. This option has effect on the } 50-50 \\
\text { MANOVA testing and the calculation of exVarSS. }\end{array} \\
\text { nSim } & \begin{array}{l}\text { nonnegative integer. The number of simulations to use in the rotation tests. Can } \\
\text { be a single nonnegative integer or a list of values for each term. }\end{array} \\
\text { verbose } & \begin{array}{l}\text { Logical. If TRUE, the rotation tests print trace information. }\end{array} \\
\text { returnModel } & \begin{array}{l}\text { When TRUE, and object, ffModel, with output from ffModelObj is included in } \\
\text { output. Must be TRUE to enable predictions by predict.ffmanova. }\end{array} \\
\text { returnY } & \begin{array}{l}\text { Response matrix, Y, in output when TRUE. }\end{array} \\
\text { returnYhat } & \begin{array}{l}\text { Matrix Yhat of fitted values corresponding to Y in output when TRUE. }\end{array} \\
\text { returnYhatStd } & \begin{array}{l}\text { Standard errors, YhatStd, in output when TRUE. } \\
\text { newdata }\end{array}
$$ <br>
Possible input to predict.ffmanova. When non-NULL, prediction results will <br>

be included output.\end{array}\right]\)| Possible input to predict.ffmanova in addition to newdata. |
| :--- | :--- |

## Details

An overall $p$-value for all responses is calculated for each model term. This is done using the $50-$ 50 MANOVA method, which is a modified variant of classical MANOVA made to handle several highly correlated responses.
Ordinary single response $p$-values are produced. By using rotation testing these can be adjusted for multiplicity according to familywise error rates or false discovery rates. Rotation testing is a Monte Carlo simulation framework for doing exact significance testing under multivariate normality. The number of simulation repetitions (nSim) must be chosen.
Unbalance is handled by a variant of Type II sums of squares, which has several nice properties:

1. Invariant to ordering of the model terms.
2. Invariant to scale changes.
3. Invariant to how the overparameterization problem of categorical variable models is solved (how constraints are defined).
4. Whether two-level factors are defined to be continuos or categorical does not influence the results.
5. Analysis of a polynomial model with a single experimental variable produce results equivalent to the results using an orthogonal polynomial.

In addition to significance testing an explained variance measure, which is based on sums of sums of squares, is computed for each model term.

## Value

An object of class "ffmanova", which consists of the concatenated results from the underlying functions manova5050, rotationtests and unitests:

| termNames | model term names |
| :--- | :--- |
| exVarSS | explained variances calculated from sums of squares summed over all responses |
| df | degrees of freedom - adjusted for other terms in model |
| df_om | degrees of freedom - adjusted for terms contained in actual term <br> nPC |
| nBU | number of principal components used for testing |
| exVarPC | variance explained by nPC components |
| exVarBU | variance explained by (nPC+nBU) components |
| pValues | $50-50$ MANOVA $p$-values |
| stand | logical. Whether the responses are standardised. |
| stat | The test statistics as $t$-statistics (when single degree of freedom) or $F$-statistics |
| pRaw | matrix of ordinary $p$-values from F- or t-testing |
| pAdjusted | matrix of adjusted $p$-values according to familywise error rates |
| pAdjFDR | matrix of adjusted $p$-values according to false discovery rates |
| simN | number of simulations performed for each term (same as input) |

The matrices stat, pRaw, pAdjusted and pAdjFDR have one row for each model term and one column for each response.
According to the input parameters, additional elements can be included in output.

## Note

The model is specified with formula, in the same way as in lm (except that offsets are not supported). See lm for details. Input parameters formula and data will be interpreted by model. frame.

## Author(s)

Øyvind Langsrud and Bjørn-Helge Mevik

## References

Langsrud, Ø. (2002) 50-50 Multivariate Analysis of Variance for Collinear Responses. The Statistician, 51, 305-317.
Langsrud, Ø. (2003) ANOVA for Unbalanced Data: Use Type II Instead of Type III Sums of Squares. Statistics and Computing, 13, 163-167.

