

# Package ‘DAAG’

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**Version** 1.20

**Title** Data Analysis And Graphics data and functions

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**Description** various data sets used in examples and exercises in the book Maindonald, J.H. and Braun, W.J. (2003, 2007, 2010) ``Data Analysis and Graphics Using R".

**LazyLoad** true

**LazyData** true

**Depends** R (>= 2.10.0), lattice

**Imports** latticeExtra

**Suggests** leaps, oz, lme4, quantreg, knitr, boot, rpart, randomForest, MASS, survival

**ZipData** yes

**License** GPL-3

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DAAG-package	<i>The R DAAG Package</i>
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## Description

Various data sets and functions used or referred to in the book Maindonald, J.H. and Braun, W.J. (3rd edn 2010) "Data Analysis and Graphics Using R", plus other selected datasets and functions.

## Details

For a list of , use `library(help="DAAG")`.

## Author(s)

Author: John H Maindonald

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ACF1

*Aberrant Crypt Foci in Rat Colons***Description**

Numbers of aberrant crypt foci (ACF) in the section 1 of the colons of 22 rats subjected to a single dose of the carcinogen azoxymethane (AOM), sacrificed at 3 different times.

**Usage**

ACF1

**Format**

This data frame contains the following columns:

**count** The number of ACF observed in section 1 of each rat colon

**endtime** Time of sacrifice, in weeks following injection of AOM

**Source**

Ranjana P. Bird, Faculty of Human Ecology, University of Manitoba, Winnipeg, Canada.

**References**

E.A. McLellan, A. Medline and R.P. Bird. Dose response and proliferative characteristics of aberrant crypt foci: putative preneoplastic lesions in rat colon. *Carcinogenesis*, 12(11): 2093-2098, 1991.

**Examples**

```
sapply(split(ACF1$count,ACF1$endtime),var)
plot(count ~ endtime, data=ACF1, pch=16)
pause()
print("Poisson Regression - Example 8.3")
ACF.glm0 <- glm(formula = count ~ endtime, family = poisson, data = ACF1)
summary(ACF.glm0)

# Is there a quadratic effect?
pause()

ACF.glm <- glm(formula = count ~ endtime + I(endtime^2),
  family = poisson, data = ACF1)
summary(ACF.glm)

# But is the data really Poisson? If not, try quasipoisson:
pause()

ACF.glm <- glm(formula = count ~ endtime + I(endtime^2),
```

```
family = quasipoisson, data = ACF1)
summary(ACF.glm)
```

---

ais	<i>Australian athletes data set</i>
-----	-------------------------------------

---

## Description

These data were collected in a study of how data on various characteristics of the blood varied with sport body size and sex of the athlete.

## Usage

```
data(ais)
```

## Format

A data frame with 202 observations on the following 13 variables.

**rcc** red blood cell count, in  $10^{12}l^{-1}$

**wcc** white blood cell count, in  $10^{12}$  per liter

**hc** hematocrit, percent

**hg** hemoglobin concentration, in g per decaliter

**ferr** plasma ferritins, ng  $dl^{-1}$

**bmi** Body mass index, kg  $cm^{-2}10^2$

**ssf** sum of skin folds

**pcBfat** percent Body fat

**lbm** lean body mass, kg

**ht** height, cm

**wt** weight, kg

**sex** a factor with levels f m

**sport** a factor with levels B\_Ball Field Gym Netball Row Swim T\_400m T\_Sprnt Tennis W\_Polo

## Details

Do blood hemoglobin concentrations of athletes in endurance-related events differ from those in power-related events?

## Source

These data were the basis for the analyses that are reported in Telford and Cunningham (1991).

## References

Telford, R.D. and Cunningham, R.B. 1991. Sex, sport and body-size dependency of hematology in highly trained athletes. *Medicine and Science in Sports and Exercise* 23: 788-794.

align2D

*Function to align points from ordination with known locations***Description**

Find the linear transformation which, applied to one set of points in the ( $x$ ,  $y$ ) plane, gives the best match in a least squares sense to a second set of points.

**Usage**

```
align2D(lat, long, x1, x2, wts=NULL)
```

**Arguments**

lat	Latitude or other co-ordinate of point to align to
long	Longitude or other co-ordinate of point to align to
x1	First coordinate of point to align
x2	First coordinate of point to align
wts	If non-NULL, specifies weights for the points.

**Details**

Achieves the best match, in a least squares sense, between an ordination and known locations in two-dimensional space.

**Value**

fitlat	Fitted values of lat
fitlong	Fitted values of long
lat	Input values of lat
long	Input values of long

**Note**

An ordination that is designed to reproduce distances between points is specified only to within an arbitrary rotation about the centroid. What linear transformation of the points ( $x_1$ ,  $x_2$ ) given by the ordination gives the best match to the known co-ordinates?

**Author(s)**

John H Maindonald



## Examples

```

if(require(DAAG)&require(oz)){
  aupts <- cmdscale(audists)
  xy <- align2D(lat = aulatlong$latitude, long = aulatlong$longitude,
               x1 = aupts[, 1], x2 = aupts[, 2], wts = NULL)

  oz()
  fitcoords <- align2D(lat=aulatlong$latitude,
                      long=aulatlong$longitude,
                      x1=aupts[,1], x2 = aupts[,2],
                      wts=NULL)

  x <-with(fitcoords,
          as.vector(rbind(lat, fitlat, rep(NA,length(lat)))))
  y <-with(fitcoords,
          as.vector(rbind(long, fitlong, rep(NA,length(long)))))
  points(aulatlong, col="red", pch=16, cex=1.5)
  lines(x, y, col="gray40", lwd=3)
}

## The function is currently defined as
function(lat, long, x1, x2, wts=NULL){
  ## Get best fit in space of (latitude, longitude)
  if(is.null(wts))wts <- rep(1,length(x1))
  fitlat <- predict(lm(lat ~ x1+x2, weights=wts))
  fitlong <- predict(lm(long ~ x1+x2, weights=wts))
  list(fitlat = fitlat, fitlong=fitlong, lat=lat, long=long)
}

```

---

allbacks

---

*Measurements on a Selection of Books*


---

## Description

The allbacks data frame gives measurements on the volume and weight of 15 books, some of which are softback (pb) and some of which are hardback (hb). Area of the hardback covers is also included.

