Package ‘gmodels’

July 22, 2015

Version 2.16.2
Date 2015-07-21
Title Various R Programming Tools for Model Fitting
Author Gregory R. Warnes, Ben Bolker, Thomas Lumley, and Randall C
        Johnson. Contributions from Randall C. Johnson are Copyright
        (2005) SAIC-Frederick, Inc. Funded by the Intramural Research
        Program, of the NIH, National Cancer Institute, Center for
        Cancer Research under NCI Contract NO1-CO-12400.
Maintainer Gregory R. Warnes <greg@warnes.net>
Description Various R programming tools for model fitting.
Depends R (>= 1.9.0)
Suggests gplots, gtools, Matrix, nlme, lme4 (>= 0.999999-0)
Imports MASS, gdata
License GPL-2
URL http://www.sf.net/projects/r-gregmisc
NeedsCompilation no
Repository CRAN
Date/Publication 2015-07-22 06:11:15

R topics documented:

<table>
<thead>
<tr>
<th>Topic</th>
<th>Page</th>
</tr>
</thead>
<tbody>
<tr>
<td>ci</td>
<td>2</td>
</tr>
<tr>
<td>coefFrame</td>
<td>3</td>
</tr>
<tr>
<td>CrossTable</td>
<td>5</td>
</tr>
<tr>
<td>estimable</td>
<td>7</td>
</tr>
<tr>
<td>fast.prcomp</td>
<td>9</td>
</tr>
<tr>
<td>fit.contrast</td>
<td>11</td>
</tr>
<tr>
<td>glh.test</td>
<td>15</td>
</tr>
<tr>
<td>make.contrasts</td>
<td>17</td>
</tr>
</tbody>
</table>

Index 20
ci

Compute Confidence Intervals

Description

Compute and display confidence intervals for model estimates. Methods are provided for the mean
of a numeric vector ci.default, the probability of a binomial vector ci.binom, and for lm, and
lme
objects are provided.

Usage

```r
.ci(x, confidence=0.95, alpha=1-confidence, ...)  # S3 method for class 'numeric'
.ci(x, confidence=0.95, alpha=1-confidence, na.rm=FALSE, ...)  # S3 method for class 'binom'
.ci(x, confidence=0.95, alpha=1-confidence, ...)  # S3 method for class 'lm'
.ci(x, confidence=0.95, alpha=1-confidence, ...)  # S3 method for class 'lme'
.ci(x, confidence=0.95, alpha=1-confidence, ...)  # S3 method for class 'estimable'
```

Arguments

- `x` object from which to compute confidence intervals.
- `confidence` confidence level. Defaults to 0.95.
- `alpha` type one error rate. Defaults to 1.0-confidence
- `na.rm` boolean indicating whether missing values should be removed. Defaults to FALSE.
- `...` Arguments for methods

Details

ci.binom computes binomial confidence intervals using the Clopper-Pearson 'exact' method based
on the binomial quantile function. Due to the discrete nature of the binomial distribution, this
interval is conservative.

Value

vector or matrix with one row per model parameter and elements/columns Estimate, CI lower,
CI upper, Std. Error, DF (for lme objects only), and p-value.
coeftFrame

Author(s)
Gregory R. Warnes <greg@warnes.net>

See Also
coeft, lm, summary.lm

Examples

# mean and confidence interval
ci(rnorm(10))

# binomial proportion and exact confidence interval
b <- rbinom(prob=0.75, size=1, n=20)
ci.binom(b) # direct call
class(b) <- c('binom')
ci(b) # indirect call

# confidence intervals for regression parameteres
data(state)
reg <- lm(Area ~ Population, data=as.data.frame(state.x77))
ci(reg)

# mer example
library(nlme)
Orthodont$AgeGroup <- gtools::quantcut(Orthodont$age)
fm2 <- lme(distance ~ Sex + AgeGroup, data = Orthodont, random=~1|Subject)
ci(fm2)

coeffFrame

Return model parameters in a data frame

Description

Fits a model to each subgroup defined by by, then returns a data frame with one row for each fit and one column for each parameter.