Langsrud, Ø. (2005) Rotation Tests. Statistics and Computing, 15, 53-60.
Moen, B., Oust, A., Langsrud, Ø., Dorrell, N., Gemma, L., Marsden, G.L., Hinds, J., Kohler, A., Wren, B.W. and Rudi, K. (2005) An explorative multifactor approach for investigating global survival mechanisms of Campylobacter jejuni under environmental conditions. Applied and Environmental Microbiology, 71, 2086-2094.

See also https://www.langsrud.com/stat/program.htm.

## See Also

ffAnova and predict.ffmanova.

## Examples

```
data(dressing)
# An ANOVA model with all design variables as factors
# and with visc as the only response variable.
# Classical univariate Type II test results are produced.
ffmanova(visc ~ (factor(press) + factor(stab) + factor(emul))^2 + day,
    data = dressing)
# A second order response surface model with day as a block factor.
# The properties of the extended Type II approach is utilized.
ffmanova(visc ~ (press + stab + emul)^2 + I(press^2)+ I(stab^2)+ I(emul^2)+ day,
    data = dressing)
# 50-50 MANOVA results with the particle-volume curves as
# multivariate responses. The responses are not standardized.
ffmanova(pvol ~ (press + stab + emul)^2 + I(press^2)+ I(stab^2)+ I(emul^2)+ day,
    stand = FALSE, data = dressing)
# 50-50 MANOVA results with 9 rheological responses (standardized).
# 99 rotation simulation repetitions are performed.
res <- ffmanova(rheo ~ (press + stab + emul)^2 + I(press^2)+ I(stab^2)+ I(emul^2)+ day,
```

```
            nSim = 99, data = dressing)
res$pRaw # Unadjusted single responses p-values
res$pAdjusted # Familywise error rate adjusted p-values
res$pAdjFDR # False discovery rate adjusted p-values
# As above, but this time 9999 rotation simulation repetitions
# are performed, but only for the model term stab^2.
res <- ffmanova(rheo ~ (press + stab + emul)^2 + I(press^2)+I(stab^2)+I(emul^2)+ day,
    nSim = c(0,0,0,0,0,9999,0,0,0,0,0), data = dressing)
res$pAdjusted[6,] # Familywise error rate adjusted p-values for stab^2
res$pAdjFDR[6,] # False discovery rate adjusted p-values for stab^2
# Note that the results of the first example above can also be
# obtained by using the car package.
## Not run:
        require(car)
    Anova(lm(visc ~ (factor(press) + factor(stab) + factor(emul))^2 + day,
                data = dressing), type = "II")
## End(Not run)
# The results of the second example differ because Anova does not recognise
# linear terms (emul) as being contained in quadratic terms (I(emul^2)).
# A consequence here is that the clear significance of emul disappears.
## Not run:
    require(car)
    Anova(lm(visc ~ (press + stab + emul)^2 + I(press^2)+ I(stab^2)+ I(emul^2)+ day,
                data = dressing), type="II")
## End(Not run)
```


## Description

The function takes the factor matrix of the terms object corresponding to a model formula and changes it so that model hierarchy is preserved also for powers of terms (e.g., I (a^2)).

## Usage

fixModelMatrix(mOld)

## Arguments

mOld
The factor matrix (i.e. the "factor" attribute) of a terms object.

## Details

The ordinary model handling functions in do not treat powers of terms $\left(a^{n}\right)$ as being higher order terms (like interaction terms). fixModelMatrix takes the "factor" attribute of a terms object (usually created from a model formula) and changes it such that power terms can be treated hierarchically just like interaction terms.

The factor matrix has one row for each variable and one coloumn for each term. Originally, an entry is 0 if the term does not contain the variable. If it contains the variable, the entry is 1 if the variable should be coded with contrasts, and 2 if it should be coded with dummy variables. See terms.object for details.

The changes performed by fixModelMatrix are:

- Any 2 's are changed to 1 .
- In any coloumn corresponding to a term that contains $I\left(a^{\wedge} n\right)$, where $a$ is the name of a variable and $n$ is a positive integer, the element in the row corresponding to $a$ is set to $n$. For instance, the entry of row $D$ and coloumn $C: I\left(D^{\wedge} 2\right)$ is set to 2 .
- Rows corresponding to $I\left(a^{\wedge} n\right)$ are deleted.