## Usage

```
allbacks
```

## Format

This data frame contains the following columns:

**volume** book volumes in cubic centimeters  
**area** hard board cover areas in square centimeters  
**weight** book weights in grams  
**cover** a factor with levels hb hardback, pb paperback

**Source**

The bookshelf of J. H. Maindonald.

**Examples**

```
print("Multiple Regression - Example 6.1")
attach(allbacks)
volume.split <- split(volume, cover)
weight.split <- split(weight, cover)
plot(weight.split$hb ~ volume.split$hb, pch=16, xlim=range(volume), ylim=range(weight),
      ylab="Weight (g)", xlab="Volume (cc)")
points(weight.split$pb ~ volume.split$pb, pch=16, col=2)
pause()

allbacks.lm <- lm(weight ~ volume+area)
summary(allbacks.lm)
detach(allbacks)
pause()

anova(allbacks.lm)
pause()

model.matrix(allbacks.lm)
pause()

print("Example 6.1.1")
allbacks.lm0 <- lm(weight ~ -1+volume+area, data=allbacks)
summary(allbacks.lm0)
pause()

print("Example 6.1.2")
oldpar <- par(mfrow=c(2,2))
plot(allbacks.lm0)
par(oldpar)
allbacks.lm13 <- lm(weight ~ -1+volume+area, data=allbacks[-13,])
summary(allbacks.lm13)
pause()

print("Example 6.1.3")
round(coef(allbacks.lm0),2) # Baseline for changes
round(lm.influence(allbacks.lm0)$coef,2)
```

---

anesthetic

*Anesthetic Effectiveness*


---

**Description**

Thirty patients were given an anesthetic agent maintained at a predetermined level (conc) for 15 minutes before making an incision. It was then noted whether the patient moved, i.e. jerked or twisted.

**Usage**

```
anesthetic
```

**Format**

This data frame contains the following columns:

**move** a binary numeric vector coded for patient movement (0 = no movement, 1 = movement)

**conc** anesthetic concentration

**logconc** logarithm of concentration

**nomove** the complement of move

**Details**

The interest is in estimating how the probability of jerking or twisting varies with increasing concentration of the anesthetic agent.

**Source**

unknown

**Examples**

```
print("Logistic Regression - Example 8.1.4")

z <- table(anesthetic$nomove, anesthetic$conc)
tot <- apply(z, 2, sum)          # totals at each concentration
prop <- z[2, ]/(tot)            # proportions at each concentration
oprop <- sum(z[2, ])/sum(tot)    # expected proportion moving if concentration had no effect
conc <- as.numeric(dimnames(z)[[2]])
plot(conc, prop, xlab = "Concentration", ylab = "Proportion", xlim = c(.5, 2.5),
      ylim = c(0, 1), pch = 16)
chw <- par()$cxy[1]
text(conc - 0.75 * chw, prop, paste(tot), adj = 1)
abline(h = oprop, lty = 2)

pause()

anes.logit <- glm(nomove ~ conc, family = binomial(link = logit),
  data = anesthetic)
anova(anes.logit)
summary(anes.logit)
```

---

ant111b	<i>Averages by block of corn yields, for treatment 111 only</i>
---------	---

---

### Description

These data frames have averages by blocks (parcels) for the treatment 111.

### Usage

```
ant111b
```

### Format

A data frame with 36 observations on 9 variables.

**site** a factor with levels (ant111b:) DBAN LFAN NSAN ORAN OVAN TEAN WEAN WLAN

**parcel** a factor with levels I II III IV

**code** a numeric vector

**island** a numeric vector

**id** a numeric vector

**plot** a numeric vector

**trt** a numeric vector

**ears** a numeric vector

**harvwt** a numeric vector

### Source

Andrews DF; Herzberg AM, 1985. Data. A Collection of Problems from Many Fields for the Student and Research Worker. Springer-Verlag. (pp. 339-353)

---

antigua	<i>Averages by block of yields for the Antigua Corn data</i>
---------	--

---

### Description

These data frames have yield averages by blocks (parcels). The ant111b data set is a subset of this.

### Usage

```
antigua
```

**Format**

A data frame with 324 observations on 7 variables.

**id** a numeric vector

**site** a factor with 8 levels.

**block** a factor with levels I II III IV

**plot** a numeric vector

**trt** a factor consisting of 12 levels

**ears** a numeric vector; note that -9999 is used as a missing value code.

**harvwt** a numeric vector; the average yield

**Source**

Andrews DF; Herzberg AM, 1985. Data. A Collection of Problems from Many Fields for the Student and Research Worker. Springer-Verlag. (pp. 339-353)

---

appletaste

*Tasting experiment that compared four apple varieties*


---

**Description**

Each of 20 tasters each assessed three out of the four varieties. The experiment was conducted according to a balanced incomplete block design.

**Usage**

```
data(appletaste)
```

**Format**

A data frame with 60 observations on the following 3 variables.

**aftertaste** a numeric vector Apple samples were rated for aftertaste, by making a mark on a continuous scale that ranged from 0 (extreme dislike) to 150 (like very much).

**panelist** a factor with levels a b c d e f g h i j k l m n o p q r s t

**product** a factor with levels 298 493 649 937

**Examples**

```
data(appletaste)
appletaste.aov <- aov(aftertaste ~ panelist + product, data=appletaste)
termplot(appletaste.aov)
```

---

audists

*Road distances between 10 Australian cities*


---

### Description

Distances between the Australian cities of Adelaide, Alice, Brisbane, Broome, Cairns, Canberra, Darwin, Melbourne, Perth and Sydney

### Usage

```
audists
```

### Format

The format is: Class 'dist', i.e., a distance matrix.

### Source

Australian road map

### Examples

```
data(audists)
audists.cmd <- cmdscale(audists)
xyplot(audists.cmd[,2] ~ audists.cmd[,1], groups=row.names(audists.cmd),
       panel = function(x, y, subscripts, groups)
         ltext(x = x, y = y, label = groups[subscripts],
              cex=1, fontfamily = "HersheySans"))
```

---

aulatlong

*Latitudes and longitudes for ten Australian cities*


---

### Description

Latitudes and longitudes for Adelaide, Alice, Brisbane, Broome, Cairns, Canberra, Darwin, Melbourne, Perth and Sydney; i.e., for the cities to which the road distances in audists relate.