Usage

coeffFrame(mod, data, by = NULL, fit.on = TRUE, fitfun,
    keep.unused.levels = TRUE, byvar.sep = "\001", ...)
Arguments

- **mod**: a model formula, to be passed to by fitfun.
- **data**: a data frame, row subsets of which will be used as the data argument to fitfun.
- **by**: names of columns in x that will be used to define the subgroups.
- **fit.on**: a logical vector indicating which rows of x are to be used to fit the model (like the subset argument in a lot of other functions). Can be given in terms of variables in x.
- **fitfun**: a model fitting function (e.g., lm, nls). More specifically, a function that expects at least a formula object (as the first argument) and a data.frame object (passed as an argument named data) and returns a model object for which a coef method has been defined (e.g., coef.lm, coef.nls) to extract fit values of model parameters.
- **keep.unused.levels**: Include rows in output for all unique values of by, even those which were excluded by fit.on. The default value TRUE should be left alone if you are going to go on to pass the result to backfit.
- **byvar.sep**: passed to frameApply, used to form the subsets of the data.
- **...**: other arguments to pass to fitfun.

Value

A data frame with a row for each unique row of x[by], and column for each model parameter, as well as columns specified in by.

Author(s)

Jim Rogers <james.a.rogers@pfizer.com>

Examples

```r
# load example data
library(gtools)
data(ELISA)

# Coefficients for four parameter logistic fits:
coefFrame(log(Signal) ~ SSfpl(log(Concentration), A, B, xmid, scal),
  data = ELISA, fitfun = nls,
  by = c("PlateDay", "Read"),
  fit.on = Description == "Standard" & Concentration != 0)

# Coefficients for linear fits:
coefFrame(log(Signal) ~ log(Concentration),
  data = ELISA, fitfun = lm,
  by = c("PlateDay", "Read"),
  fit.on = Description == "Standard" & Concentration != 0 )

# Example passing arguments to fitfun, and example of
# error handling during model fitting:
```
Cross Table

CrossTabulation with Tests for Factor Independence

Description

An implementation of a cross-tabulation function with output similar to S-Plus crosstabs() and SAS Proc Freq (or SPSS format) with Chi-square, Fisher and McNemar tests of the independence of all table factors.

Usage

CrossTable(x, y, digits=3, max.width = 5, expected=FALSE, prop.r=TRUE, prop.c=TRUE, prop.t=TRUE, prop.chisq=TRUE, chisq = FALSE, fisher=FALSE, mcnemar=FALSE, resid=FALSE, sresid=FALSE, asresid=FALSE, missing.include=FALSE, format=c("SAS","SPSS"), dnn = NULL, ...)

Arguments

x A vector or a matrix. If y is specified, x must be a vector
y A vector in a matrix or a dataframe
digits Number of digits after the decimal point for cell proportions
max.width In the case of a 1 x n table, the default will be to print the output horizontally. If the number of columns exceeds max.width, the table will be wrapped for each successive increment of max.width columns. If you want a single column vertical table, set max.width to 1
expected If TRUE, chisq will be set to TRUE and expected cell counts from the \( \chi^2 \) will be included
prop.r If TRUE, row proportions will be included
prop.c If TRUE, column proportions will be included
prop.t If TRUE, table proportions will be included
prop.chisq If TRUE, chi-square contribution of each cell will be included
chisq If TRUE, the results of a chi-square test will be included
fisher If TRUE, the results of a Fisher Exact test will be included
mcnemar If TRUE, the results of a McNemar test will be included
resid If TRUE, residual (Pearson) will be included

ELISA$Signal[1] &lt;- NA
coeffFrame(log(Signal) - log(Concentration),
data = ELISA, fitfun = lm, na.action = na.fail,
by = c("PlateDay", "Read"),
fit.on = Description == "Standard" & Concentration != 0 )

CrossTable

Cross Tabulation with Tests for Factor Independence

Description

An implementation of a cross-tabulation function with output similar to S-Plus crosstabs() and SAS Proc Freq (or SPSS format) with Chi-square, Fisher and McNemar tests of the independence of all table factors.

Usage

CrossTable(x, y, digits=3, max.width = 5, expected=FALSE, prop.r=TRUE, prop.c=TRUE, prop.t=TRUE, prop.chisq=TRUE, chisq = FALSE, fisher=FALSE, mcnemar=FALSE, resid=FALSE, sresid=FALSE, asresid=FALSE, missing.include=FALSE, format=c("SAS","SPSS"), dnn = NULL, ...)