Note that this changes the semantics of the factor matrix: 2 no longer means 'code via dummy variables'.

## Value

A factor matrix.

## Author(s)

Øyvind Langsrud and Bjørn-Helge Mevik

## See Also

terms, terms.object

## Examples

```
mt <- terms(y ~ a + b + a:b + a:c + I (a^2) + I (a^3) + I (a^2):b)
print(mOld <- attr(mt, "factor"))
fixModelMatrix(mOld)
```


## Description

The function takes a design-with-responses object created by xy_Obj and produces 50-50 MANOVA output. Results are produced for each term in the model.

## Usage

manova5050(xy0bj, stand)

## Arguments

xyObj design-with-responses object
stand standardisation of responses (0 or 1 )

## Details

Classical multivariate ANOVA (MANOVA) are useless in many practical cases. The tests perform poorly in cases with several highly correlated responses and the method collapses when the number of responses exceeds the number of observations. 50-50 MANOVA is made to handle this problem. Principal component analysis (PCA) is an important part of this methodology. Each test is based on a separate PCA.

## Value

A list with components

| termNames | model term names |
| :--- | :--- |
| exVarSS | explained variances calculated from sums of squares summed over all responses |
| df | degrees of freedom - adjusted for other terms in model |
| df_om | degrees of freedom - adjusted for terms contained in actual term |
| nPC | number of principal components used for testing |
| nBU | number of principal components used as buffer components |
| exVarPC | variance explained by nPC components |
| exVarBU | variance explained by (nPC+nBU) components |
| pValues | $50-50$ MANOVA p-values |
| stand | logical. Whether the responses are standardised. |

## Note

The 50-50 MANOVA $p$-values are based on the Hotelling-Lawley Trace Statistic. The number of components for testing and the number of buffer components are chosen according to default rules.

## Author(s)

Øyvind Langsrud and Bjørn-Helge Mevik

## References

Langsrud, Ø. (2002) Rotation Tests. The Statistician, 51, 305-317.

## See Also

ffmanova

```
matlabColon Simulate Matlab's ':'
```


## Description

A function to simulate Matlab's ' $:$ ' operator.

## Usage

matlabColon(from, to)

## Arguments

| from | numeric. The start value. |
| :--- | :--- |
| to | numeric. The end value. |

## Details

matlabCode $(a, b)$ returns $a: b$ ('s version) unless $a>b$, in which case it returns integer (0).

## Value

A numeric vector, possibly empty.

## Author(s)

Bjørn-Helge Mevik

## See Also

seq

## Examples

```
identical(3:5, matlabColon(3, 5)) ## => TRUE
3:1 ## => 3 2 1
matlabColon(3, 1) ## => integer(0)
```

```
    multiPvalues p-values from MANOVA test statistics
```


## Description

$p$-values from the four MANOVA test statistics are calculated according to the traditional F-distribution approximations (exact in some cases).

## Usage

multiPvalues(D, E, A, M, dim, dimX, dimY)

## Arguments

D Wilks' Lambda
E Roy's Largest Root
A Hotelling-Lawley Trace Statistic
M Pillay-Bartlett Trace Statistic
dim Number of observations
$\operatorname{dimX} \quad$ Number of x -variables
$\operatorname{dim} Y \quad$ Number of $y$-variables

## Details

The parameters dim, dimX and dimY corresponds to a situation where the test statistics are calculated from two data matrices with zero mean (test of independence).

## Value

pD $\quad p$-value: Wilks' Lambda
pE $\quad p$-value: LOWER BOUND for Roy's Largest Root
pA $\quad p$-value: Hotelling-Lawley Trace Statistic
pM $\quad p$-value: Pillay-Bartlett Trace Statistic

## Author(s)

Øyvind Langsrud and Bjørn-Helge Mevik

## See Also

ffmanova

```
multiStatistics MANOVA test statistics
```


## Description

The four classical MANOVA test statistics are calculated from a set of eigenvalues.

## Usage

multiStatistics(ss)

## Arguments

ss
A list of eigenvalues

## Details

These eigenvalues are also known as the squared canonical correlation coefficients.