### Usage

```
aulatlong
```

### Format

A data frame with 10 observations on the following 2 variables.

latitude Latitude, as a decimal number

longitude Longitude, as a decimal number

**Source**

Map of Australia showing latitude and longitude information.

**Examples**

```
data(aulatlong)
## maybe str(aulatlong) ; plot(aulatlong) ...
```

---

austpop

*Population figures for Australian States and Territories*

---

**Description**

Population figures for Australian states and territories for 1917, 1927, ..., 1997.

**Usage**

```
austpop
```

**Format**

This data frame contains the following columns:

**year** a numeric vector

**NSW** New South Wales population counts

**Vic** Victoria population counts

**Qld** Queensland population counts

**SA** South Australia population counts

**WA** Western Australia population counts

**Tas** Tasmania population counts

**NT** Northern Territory population counts

**ACT** Australian Capital Territory population counts

**Aust** Population counts for the whole country

**Source**

Australian Bureau of Statistics

**Examples**

```

print("Looping - Example 1.7")

growth.rates <- numeric(8)
for (j in seq(2,9)) {
  growth.rates[j-1] <- (austpop[9, j]-austpop[1, j])/austpop[1, j] }
growth.rates <- data.frame(growth.rates)
row.names(growth.rates) <- names(austpop[c(-1,-10)])
# Note the use of row.names() to name the rows of the data frame
growth.rates

pause()
print("Avoiding Loops - Example 1.7b")

sapply(austpop[, -c(1,10)], function(x){(x[9]-x[1])/x[1]})

pause()
print("Plot - Example 1.8a")
attach(austpop)
plot(year, ACT, type="l") # Join the points ("l" = "line")
detach(austpop)

pause()
print("Exerice 1.12.9")
attach(austpop)
oldpar <- par(mfrow=c(2,4))
for (i in 2:9){
  plot(austpop[,1], log(austpop[, i]), xlab="Year",
       ylab=names(austpop)[i], pch=16, ylim=c(0,10))}
par(oldpar)
detach(austpop)

```

---

bestsetNoise

*Best Subset Selection Applied to Noise*


---

**Description**

Best subset selection applied to completely random noise. This function demonstrates how variable selection techniques in regression can often err in including explanatory variables that are indistinguishable from noise.

**Usage**

```

bestsetNoise(m = 100, n = 40, method = "exhaustive", nvmax = 3,
             X = NULL, y=NULL, intercept=TRUE,
             print.summary = TRUE, really.big = FALSE, ...)

bestset.noise(m = 100, n = 40, method = "exhaustive", nvmax = 3,

```



```

X = NULL, y=NULL, intercept=TRUE,
print.summary = TRUE, really.big = FALSE, ...)

bsnCV(m = 100, n = 40, method = "exhaustive", nvmax = 3,
      X = NULL, y=NULL, intercept=TRUE, nfolds = 2,
      print.summary = TRUE, really.big = FALSE)

bsnOpt(X = matrix(rnorm(25 * 10), ncol = 10), y = NULL, method = "exhaustive",
        nvmax = NULL, nbest = 1, intercept = TRUE, criterion = "cp",
        tcrit = NULL, print.summary = TRUE, really.big = FALSE,
        ...)

bsnVaryNvar(m = 100, nvar = nvmax:50, nvmax = 3, method = "exhaustive",
            intercept=TRUE,
            plotit = TRUE, xlab = "# of variables from which to select",
            ylab = "p-values for t-statistics", main = paste("Select 'best'",
                                                            nvmax, "variables"),
            details = FALSE, really.big = TRUE, smooth = TRUE)

```

## Arguments

<code>m</code>	the number of observations to be simulated, ignored if <code>X</code> is supplied.
<code>n</code>	the number of predictor variables in the simulated model, ignored if <code>X</code> is supplied.
<code>method</code>	Use exhaustive search, or backward selection, or forward selection, or sequential replacement.
<code>nvmax</code>	Number of explanatory variables in model.
<code>X</code>	Use columns from this matrix. Alternatively, <code>X</code> may be a data frame, in which case a model matrix will be formed from it. If not <code>NULL</code> , <code>m</code> and <code>n</code> are ignored.
<code>y</code>	If not supplied, random normal noise will be generated.
<code>nbest</code>	Number of models, for each choice of number of columns of explanatory variables, to return ( <code>bsnOpt</code> ). If <code>tcrit</code> is non- <code>NULL</code> , it may be important to set this greater than one, in order to have a good chance of finding models with minimum absolute <i>t</i> -statistic greater than <code>tcrit</code> .
<code>intercept</code>	Should an intercept be added?
<code>nvar</code>	range of number of candidate variables ( <code>bsnVaryVvar</code> ).
<code>nfolds</code>	For splitting the data into training and test sets, the number of folds.
<code>criterion</code>	Criterion to use in choosing between models with different numbers of explanatory variables ( <code>bsnOpt</code> ). Alternatives are "bic", or "cip" or "adjr2".
<code>tcrit</code>	Consider only those models for which the minimum absolute <i>t</i> -statistic is greater than <code>tcrit</code> .
<code>print.summary</code>	Should summary information be printed.
<code>plotit</code>	Plot a graph? ( <code>bsnVaryVvar</code> )
<code>xlab</code>	<i>x</i> -label for graph ( <code>bsnVaryVvar</code> )

<code>ylab</code>	y-label for graph ( <code>bsnVaryVvar</code> .)
<code>main</code>	main title for graph ( <code>bsnVaryVvar</code> .)
<code>details</code>	Return detailed output list ( <code>bsnVaryVvar</code> )
<code>really.big</code>	Set to TRUE to allow (currently) for more than 50 explanatory variables.
<code>smooth</code>	Fit smooth to graph? ( <code>bsnVaryVvar</code> ).
<code>...</code>	Additional arguments, to be passed through to <code>regsubsets()</code> .

## Details

If  $X$  is not supplied, and in any case for `bsnVaryNvar`, a set of  $n$  predictor variables are simulated as independent standard normal, i.e.  $N(0,1)$ , variates. Additionally a  $N(0,1)$  response variable is simulated. The function `bsnOpt` selects the ‘best’ model with `nvmax` or fewer explanatory variables, where the argument `criterion` specifies the criterion that will be used to choose between models with different numbers of explanatory columns. Other functions select the ‘best’ model with `nvmax` explanatory columns. In any case, the selection is made using the `regsubsets()` function from the `leaps` package. (The `leaps` package must be installed for this function to work.)