Arguments

x A vector or a matrix. If y is specified, x must be a vector
y A vector in a matrix or a dataframe
digits Number of digits after the decimal point for cell proportions
max.width In the case of a 1 x n table, the default will be to print the output horizontally. If the number of columns exceeds max.width, the table will be wrapped for each successive increment of max.width columns. If you want a single column vertical table, set max.width to 1
expected If TRUE, chisq will be set to TRUE and expected cell counts from the \( \chi^2 \) will be included
prop.r If TRUE, row proportions will be included
prop.c If TRUE, column proportions will be included
prop.t If TRUE, table proportions will be included
prop.chisq If TRUE, chi-square contribution of each cell will be included
chisq If TRUE, the results of a chi-square test will be included
fisher If TRUE, the results of a Fisher Exact test will be included
mcnemar If TRUE, the results of a McNemar test will be included
resid If TRUE, residual (Pearson) will be included

ELISA$Signal[1] &lt;- NA
coeffFrame(log(Signal) - log(Concentration),
data = ELISA, fitfun = lm, na.action = na.fail,
by = c("PlateDay", "Read"),
fit.on = Description == "Standard" & Concentration != 0 )
CrossTable

sresid     If TRUE, standardized residual will be included
asresid    If TRUE, adjusted standardized residual will be included
missing.include
           If TRUE, then remove any unused factor levels
format     Either SAS (default) or SPSS, depending on the type of output desired.
dnn         the names to be given to the dimensions in the result (the dimnames names).
            ... optional arguments

Details

A summary table will be generated with cell row, column and table proportions and marginal totals and proportions. Expected cell counts can be printed if desired (if `chisq = TRUE`). In the case of a 2 x 2 table, both corrected and uncorrected values will be included for appropriate tests. In the case of tabulating a single vector, cell counts and table proportions will be printed.

Note: If 'x' is a vector and 'y' is not specified, no statistical tests will be performed, even if any are set to TRUE.

Value

A list with multiple components including key table data and statistical test results, where performed.

t: An n by m matrix containing table cell counts
prop.col: An n by m matrix containing cell column proportions
prop.row: An n by m matrix containing cell row proportions
prop.tbl: An n by m matrix containing cell table proportions
chisq: Results from the Chi-Square test. A list with class 'htest'. See `?chisq.test` for details
chisq.corr: Results from the corrected Chi-Square test. A list with class 'htest'. See `?chisq.test` for details. ONLY included in the case of a 2 x 2 table.
fisher.ts: Results from the two-sided Fisher Exact test. A list with class 'htest'. See `?fisher.test` for details. ONLY included if 'fisher' = TRUE.
fisher.lt: Results from the Fisher Exact test with HA = "less". A list with class 'htest'. See `?fisher.test` for details. ONLY included if 'fisher' = TRUE and in the case of a 2 x 2 table.
fisher.gt: Results from the Fisher Exact test with HA = "greater". A list with class 'htest'. See `?fisher.test` for details. ONLY included if 'fisher' = TRUE and in the case of a 2 x 2 table.
mcnemar: Results from the McNemar test. A list with class 'htest'. See `?mcnemar.test` for details. ONLY included if 'mcnemar' = TRUE.
mcnemar.corr: Results from the corrected McNemar test. A list with class 'htest'. See `?mcnemar.test` for details. ONLY included if 'mcnemar' = TRUE and in the case of a 2 x 2 table.
resid/sresid/asresid: Pearson Residuals (from chi-square tests).

Author(s)

Marc Schwartz <marc_schwartz@comcast.net>. Original version posted to r-devel on Jul 27, 2002. SPSS format modifications added by Nitin Jain based upon code provided by Dirk Enzmann <dirk.enzmann@jura.uni-hamburg.de>
estimable

See Also
xtabs, table, prop.table

Examples

# Simple cross tabulation of education versus prior induced abortions
# using infertility data
data(infert, package = "datasets")
CrossTable(infert$education, infert$induced, expected = TRUE)
CrossTable(infert$education, infert$induced, expected = TRUE, format="SAS")
CrossTable(infert$education, infert$induced, expected = TRUE, format="SPSS")
CrossTable(warpbreaks$wool, warpbreaks$tension, dnn = c("Wool", "Tension"))

estimable

Contrasts and estimable linear functions of model coefficients

Description

Compute and test contrasts and other estimable linear functions of model coefficients for for lm,
glm, lme, and geese objects

Usage

estimable(obj, cm, beta0, conf.int=NULL, show.beta0, ...)
## Default S3 method:
estimable(obj, cm, beta0, conf.int=NULL, show.beta0, joint.test=FALSE, ...)