## Value

A list with elements
D Wilks’ Lambda
E Roy's Largest Root
A Hotelling-Lawley Trace Statistic
M Pillay-Bartlett Trace Statistic

## Author(s)

Øyvind Langsrud and Bjørn-Helge Mevik
predict.ffmanova Predictions, mean predictions, adjusted means and linear combinations

## Description

The same predictions as $1 m$ can be obtained. With some variables missing in input, adjusted means or mean predictions are computed (Langsrud et al., 2007). Linear combinations of such predictions, with standard errors, can also be obtained.

## Usage

\#\# S3 method for class 'ffmanova'
predict(object, newdata $=$ NULL, linComb $=$ NULL, nonEstimableAsNA $=$ TRUE, ...)

## Arguments

object Output from ffmanova.
newdata Data frame or list. Missing values and missing variables are possible.
linComb A matrix defining linear combinations.
nonEstimableAsNA
When TRUE missing values are retuned when predictions cannot be made. When FALSE predictions are made anyway, but the logical vector, estimable, is added to output in cases of non-estimable results.
... further arguments (not used)

## Value

A list of two matrices:

| YnewPred | Predictions, mean predictions, adjusted means or linear combinations of such <br> predictions. |
| :--- | :--- |
| YnewStd | Corresponding standard errors. |

## References

Langsrud, Ø., Jørgensen, K., Ofstad, R. and Næs, T. (2007): "Analyzing Designed Experiments with Multiple Responses", Journal of Applied Statistics, 34, 1275-1296.

## Examples

```
# Generate data
x1 <- 1:6
x2 <- rep(c(100, 200), each = 3)
y1 <- x1 + rnorm(6)/10
y2 <- y1 + x2 + rnorm(6)/10
# Create ffmanova object
ff <- ffmanova(cbind(y1, y2) ~ x1 + x2)
# Predictions from the input data
predict(ff)
# Rows 1 and 5 from above predictions
predict(ff, data.frame(x1 = c(1, 5), x2 = c(100, 200)))
# Rows 1 as above and row 2 different
predict(ff, data.frame(x1 = c(1, 5), x2 = 100))
# Three ways of making the same mean predictions
predict(ff, data.frame(x1 = c(1, 5), x2 = 150))
predict(ff, data.frame(x1 = c(1, 5), x2 = NA))
predict(ff, data.frame(x1 = c(1, 5)))
# Using linComb input specified to produce regression coefficients
# with std. As produced by summary(lm(cbind(y1, y2) ~ x1 + x2))
```

```
predict(ff, data.frame(x1 = c(1, 2)), matrix(c(-1, 1), 1, 2))
predict(ff, data.frame(x2 = c(101, 102)), matrix(c(-1, 1), 1, 2))
# Above results by a 2*4 linComb matrix and with rownames
lC <- t(matrix(c(-1, 1, 0, 0, 0, 0, -1, 1), 4, 2))
rownames(lC) <- c("x1", "x2")
predict(ff, data.frame(x1 = c(1, 2, 1, 1), x2 = c(100, 100, 101, 102)), lC)
```

```
rotationtests Rotation testing
```


## Description

The functions perform rotation testing based on a matrix of hypothesis observations and a matrix of error observations. Adjusted $p$-values according to familywise error rates and false discovery rates are calculated.

## Usage

rotationtests(xyObj, nSim, verbose = TRUE)
rotationtest(modelData, errorData, $\operatorname{simN}=999, \mathrm{dfE}=-1$, dispsim = TRUE)

## Arguments

xyObj a design-with-responses object created by xy_Obj
nSim vector of nonnegative integers. The number of simulations to use for each term.
verbose logical. Whether rotationtests (and rotationtest) should be verbose.
modelData matrix of hypothesis observations
errorData matrix of error observations
$\operatorname{simN} \quad$ Number of simulations for each test. Can be a single value or a list of values for each term.
dfE Degrees of freedom for error needs to be specified if errorData is incomplete
dispsim When TRUE, dots are displayed to illustrate simulation progress.

## Details

modelData and errorObs correspond to hypObs and errorObs calculated by xy_Obj. These matrices are efficient representations of sums of squares and cross-products (see xy_0bj for details). This means that rotationtest can be viewed as a generalised $F$-test function.
rotationtests is a wrapper function that calls rotationtest for each term in the xyObj and collects the results.