The function `bsnCV` splits the data (randomly) into `nfolds` (2 or more) parts. It puts each part aside in turn for use to fit the model (effectively, test data), with the remaining data used for selecting the variables that will be used for fitting. One model fit is returned for each of the `nfolds` parts.

The function `bsnVaryVvar` makes repeated calls to `bestsetNoise`

## Value

`bestsetNoise` returns the `lm` model object for the “best” model with `nvmax` explanatory columns.

`bsnCV` returns as many models as there are folds.

`bsnVaryVvar` silently returns either (`details=FALSE`) a matrix that has  $p$ -values of the coefficients for the ‘best’ choice of model for each different number of candidate variables, or (`details=TRUE`) a list with elements:

<code>coef</code>	A matrix of sets of regression coefficients
<code>SE</code>	A matrix of standard errors
<code>pval</code>	A matrix of $p$ -values

Matrices have one row for each choice of `nvar`. The statistics returned are for the ‘best’ model with `nvmax` explanatory variables.

`bsnOpt` silently returns a list with elements:

<code>u1</code>	‘best’ model ( <code>lm</code> object) with <code>nvmax</code> or fewer columns of predictors. If <code>tcrit</code> is non-NULL, and there is no model for which all coefficients have $t$ -statistics less than <code>tcrit</code> in absolute value, <code>u1</code> will be NULL.
-----------------	---

`tcrit` For each model, the minimum of the absolute values of the  $t$ -statistics. `regsubsets_obj` The object returned by the call to `regsubsets`.

**Note**

These functions are primarily designed to demonstrate the biases that can be expected, relative to theoretical estimates of standard errors of parameters and other fitted model statistics, when there is prior selection of the columns that are to be included in the model. With the exception of `bsnVaryNvar`, they can also be used with an  $X$  and  $y$  for actual data. In that case, the  $p$ -values should be compared with those obtained from repeated use of the function where  $y$  is random noise, as a check on the extent of selection effects.

**Author(s)**

J.H. Maindonald

**See Also**

[lm](#)

**Examples**

```
leaps.out <- try(require(leaps, quietly=TRUE))
leaps.out.log <- is.logical(leaps.out)
if ((leaps.out.log==TRUE)&(leaps.out==TRUE)){
  bestsetNoise(20,6) # 'best' 3-variable regression for 20 simulated observations
                    # on 7 unrelated variables (including the response)
  bsnCV(20,6) # 'best' 3-variable regressions (one for each fold) for 20
              # simulated observations on 7 unrelated variables
              # (including the response)
  bsnVaryNvar(m = 50, nvar = 3:6, nvmax = 3, method = "exhaustive",
              plotit=FALSE, details=TRUE)
  bsnOpt()
}
```

---

biomass

*Biomass Data*

---

**Description**

The biomass data frame has 135 rows and 8 columns. The rainforest data frame is a subset of this one.

**Usage**

biomass

**Format**

This data frame contains the following columns:

**dbh** a numeric vector

**wood** a numeric vector

**bark** a numeric vector

**fac26** a factor with 3 levels

**root** a numeric vector

**rootsk** a numeric vector

**branch** a numeric vector

**species** a factor with levels *Acacia mabellae*, *C. fraseri*, *Acmena smithii*, *B. myrtifolia*

**Source**

J. Ash, Australian National University

**References**

Ash, J. and Helman, C. (1990) Floristics and vegetation biomass of a forest catchment, Kioloa, south coastal N.S.W. *Cunninghamia*, 2: 167-182.

---

bomregions

*Australian and Related Historical Annual Climate Data, by region*


---

**Description**

Australian regional temperature data, Australian regional rainfall data, and Annual SOI, are given for the years 1900-2008. The regional rainfall and temperature data are area-weighted averages for the respective regions. The Southern Oscillation Index (SOI) is the difference in barometric pressure at sea level between Tahiti and Darwin.

**Usage**

bomregions

**Format**

This data frame contains the following columns:

**Year** Year

**eastAVt** Eastern temperature

**seAVt** Southeastern region average temperature (degrees C)

**southAVt** Southern temperature

**swAVt** Southwestern temperature

**westAVt** Western temperature  
**northAVt** Northern temperature  
**mdbAVt** Murray-Darling basin temperature  
**auAVt** Australian average temperature, area-weighted mean  
**eastRain** Eastern rainfall  
**seRain** Southeast Australian annual rainfall (mm)  
**southRain** Southern rainfall  
**swRain** Southwest rainfall  
**westRain** Western rainfall  
**northRain** Northern rainfall  
**mdbRain** Murray-Darling basin rainfall  
**auRain** Australian average rainfall, area weighted  
**SOI** Annual average Southern Oscillation Index  
**co2mlo** Moana Loa CO<sub>2</sub> concentrations, from 1959  
**co2law** Moana Loa CO<sub>2</sub> concentrations, 1900 to 1978  
**CO<sub>2</sub>** CO<sub>2</sub> concentrations, composite series  
**sunspot** Annual average sunspot counts

## Source

Australian Bureau of Meteorology web pages:

<http://www.bom.gov.au/climate/change/> <http://www.bom.gov.au/climate/current/soihtml.shtml>

Regions are identified on a map that can be found at: [http://www.bom.gov.au/silo/products/cli\\_chg/rain\\_timeseries.shtml](http://www.bom.gov.au/silo/products/cli_chg/rain_timeseries.shtml)

The CO<sub>2</sub> series co2law, from <http://cdiac.ornl.gov/trends/co2/lawdome.html>, is from Law Dome ice core data.