## S3 method for class 'mlm'
estimable(obj, cm, beta0, conf.int=NULL, show.beta0, ...)

Arguments

obj Regression (lm, glm, lme, mlm) object.

cm Vector, List, or Matrix specifying estimable linear functions or contrasts. See below for details.

beta0 Vector of null hypothesis values

conf.int Confidence level. If provided, confidence intervals will be computed.

joint.test Logical value. If TRUE a 'joint' Wald test for the hypothesis \( L\beta = \beta_0 \) is performed. Otherwise 'row-wise' tests are performed, i.e. \( (L\beta)_i = \beta_{0i} \).

show.beta0 Logical value. If TRUE a column for beta0 will be included in the output table. Defaults to TRUE when beta0 is specified, FALSE otherwise.

... ignored
estimable computes an estimate, test statistic, significance test, and (optional) confidence interval for each linear functions of the model coefficients specified by cm.

The estimable function(s) may be specified via a vector, list, or matrix. If cm is a vector, it should contained named elements each of which gives the coefficient to be applied to the corresponding parameter. These coefficients will be used to construct the contrast matrix, with unspecified model parameters assigned zero coefficients. If cm is a list, it should contain one or more coefficient vectors, which will be used to construct rows of the contrast matrix. If cm is a matrix, column names must match (a subset of) the model parameters, and each row should contain the corresponding coefficient to be applied. Model parameters which are not present in the set of column names of cm will be set to zero.

The estimates and their variances are obtained by applying the contrast matrix (generated from) cm to the model estimates variance-covariance matrix. Degrees of freedom are obtained from the appropriate model terms.

The user is responsible for ensuring that the specified linear functions are meaningful.

For computing contrasts among levels of a single factor, fit.contrast may be more convenient.

For computing contrasts between two specific combinations of model parameters, the contrast function in Frank Harrell’s ‘rms’ library (formerly ‘Design’) may be more convenient.

Value

Returns a matrix with one row per linear function. Columns contain the beta0 value (optional, see show.beta0 above), estimated coefficients, standard errors, t values, degrees of freedom, two-sided p-values, and the lower and upper endpoints of the 1-alpha confidence intervals.

Note

The estimated fixed effect parameters of lme objects may have different degrees of freedom. If a specified contrast includes nonzero coefficients for parameters with differing degrees of freedom, the smallest number of degrees of freedom is used and a warning message is issued.

Author(s)

BXC (Bendix Carstensen) <bxc@novonordisk.com>, Gregory R. Warnes <greg@warnes.net>, Soren Hojsgaard <sorenh@agrsci.dk>, and Randall C Johnson <rjohnson@ncifcrf.gov>

See Also

fit.contrast.lm, lme, contrasts, contrast

Examples

```r
# setup example data
y <- rnorm(100)
x <- cut(rnorm(100, mean=y, sd=0.25),c(-4,-1.5,0,1.5,4))
levels(x) <- c("A","B","C","D")
x2 <- rnorm(100, mean=y, sd=0.5)
```
# simple contrast and confidence interval
reg <- lm(y ~ x)
estimable(reg, c(0, 1, 0, -1)) # full coefficient vector
estimable(reg, c("xB"=1,"xD"=-1)) # just the nonzero terms

# fit a spline with a single knot at 0.5 and plot the *pointwise*
# confidence intervals
library(gplots)
pm <- pmax(x2-0.5, 0) # knot at 0.5
reg2 <- lm(y ~ x + x2 + pm)
range <- seq(-2, 2, , 0.5)
tmp <- estimable(reg2, cm=cbind(
    '(Intercept)'=1,
    'xC'=1,
    'x2'=range,
    'pm'=pmax(range-0.5, 0)
),
    conf.int=0.95)
plotCI(x=range, y=tmp[, 1], li=tmp[, 6], ui=tmp[, 7])