## Value

Both functions return a list with components

| pAdjusted | adjusted $p$-values according to familywise error rates |
| :--- | :--- |
| pAdjFDR | adjusted $p$-values according to false discovery rates |
| simN | number of simulations performed for each term |

## Author(s)

Øyvind Langsrud and Bjørn-Helge Mevik

## References

Langsrud, Ø. (2005) Rotation Tests. Statistics and Computing, 15, 53-60.
Moen, B., Oust, A., Langsrud, Ø., Dorrell, N., Gemma, L., Marsden, G.L., Hinds, J., Kohler, A., Wren, B.W. and Rudi, K. (2005) An explorative multifactor approach for investigating global survival mechanisms of Campylobacter jejuni under environmental conditions. Applied and Environmental Microbiology, 71, 2086-2094.

See Also
unitest, unitests
stdize Centering and scaling of matrices

## Description

Function to center and/or scale the coloumns of a matrix in various ways. The coloumns can be centered with their means or with supplied values, and they can be scaled with their standard deviations or with supplied values.

## Usage

stdize(x, center = TRUE, scale = TRUE, avoid.zero.divisor = FALSE)
stdize3(x, center = TRUE, scale = TRUE, avoid.zero.divisor = FALSE)

## Arguments

| $x$ | A matrix. |
| :--- | :--- |
| center | A logical, or a numeric vector. The values to subtract from each column. If <br> center is TRUE, the mean values are used. |
| scale | A lgical, or a numeric vector. The values to divide each column with. If scale <br> is TRUE, the standard deviations are used. |
| avoid.zero.divisor |  |

A logical. If TRUE, each occurence of 0 in scale is replaced with a 1.

## Details

stdize standardizes the coloumns of a matrix by subtracting their means (or the supplied values) and dividing by their standard deviations (or the supplied values).
If avoid.zero.divisor is TRUE, division-by-zero is guarded against by substituting any 0 in center (either calculated or supplied) with 1 prior to division.

The main difference between stdize and scale is that stdize divides by the standard deviations even when center is not TRUE.

## Value

A matrix.

## Note

stdize3 is a variant with a three-element list as output ( x , center, scale) and where avoid. zero.divisor is also used to avoid centring (constant term in model matrix is unchanged).

## Author(s)

Bjørn-Helge Mevik and Øyvind Langsrud

## See Also

scale

## Examples

```
A <- matrix(rnorm(15, mean = 1), ncol = 3)
stopifnot(all.equal(stdize(A), scale(A), check.attributes = FALSE))
## These are different:
stdize(A, center = FALSE)
scale(A, center = FALSE)
```

unitests Univariate F or t testing

## Description

The functions perform $F$ or $t$ testing for several responses based on a matrix of hypothesis observations and a matrix of error observations.

## Usage

unitests(xyObj)
unitest(modelData, errorData, dfError $=\operatorname{dim}(e r r o r D a t a)[1])$

## Arguments

xyObj a design-with-responses object created by xy_Obj
modelData matrix of hypothesis observations
errorData matrix of error observations
dfError Degrees of freedom for error needs to be specified if errorData is incomplete

## Details

modelData and errorObs correspond to hypObs and errorObs calculated by xy_Obj. These matrices are efficient representations of sums of squares and cross-products (see xy_0bj for details). This means the univariate $F$-statistics can be calculated straightforwardly from these input matrices. Furthermore, in the single-degree-of-freedom case, $t$-statistics with correct sign can be obtained.
unitests is a wrapper function that calls unitest for each term in the xyObj (see xy_Obj for details) and collects the results.

## Value

unitest returns a list with components
pValues $\quad p$-values
stat $\quad$ The test statistics as $t$-statistics (when single degree of freedom) or $F$-statistics
unitests returns a list with components
pRaw $\quad$ Matrix of $p$-values from unitest, one row for each term.
stat Matrix of test statistics from unitest, one row for each term.

## Note

The function calculates the $p$-values by making a call to pf .

## Author(s)

Øyvind Langsrud and Bjørn-Helge Mevik

## See Also

rotationtest, rotationtests

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