The CO<sub>2</sub> series co2mlo is from Dr. Pieter Tans, NOAA/ESRL ([www.esrl.noaa.gov/gmd/ccgg/trends/](http://www.esrl.noaa.gov/gmd/ccgg/trends/))

The series CO<sub>2</sub> is a composite series, obtained by adding 0.46 to the Law data for 1900 to 1958, then following this with the Moana Loa data that is available from 1959. The addition of 0.46 is designed so that the averages from the two series agree for the period 1959 to 1968

Sunspot data is from <http://sidc.oma.be/sunspot-data/>

## References

D.M. Etheridge, L.P. Steele, R.L. Langenfelds, R.J. Francey, J.-M. Barnola and V.I. Morgan, 1998, *Historical CO<sub>2</sub> records from the Law Dome DE08, DE08-2, and DSS ice cores*, in Trends: A Compendium of Data on Global Change, on line at Carbon Dioxide Information Analysis Center, Oak Ridge National Laboratory, U.S. Department of Energy, Oak Ridge, Tenn., U.S.A. <http://cdiac.ornl.gov/trends/co2/lawdome.html>

Lavery, B., Joung, G. and Nicholls, N. 1997. An extended high-quality historical rainfall dataset for Australia. Australian Meteorological Magazine, 46, 27-38.

Nicholls, N., Lavery, B., Frederiksen, C. and Drosowsky, W. 1996. Recent apparent changes in relationships between the El Nino – southern oscillation and Australian rainfall and temperature. Geophysical Research Letters 23: 3357-3360.

## Examples

```
plot(ts(bomregions[, c("mdbRain", "SOI")], start=1900),
     panel=function(y,...)panel.smooth(bomregions$Year, y,...))
avrain <- bomregions[, "mdbRain"]
xbomsoi <- with(bomregions, data.frame(Year=Year, SOI=SOI,
                                       cuberootRain=avrain^0.33))
xbomsoi$trendSOI <- lowess(xbomsoi$SOI, f=0.1)$y
xbomsoi$trendRain <- lowess(xbomsoi$cuberootRain, f=0.1)$y
xbomsoi$detrendRain <-
  with(xbomsoi, cuberootRain - trendRain + mean(trendRain))
xbomsoi$detrendSOI <-
  with(xbomsoi, SOI - trendSOI + mean(trendSOI))
## Plot time series avrain and SOI: ts object xbomsoi
plot(ts(xbomsoi[, c("cuberootRain", "SOI")], start=1900),
     panel=function(y,...)panel.smooth(xbomsoi$Year, y,...),
     xlab = "Year", main="", ylim=list(c(250, 800), c(-20, 25)))
par(mfrow=c(1,2))
rainpos <- pretty(xbomsoi$cuberootRain^3, 6)
plot(cuberootRain ~ SOI, data = xbomsoi,
     ylab = "Rainfall (cube root scale)", yaxt="n")
axis(2, at = rainpos^0.33, labels=paste(rainpos))
mtext(side = 3, line = 0.8, "A", adj = -0.025)
with(xbomsoi, lines(lowess(cuberootRain ~ SOI, f=0.75)))
plot(detrendRain ~ detrendSOI, data = xbomsoi,
     xlab="Detrended SOI", ylab = "Detrended rainfall", yaxt="n")
axis(2, at = rainpos^0.33, labels=paste(rainpos))
with(xbomsoi, lines(lowess(detrendRain ~ detrendSOI, f=0.75)))
mtext(side = 3, line = 0.8, "B", adj = -0.025)
par(mfrow=c(1,1))
```

---

bomregions2012

---

*Australian and Related Historical Annual Climate Data, by region*


---

## Description

Australian regional temperature data, Australian regional rainfall data, Annual SOI, and average sunspot counts, are given for the years 1900-2011 or 1900-2012.. The regional rainfall and temperature data are area-weighted averages for the respective regions. The Southern Oscillation Index (SOI) is the difference in barometric pressure at sea level between Tahiti and Darwin.

**Usage**

bomregions2012

**Format**

This data frame contains the following columns:

**Year** Year

**eastAVt** Eastern temperature

**seAVt** Southeastern region average temperature (degrees C)

**southAVt** Southern temperature

**swAVt** Southwestern temperature

**westAVt** Western temperature

**northAVt** Northern temperature

**mdbAVt** Murray-Darling basin temperature

**auAVt** Australian average temperature, area-weighted mean

**eastRain** Eastern rainfall

**seRain** Southeast Australian annual rainfall (mm)

**southRain** Southern rainfall

**swRain** Southwest rainfall

**westRain** Western rainfall

**northRain** Northern rainfall

**mdbRain** Murray-Darling basin rainfall

**auRain** Australian average rainfall, area weighted

**SOI** Annual average Southern Oscillation Index

**co2mlo** Moana Loa CO2 concentrations, from 1959

**co2law** Moana Loa CO2 concentrations, 1900 to 1978

**CO2** CO2 concentrations, composite series

**sunspot** Annual average sunspot counts

**Source**

Rainfall, temperature and SOI data are from Australian Bureau of Meteorology web pages:

<http://www.bom.gov.au/climate/change/> <http://www.bom.gov.au/climate/current/soihtml.shtml>

Regions are identified on a map that can be found at: [http://www.bom.gov.au/silo/products/cli\\_chg/rain\\_timeseries.shtml](http://www.bom.gov.au/silo/products/cli_chg/rain_timeseries.shtml)

The CO2 series co2law, from <http://cdiac.ornl.gov/trends/co2/lawdome.html>, is from Law Dome ice core data.

The CO2 series co2mlo is from Dr. Pieter Tans, NOAA/ESRL ([www.esrl.noaa.gov/gmd/ccgg/trends/](http://www.esrl.noaa.gov/gmd/ccgg/trends/))

The series CO2 is a composite series, obtained by adding 0.46 to the Law data for 1900 to 1958, then following this with the Moana Loa data that is available from 1959. The addition of 0.46 is designed so that the averages from the two series agree for the period 1959 to 1968

Sunspot data is from <http://sidc.oma.be/sunspot-data/>

## References

D.M. Etheridge, L.P. Steele, R.L. Langenfelds, R.J. Francey, J.-M. Barnola and V.I. Morgan, 1998, *Historical CO2 records from the Law Dome DE08, DE08-2, and DSS ice cores*, in Trends: A Compendium of Data on Global Change, on line at Carbon Dioxide Information Analysis Center, Oak Ridge National Laboratory, U.S. Department of Energy, Oak Ridge, Tenn., U.S.A. <http://cdiac.ornl.gov/trends/co2/lawdome.html>

Lavery, B., Joung, G. and Nicholls, N. 1997. An extended high-quality historical rainfall dataset for Australia. Australian Meteorological Magazine, 46, 27-38.