# fit both linear and quasi-Poisson models to iris data, then compute
# joint confidence intervals on contrasts for the Species and
# Sepal.Width by Species interaction terms.
data(iris)
lm1 <- lm(Sepal.Length ~ Sepal.Width + Species + Sepal.Width:Species, data=iris)
glm1 <- glm(Sepal.Length ~ Sepal.Width + Species + Sepal.Width:Species, data=iris, family=quasipoisson("identity"))

cm <- rbind(
    'Setosa vs. Versicolor' = c(0, 0, 1, 0, 1, 0),
    'Setosa vs. Virginica' = c(0, 0, 0, 1, 0, 1),
    'Versicolor vs. Virginica' = c(0, 0, 1,-1, 1,-1)
)
estimable(lm1, cm)
estimable(glm1, cm)

---

**fast.pcomp**

*Efficient computation of principal components and singular value decompositions.*

**Description**

The standard `prcomp` and `svd` function are very inefficient for wide matrixes. `fast.pcomp` and `fast.svd` are modified versions which are efficient even for matrixes that are very wide.
Usage

fast.prcomp(x, retx = TRUE, center = TRUE, scale. = FALSE, tol = NULL)
fast.svd(x, nu = min(n, p), nv = min(n, p), ...)

Arguments

x data matrix
retx, center, scale., tol
See documentation for prcomp
nu, nv, ... See documentation for svd

Details

The current implementation of the function svd in S-Plus and R is much slower when operating on a matrix with a large number of columns than on the transpose of this matrix, which has a large number of rows. As a consequence, prcomp, which uses svd, is also very slow when applied to matrices with a large number of rows.

For R, the simple solution is to use La.svd instead of svd. A suitable patch to prcomp has been submitted. In the meantime, the function fast.prcomp has been provided as a short-term workaround.

For S-Plus the solution is to replace the standard svd with a version that checks the dimensions of the matrix, and performs the computation on the transposed the matrix if it is wider than tall.

For R:

fast.prcomp is a modified version of prcomp that calls La.svd instead of svd
fast.svd is simply a wrapper around La.svd.

For S-Plus:

fast.prcomp is a modified version of prcomp that calls fast.svd instead of svd
fast.svd checks the dimensions of the matrix. When it is wider than tall, it transposes the input matrix and calls svd. It then swaps u and v and returns the result. Otherwise, it just calls svd and returns the results unchanged.

Value

See the documentation for prcomp or svd.

Author(s)

Modifications by Gregory R. Warnes <greg@warne.net>

See Also

prcomp, svd, La.svd
fit.contrast

Examples

```r
# create test matrix
set.seed(4943546)
nr <- 50
nc <- 2000
x <- matrix(rnorm(nr*nc), nrow=nr, ncol=nc)
tx <- t(x)

# SVD directly on matrix is SLOW:
system.time(val.x <- svd(x)$u)

# SVD on t(matrix) is FAST:
system.time(val.tx <- svd(tx)$v)

# and the results are equivalent:
max(abs(val.x) - abs(val.tx))

# Time gap dissapears using fast.svd:
library(stats)

# prcomp directly on matrix is SLOW:
system.time(pr.x <- prcomp(x))

# prcomp.fast is much faster
system.time(fast.pr.x <- fast.prcomp(x))

# and the results are equivalent
max(pr.x$sdev - fast.pr.x$sdev)
max(abs(pr.x$rotation[,1:49]) - abs(fast.pr.x$rotation[,1:49]))
max(abs(pr.x$x) - abs(fast.pr.x$x))

# (except for the last and least significant component):
max(abs(pr.x$rotation[,50]) - abs(fast.pr.x$rotation[,50]))
```

Description

Compute and test arbitrary contrasts for regression objects.
Usage

fit.contrast(model, varname, coeff, ...)
## S3 method for class 'lm'
fit.contrast(model, varname, coeff, showall=FALSE,
             conf.int=NULL, df=FALSE, ...)
## S3 method for class 'lme'
fit.contrast(model, varname, coeff, showall=FALSE,
             conf.int=NULL, df=FALSE, ...)

Arguments

- **model**: regression (lm, glm, aov, lme) object for which the contrast(s) will be computed.
- **varname**: variable name
- **coeff**: vector or matrix specifying contrasts (one per row).
- **showall**: return all regression coefficients. If TRUE, all model coefficients will be returned. If FALSE (the default), only the coefficients corresponding to the specified contrast will be returned.
- **conf.int**: numeric value on (0,1) or NULL. If a numeric value is specified, confidence intervals with nominal coverage probability `conf.int` will be computed. If NULL, confidence intervals will not be computed.
- **df**: boolean indicating whether to return a column containing the degrees of freedom.
- **...**: optional arguments provided by methods.