Nicholls, N., Lavery, B., Frederiksen, C. and Drosowsky, W. 1996. Recent apparent changes in relationships between the El Nino – southern oscillation and Australian rainfall and temperature. Geophysical Research Letters 23: 3357-3360.

SIDC-team, World Data Center for the Sunspot Index, Royal Observatory of Belgium, Monthly Report on the International Sunspot Number, online catalogue of the sunspot index: <http://www.sidc.be/sunspot-data/>, 1900-2011

## Examples

```
plot(ts(bomregions2011[, c("mdbRain", "SOI")], start=1900),
      panel=function(y,...)panel.smooth(bomregions2011$Year, y,...))
avrain <- bomregions2011[, "mdbRain"]
xbomsoi <- with(bomregions2011, data.frame(Year=Year, SOI=SOI,
      cuberootRain=avrain^0.33))
xbomsoi$trendSOI <- lowess(xbomsoi$SOI, f=0.1)$y
xbomsoi$trendRain <- lowess(xbomsoi$cuberootRain, f=0.1)$y
xbomsoi$detrendRain <-
  with(xbomsoi, cuberootRain - trendRain + mean(trendRain))
xbomsoi$detrendSOI <-
  with(xbomsoi, SOI - trendSOI + mean(trendSOI))
## Plot time series avrain and SOI: ts object xbomsoi
plot(ts(xbomsoi[, c("cuberootRain", "SOI")], start=1900),
      panel=function(y,...)panel.smooth(xbomsoi$Year, y,...),
      xlab = "Year", main="", ylim=list(c(250, 800), c(-20, 25)))
par(mfrow=c(1,2))
rainpos <- pretty(xbomsoi$cuberootRain^3, 6)
plot(cuberootRain ~ SOI, data = xbomsoi,
      ylab = "Rainfall (cube root scale)", yaxt="n")
axis(2, at = rainpos^0.33, labels=paste(rainpos))
with(xbomsoi, lines(lowess(cuberootRain ~ SOI, f=0.75)))
plot(detrendRain ~ detrendSOI, data = xbomsoi,
      xlab="Detrended SOI", ylab = "Detrended rainfall", yaxt="n")
axis(2, at = rainpos^0.33, labels=paste(rainpos))
with(xbomsoi, lines(lowess(detrendRain ~ detrendSOI, f=0.75)))
par(mfrow=c(1,1))
```



---

bomsoi*Southern Oscillation Index Data*

---

**Description**

The Southern Oscillation Index (SOI) is the difference in barometric pressure at sea level between Tahiti and Darwin. Annual SOI and Australian rainfall data, for the years 1900-2001, are given. Australia's annual mean rainfall is an area-weighted average of the total annual precipitation at approximately 370 rainfall stations around the country.

**Usage**

bomsoi

**Format**

This data frame contains the following columns:

**Year** a numeric vector

**Jan** average January SOI values for each year

**Feb** average February SOI values for each year

**Mar** average March SOI values for each year

**Apr** average April SOI values for each year

**May** average May SOI values for each year

**Jun** average June SOI values for each year

**Jul** average July SOI values for each year

**Aug** average August SOI values for each year

**Sep** average September SOI values for each year

**Oct** average October SOI values for each year

**Nov** average November SOI values for each year

**Dec** average December SOI values for each year

**SOI** a numeric vector consisting of average annual SOI values

**avrain** a numeric vector consisting of a weighted average annual rainfall at a large number of Australian sites

**NTrain** Northern Territory rain

**northRain** north rain

**seRain** southeast rain

**eastRain** east rain

**southRain** south rain

**swRain** southwest rain

## Source

Australian Bureau of Meteorology web pages:

<http://www.bom.gov.au/climate/change/rain02.txt> and <http://www.bom.gov.au/climate/current/soihtml1.shtml>

## References

Nicholls, N., Lavery, B., Frederiksen, C. and Drosowsky, W. 1996. Recent apparent changes in relationships between the El Nino – southern oscillation and Australian rainfall and temperature. *Geophysical Research Letters* 23: 3357-3360.

## Examples

```
plot(ts(bomsoi[, 15:14], start=1900),
      panel=function(y,...)panel.smooth(1900:2005, y,...))
pause()

# Check for skewness by comparing the normal probability plots for
# different a, e.g.
par(mfrow = c(2,3))
for (a in c(50, 100, 150, 200, 250, 300))
  qqnorm(log(bomsoi[, "avrain"] - a))
  # a = 250 leads to a nearly linear plot

pause()

par(mfrow = c(1,1))
plot(bomsoi$SOI, log(bomsoi$avrain - 250), xlab = "SOI",
      ylab = "log(avrain = 250)")
lines(lowess(bomsoi$SOI)$y, lowess(log(bomsoi$avrain - 250))$y, lwd=2)
# NB: separate lowess fits against time
lines(lowess(bomsoi$SOI, log(bomsoi$avrain - 250)))
pause()

xbomsoi <-
  with(bomsoi, data.frame(SOI=SOI, cuberootRain=avrain^0.33))
xbomsoi$trendSOI <- lowess(xbomsoi$SOI)$y
xbomsoi$trendRain <- lowess(xbomsoi$cuberootRain)$y
rainpos <- pretty(bomsoi$avrain, 5)
with(xbomsoi,
      {plot(cuberootRain ~ SOI, xlab = "SOI",
            ylab = "Rainfall (cube root scale)", yaxt="n")
       axis(2, at = rainpos^0.33, labels=paste(rainpos))
      })
## Relative changes in the two trend curves
lines(lowess(cuberootRain ~ SOI))
lines(lowess(trendRain ~ trendSOI), lwd=2)
})
pause()

xbomsoi$detrendRain <-
  with(xbomsoi, cuberootRain - trendRain + mean(trendRain))
```

```

xbomsoi$detrendSOI <-
  with(xbomsoi, SOI - trendSOI + mean(trendSOI))
oldpar <- par(mfrow=c(1,2), pty="s")
plot(cuberootRain ~ SOI, data = xbomsoi,
     ylab = "Rainfall (cube root scale)", yaxt="n")
axis(2, at = rainpos^0.33, labels=paste(rainpos))
with(xbomsoi, lines(lowess(cuberootRain ~ SOI)))
plot(detrendRain ~ detrendSOI, data = xbomsoi,
     xlab="Detrended SOI", ylab = "Detrended rainfall", yaxt="n")
axis(2, at = rainpos^0.33, labels=paste(rainpos))
with(xbomsoi, lines(lowess(detrendRain ~ detrendSOI)))
pause()

par(oldpar)
attach(xbomsoi)
xbomsoi.ma0 <- arima(detrendRain, xreg=detrendSOI, order=c(0,0,0))
# ordinary regression model

xbomsoi.ma12 <- arima(detrendRain, xreg=detrendSOI,
                     order=c(0,0,12))
# regression with MA(12) errors -- all 12 MA parameters are estimated
xbomsoi.ma12
pause()

xbomsoi.ma12s <- arima(detrendRain, xreg=detrendSOI,
                      seasonal=list(order=c(0,0,1), period=12))
# regression with seasonal MA(1) (lag 12) errors -- only 1 MA parameter
# is estimated
xbomsoi.ma12s
pause()

xbomsoi.maSel <- arima(x = detrendRain, order = c(0, 0, 12),
                      xreg = detrendSOI, fixed = c(0, 0, 0,
NA, rep(0, 4), NA, 0, NA, NA, NA, NA),
                      transform.pars=FALSE)
# error term is MA(12) with fixed 0's at lags 1, 2, 3, 5, 6, 7, 8, 10
# NA's are used to designate coefficients that still need to be estimated
# transform.pars is set to FALSE, so that MA coefficients are not
# transformed (see help(arima))

detach(xbomsoi)
pause()

Box.test(resid(lm(detrendRain ~ detrendSOI, data = xbomsoi)),
        type="Ljung-Box", lag=20)

pause()

attach(xbomsoi)
xbomsoi2.maSel <- arima(x = detrendRain, order = c(0, 0, 12),
                      xreg = poly(detrendSOI,2), fixed = c(0,
0, 0, NA, rep(0, 4), NA, 0, rep(NA,5)),
                      transform.pars=FALSE)

```

```
xbomsoi2.maSel
qqnorm(resid(xbomsoi.maSel, type="normalized"))
detach(xbomsoi)
```

bomsoi2001

*Southern Oscillation Index Data*

## Description

The Southern Oscillation Index (SOI) is the difference in barometric pressure at sea level between Tahiti and Darwin. Annual SOI and Australian rainfall data, for the years 1900-2001, are given. Australia's annual mean rainfall is an area-weighted average of the total annual precipitation at approximately 370 rainfall stations around the country.

## Usage

```
bomsoi2001
```

## Format

This data frame contains the following columns:

**Year** a numeric vector

**Jan** average January SOI values for each year

**Feb** average February SOI values for each year

**Mar** average March SOI values for each year

**Apr** average April SOI values for each year

**May** average May SOI values for each year

**Jun** average June SOI values for each year

**Jul** average July SOI values for each year

**Aug** average August SOI values for each year

**Sep** average September SOI values for each year

**Oct** average October SOI values for each year

**Nov** average November SOI values for each year

**Dec** average December SOI values for each year

**SOI** a numeric vector consisting of average annual SOI values

**avrain** a numeric vector consisting of a weighted average annual rainfall at a large number of Australian sites

## Source

Australian Bureau of Meteorology web pages:

<http://www.bom.gov.au/climate/change/rain02.txt> and <http://www.bom.gov.au/climate/current/soihtml.shtml>

## References

Nicholls, N., Lavery, B., Frederiksen, C. and Drosowsky, W. 1996. Recent apparent changes in relationships between the El Nino – southern oscillation and Australian rainfall and temperature. *Geophysical Research Letters* 23: 3357-3360.

## See Also

bomsoi

## Examples

```
bomsoi <- bomsoi2001
plot(ts(bomsoi[, 15:14], start=1900),
     panel=function(y,...)panel.smooth(1900:2001, y,...))
pause()

# Check for skewness by comparing the normal probability plots for
# different a, e.g.
par(mfrow = c(2,3))
for (a in c(50, 100, 150, 200, 250, 300))
  qqnorm(log(bomsoi[, "avrain"] - a))
  # a = 250 leads to a nearly linear plot

pause()

par(mfrow = c(1,1))
plot(bomsoi$SOI, log(bomsoi$avrain - 250), xlab = "SOI",
     ylab = "log(avrain = 250)")
lines(lowess(bomsoi$SOI)$y, lowess(log(bomsoi$avrain - 250))$y, lwd=2)
# NB: separate lowess fits against time
lines(lowess(bomsoi$SOI, log(bomsoi$avrain - 250)))
pause()

xbomsoi <-
  with(bomsoi, data.frame(SOI=SOI, cuberootRain=avrain^0.33))
xbomsoi$trendSOI <- lowess(xbomsoi$SOI)$y
xbomsoi$trendRain <- lowess(xbomsoi$cuberootRain)$y
rainpos <- pretty(bomsoi$avrain, 5)
with(xbomsoi,
     {plot(cuberootRain ~ SOI, xlab = "SOI",
           ylab = "Rainfall (cube root scale)", yaxt="n")
      axis(2, at = rainpos^0.33, labels=paste(rainpos))
## Relative changes in the two trend curves
      lines(lowess(cuberootRain ~ SOI))
      lines(lowess(trendRain ~ trendSOI), lwd=2)
     })
pause()

xbomsoi$detrendRain <-
  with(xbomsoi, cuberootRain - trendRain + mean(trendRain))
xbomsoi$detrendSOI <-
```

```

    with(xbomsoi, SOI - trendSOI + mean(trendSOI))
oldpar <- par(mfrow=c(1,2), pty="s")
plot(cuberootRain ~ SOI, data = xbomsoi,
     ylab = "Rainfall (cube root scale)", yaxt="n")
axis(2, at = rainpos^0.33, labels=paste(rainpos))
with(xbomsoi, lines(lowess(cuberootRain ~ SOI)))
plot(detrendRain ~ detrendSOI, data = xbomsoi,
     xlab="Detrended SOI", ylab = "Detrended rainfall", yaxt="n")
axis(2, at = rainpos^0.33, labels=paste(rainpos))
with(xbomsoi, lines(lowess(detrendRain ~ detrendSOI)))
pause()

par(oldpar)
attach(xbomsoi)
xbomsoi.ma0 <- arima(detrendRain, xreg=detrendSOI, order=c(0,0,0))
# ordinary regression model

xbomsoi.ma12 <- arima(detrendRain, xreg=detrendSOI,
                    order=c(0,0,12))
# regression with MA(12) errors -- all 12 MA parameters are estimated
xbomsoi.ma12
pause()

xbomsoi.ma12s <- arima(detrendRain, xreg=detrendSOI,
                    seasonal=list(order=c(0,0,1), period=12))
# regression with seasonal MA(1) (lag 12) errors -- only 1 MA parameter
# is estimated
xbomsoi.ma12s
pause()

xbomsoi.maSel <- arima(x = detrendRain, order = c(0, 0, 12),
                    xreg = detrendSOI, fixed = c(0, 0, 0,
                    NA, rep(0, 4), NA, 0, NA, NA, NA, NA),
                    transform.pars=FALSE)
# error term is MA(12) with fixed 0's at lags 1, 2, 3, 5, 6, 7, 8, 10
# NA's are used to designate coefficients that still need to be estimated
# transform.pars is set to FALSE, so that MA coefficients are not
# transformed (see help(arima))

detach(xbomsoi)
pause()

Box.test(resid(lm(detrendRain ~ detrendSOI, data = xbomsoi)),
        type="Ljung-Box", lag=20)

pause()

attach(xbomsoi)
xbomsoi2.maSel <- arima(x = detrendRain, order = c(0, 0, 12),
                    xreg = poly(detrendSOI,2), fixed = c(0,
                    0, 0, NA, rep(0, 4), NA, 0, rep(NA,5)),
                    transform.pars=FALSE)

xbomsoi2.maSel