Details

Computes the specified contrast(s) by re-fitting the model with the appropriate arguments. A contrast of the form \( c(1,0,0,-1) \) would compare the mean of the first group with the mean of the fourth group.

Value

Returns a matrix containing estimated coefficients, standard errors, t values, two-sided p-values. If df is TRUE, an additional column containing the degrees of freedom is included. If conf.int is specified lower and upper confidence limits are also returned.

Author(s)

Gregory R. Warnes <greg@warnes.net>

References

Venables & Ripley, Section 6.2
See Also

`lm`, `contrasts`, `contr.treatment`, `contr.poly`. Computation and testing of General Linear Hypothesis: `glh.test`, Computation and testing of estimable functions of model coefficients: `estimable`, `make.contrasts`

Examples

```r
y <- rnorm(100)
x <- cut(rnorm(100, mean=y, sd=0.25),c(-4,-1.5,0,1.5,4))
reg <- lm(y ~ x)
summary(reg)

# look at the group means
gm <- sapply(split(y,x),mean)
gm

# mean of 1st group vs mean of 4th group
fit.contrast(reg, x, c( 1, 0, 0, -1 ) )
# estimate should be equal to:

# mean of 1st and 2nd groups vs mean of 3rd and 4th groups
fit.contrast(reg, x, c(-1/2, -1/2, 1/2, 1/2) )
# estimate should be equal to:
sum(-1/2*gm[1], -1/2*gm[2], 1/2*gm[3], 1/2*gm[4])

# mean of 1st group vs mean of 2nd, 3rd and 4th groups
fit.contrast(reg, x, c(-3/3, 1/3, 1/3, 1/3) )
# estimate should be equal to:
sum(-3/3*gm[1], 1/3*gm[2], 1/3*gm[3], 1/3*gm[4])

# all at once
cmat <- rbind( "1 vs 4" =c(-1, 0, 0, 1),
    "1+2 vs 3+4"=c(-1/2,-1/2, 1/2, 1/2),
    "1 vs 2+3+4"=c(-3/3, 1/3, 1/3, 1/3))
fit.contrast(reg,x,cmat)

# x2 <- rnorm(100,mean=y,sd=0.5)
reg2 <- lm(y ~ x + x2 )
fit.contrast(reg2,x,c(-1,0,0,1))

# Example for Analysis of Variance
#
set.seed(83215)
Genotype <- sample(c("WT","KO"), 1000, replace=TRUE)
Time <- factor(sample(1:3, 1000, replace=TRUE))
y <- rnorm(1000)
data <- data.frame(y, Genotype, Time)
```
# Compute Contrasts & obtain 95% confidence intervals

```r
model <- aov( y ~ Genotype + Time + Genotype:Time, data=data )
fit.contrast( model, "Genotype", rbind("KO vs WT"=c(-1,1)), conf=0.95 )
fit.contrast( model, "Time",
    rbind("1 vs 2"=c(-1,1,0),
    "2 vs 3"=c(0,-1,1)),
    conf=0.95 )
```

```r
cm.G <- rbind("KO vs WT"=c(-1,1))
cm.T <- rbind("1 vs 2"=c(-1,1,0),
    "2 vs 3"=c(0,-1,1))
```

# Compute contrasts and show SSQ decompositions

```r
model <- aov( y ~ Genotype + Time + Genotype:Time, data=data,
    contrasts=list(Genotype=make.contrasts(cm.G),
                  Time=make.contrasts(cm.T)) )
```

```r
summary(model, split=list( Genotype=list( "KO vs WT"=1 ),
                        Time = list( "1 vs 2" = 1,
                                      "2 vs 3" = 2 ) ) )
```

# example for lme

```r
library(nlme)
data(Orthodont)
fm1 <- lme(distance ~ Sex, data = Orthodont,random=~1|Subject)
```

# Contrast for sex. This example is equivalent to standard treatment
# contrast.

```r
# fit.contrast(fm1, "Sex", c(-1,1), conf.int=0.95 )
#
# and identical results can be obtained using lme built-in 'intervals'
# intervals(fm1)
```

# Cut age into quantile groups & compute some contrasts

```r
Orthodont$AgeGroup <- gtools::quantcut(Orthodont$age)
fm2 <- lme(distance ~ Sex + AgeGroup, data = Orthodont,random=~1|Subject)
```

```r
# fit.contrast(fm2, "AgeGroup", rbind("Linear"=c(-2,-1,1,2),
                       "U-Shaped"=c(-1,1,1,-1),
                       "Change-Point at 11"=c(-1,-1,1,1)),
                       conf.int=0.95)
```
Test a General Linear Hypothesis for a Regression Model

Description

Test, print, or summarize a general linear hypothesis for a regression model

Usage

```r
glh.test(reg, cm, d=rep(0, nrow(cm))
## S3 method for class 'glh.test'
print(x, digits=4,...)
## S3 method for class 'glh.test'
summary(object, digits=4,...)
```

Arguments

- `reg`: Regression model
- `cm`: matrix. Each row specifies a linear combination of the coefficients
- `d`: vector specifying the null hypothesis values for each linear combination
- `x, object`: glh.test object
- `digits`: number of digits
- `...`: optional parameters (ignored)

Details

Test the general linear hypothesis $C\hat{\beta} = d$ for the regression model `reg`.

The test statistic is obtained from the formula:

$$f = \frac{(C\hat{\beta} - d)'(C(X'X)^{-1}C')(C\hat{\beta} - d)/r}{SSE/(n - p)}$$

Under the null hypothesis, $f$ will follow a F-distribution with $r$ and $n-p$ degrees of freedom.

Value

Object of class c("glh.test","htest") with elements:

- `call`: Function call that created the object
- `statistic`: F statistic
- `parameter`: vector containing the numerator ($r$) and denominator ($n-p$) degrees of freedom
- `p.value`: p-value
**estimate**
computed estimate for each row of cm

**null.value**
d

**method**
description of the method

**data.name**
name of the model given for reg

**matrix**
matrix specifying the general linear hypothesis (cm)

**Note**
When using treatment contrasts (the default) the first level of the factors are subsumed into the intercept term. The estimated model coefficients are then contrasts versus the first level. This should be taken into account when forming contrast matrixes, particularly when computing contrasts that include 'baseline' level of factors.

See the comparison with fit.contrast in the examples below.

**Author(s)**
Gregory R. Warnes <greg@warnes.net>

**References**
R.H. Myers, Classical and Modern Regression with Applications, 2nd Ed, 1990, p. 105

**See Also**
fit.contrast, estimable, contrasts

**Examples**

```r
# fit a simple model
y <- rnorm(100)
x <- cut(rnorm(100, mean=y, sd=0.25),c(-4,-1.5,0,1.5,4))
reg <- lm(y ~ x)
summary(reg)

# test both group 1 = group 2 and group 3 = group 4
# *Note the 0 in the column for the intercept term*
C <- rbind(c(0,-1,0,0), c(0,0,-1,1))
ret <- glh.test(reg, C)
ret # same as 'print(ret)'
summary(ret)

# To compute a contrast between the first and second level of the factor
# 'x' using 'fit.contrast' gives:
fit.contrast( reg, x,c(1,-1,0,0) )

# To test this same contrast using 'glh.test', use a contrast matrix
# with a zero coefficient for the intercept term. See the Note section,
```
make.contrasts

    # above, for an explanation.
    C <- rbind( c(0,-1,0,0) )
    glh.test( reg, C )

make.contrasts 

Construct a User-Specified Contrast Matrix

Description

Construct a user-specified contrast matrix.

Usage

make.contrasts(contr, how.many = ncol(contr))

Arguments

contr 
vector or matrix specifying contrasts (one per row).

how.many 
dimensions of the desired contrast matrix. This must equal the number of levels of the target factor variable.

Details

This function converts human-readable contrasts into the form that R requires for computation.
Specifying a contrast row of the form c(1, 0, 0, -1) creates a contrast that will compare the mean of the first group with the mean of the fourth group.

Value

make.contrasts returns a matrix with dimensions (how.many, how.many) containing the specified contrasts augmented (if necessary) with orthogonal "filler" contrasts.
This matrix can then be used as the argument to contrasts or to the contrasts argument of model functions (eg, lm).

Author(s)

Gregory R. Warnes <greg@warnes.net>

See Also

lm, contrasts, contr.treatment, contr.poly, Computation and testing of General Linear Hypothesis: glh.test, Computation and testing of estimable functions of model coefficients: estimable, Estimate and Test Contrasts for a previously fit linear model: fit.contrast
Examples

```r
set.seed(4684)
y <- rnorm(100)
x.true <- rnorm(100, mean=y, sd=0.25)
x <- factor(cut(x.true, c(-4, -1.5, 0, 1.5, 4)))
reg <- lm(y ~ x)
summary(reg)

# Mirror default treatment contrasts
test <- make.contrasts(rbind( c(-1,1,0,0), c(-1,0,1,0), c(-1,0,0,1) ))
lm( y ~ x, contrasts=list(x = test ) )

# Specify some more complicated contrasts
#   - mean of 1st group vs mean of 4th group
#   - mean of 1st and 2nd groups vs mean of 3rd and 4th groups
#   - mean of 1st group vs mean of 2nd, 3rd and 4th groups
cmat <- rbind("1 vs 4" =c(-1, 0, 0, 1),
               "1+2 vs 3+4"=c(-1/2,-1/2, 1/2, 1/2),
               "1 vs 2+3+4"=c(-3/3, 1/3, 1/3, 1/3))

summary(lm( y ~ x, contrasts=list(x=make.contrasts(cmat) )))
# or
contrasts(x) <- make.contrasts(cmat)
summary(lm( y ~ x ) )

# or use contrasts.lm
reg <- lm(y ~ x)
fit.contrast( reg, "x", cmat )

# compare with values computed directly using group means
gm <- sapply(split(y,x),mean)
gm
```

# Example for Analysis of Variance

```r
set.seed(03215)
Genotype <- sample(c("WT","KO"), 1000, replace=TRUE)
Time <- factor(sample(1:3, 1000, replace=TRUE))
data <- data.frame(y, Genotype, Time)
y <- rnorm(1000)
data <- data.frame(y, Genotype, as.factor(Time))

# Compute Contrasts & obtain 95% confidence intervals
model <- aov( y ~ Genotype + Time + Genotype:Time, data=data )
fit.contrast( model, "Genotype", rbind("KO vs WT"=c(-1,1) ), conf=0.95 )
```
fit.contrast( model, "Time",
    rbind("1 vs 2"=c(-1,1,0),
           "2 vs 3"=c(0,-1,1)),
    conf=0.95 )

cm.G <- rbind("KO vs WT"=c(-1,1,0))
cm.T <- rbind("1 vs 2"=c(-1,1,0),
                   "2 vs 3"=c(0,-1,1))

# Compute contrasts and show SSQ decompositions

model <- aov( y ~ Genotype + Time + Genotype:Time, data=data,
               contrasts=list(Genotype=make.contrasts(cm.G),
                              Time=make.contrasts(cm.T))
)

summary(model, split=list( Genotype=list( "KO vs WT"=1 ),
                              Time = list( "1 vs 2" = 1,
                                           "2 vs 3" = 2 ) ) )
Index

*Topic algebra
  fast.prcomp, 9
*Topic array
  fast.prcomp, 9
*Topic category
  CrossTable, 5
*Topic models
  coeffFrame, 3
  estimable, 7
  fit.contrast, 11
  glh.test, 15
  make.contrasts, 17
*Topic multivariate
  fast.prcomp, 9
*Topic regression
  ci, 2
  estimable, 7
  fit.contrast, 11
  glh.test, 15
  make.contrasts, 17
*Topic univar
  CrossTable, 5

  ci, 2
  coeffFrame, 3
  confint, 3
  contr.poly, 13, 17
  contr.treatment, 13, 17
  contrast, 8
  contrasts, 8, 13, 16, 17
  CrossTable, 5

  estimable, 7, 13, 16, 17

  fast.prcomp, 9
  fast.svd(fast.prcomp), 9
  fit.contrast, 8, 11, 16, 17
  glh.test, 13, 15, 17
  la.svd, 10

lm, 3, 8, 13, 17
lme, 8

make.contrasts, 13, 17

prcomp, 9, 10
print.glh.test (glh.test), 15
prop.table, 7

summary.glh.test (glh.test), 15
summary.lm, 3
svd, 9, 10

table, 7

xtabs, 7