```

```
qqnorm(resid(xbomsoi.maSel, type="normalized"))
detach(xbomsoi)
```

---

bostonc	<i>Boston Housing Data – Corrected</i>
---------	--

---

### Description

The corrected Boston housing data (from <http://lib.stat.cmu.edu/datasets/>).

### Usage

```
bostonc
```

### Format

A single vector containing the contents of "boston\\_corrected.txt".

### Source

Harrison, D. and Rubinfeld, D.L. 'Hedonic prices and the demand for clean air', J. Environ. Economics & Management, vol.5, 81-102, 1978. corrected by Kelley Pace (kpace@unix1.sncc.lsu.edu)

---

bounce	<i>Separate plotting positions for labels, to avoid overlap</i>
--------	---

---

### Description

Return univariate plotting positions in which neighboring points are separated, if and as necessary, so that they are the specified minimum distance apart.

### Usage

```
bounce(y, d, log = FALSE)
```

### Arguments

y	A numeric vector of plotting positions
d	Minimum required distance between neighboring positions
log	TRUE if values are will be plotted on a logarithmic scale.

### Details

The centroid(s) of groups of points that are moved relative to each other remain the same.

**Value**

A vector of values such that, when plotted along a line, neighboring points are the required minimum distance apart.

**Note**

If values are plotted on a logarithmic scale, d is the required distance apart on that scale. If a base other than 10 is required, set log equal to that base. (Note that base 10 is the default for plot with log=TRUE.)

**Author(s)**

John Maindonald

**See Also**

See also [onewayPlot](#)

**Examples**

```
bounce(c(4, 1.8, 2, 6), d=.4)
bounce(c(4, 1.8, 2, 6), d=.1, log=TRUE)
```

---

capstring

*Converts initial character of a string to upper case*

---

**Description**

This function is useful for use before plotting, if one wants capitalized axis labels or factor levels.

**Usage**

```
capstring(names)
```

**Arguments**

names	a character vector
-------	--------------------

**Value**

a character vector with upper case initial values

**Author(s)**

W.J. Braun



**Examples**

```
capstring(names(tinting)[c(3,4)])

library(lattice)
levels(tinting$agegp) <- capstring(levels(tinting$agegp))
xyplot(csoa ~ it | sex * agegp, data=tinting)
```

carprice

*US Car Price Data***Description**

U.S. data extracted from Cars93, a data frame in the MASS package.

**Usage**

```
carprice
```

**Format**

This data frame contains the following columns:

**Type** Type of car, e.g. Sporty, Van, Compact

**Min.Price** Price for a basic model

**Price** Price for a mid-range model

**Max.Price** Price for a ‘premium’ model

**Range.Price** Difference between Max.Price and Min.Price

**RoughRange** Rough.Range plus some  $N(0,.0001)$  noise

**gpm100** The number of gallons required to travel 100 miles

**MPG.city** Average number of miles per gallon for city driving

**MPG.highway** Average number of miles per gallon for highway driving

**Source**

MASS package

**References**

Venables, W.N.\ and Ripley, B.D., 4th edn 2002. Modern Applied Statistics with S. Springer, New York.

See also ‘R’ Complements to Modern Applied Statistics with S-Plus, available from <http://www.stats.ox.ac.uk/pub/MASS3/>

## Examples

```
print("Multicollinearity - Example 6.8")
pairs(carprice[,~c(1,8,9)])

carprice1.lm <- lm(gpm100 ~ Type+Min.Price+Price+Max.Price+Range.Price,
  data=carprice)
round(summary(carprice1.lm)$coef,3)
pause()

alias(carprice1.lm)
pause()

carprice2.lm <- lm(gpm100 ~ Type+Min.Price+Price+Max.Price+RoughRange, data=carprice)
round(summary(carprice2.lm)$coef, 2)
pause()

carprice.lm <- lm(gpm100 ~ Type + Price, data = carprice)
round(summary(carprice.lm)$coef,4)
pause()

summary(carprice1.lm)$sigma # residual standard error when fitting all 3 price variables
pause()

summary(carprice.lm)$sigma # residual standard error when only price is used
pause()

vif(lm(gpm100 ~ Price, data=carprice)) # Baseline Price
pause()

vif(carprice1.lm) # includes Min.Price, Price & Max.Price
pause()

vif(carprice2.lm) # includes Min.Price, Price, Max.Price & RoughRange
pause()

vif(carprice.lm) # Price alone
```

---

Cars93.summary

---

*A Summary of the Cars93 Data set*


---

## Description

The Cars93.summary data frame has 6 rows and 4 columns created from information in the Cars93 data set in the Venables and Ripley MASS package. Each row corresponds to a different class of car (e.g. Compact, Large, etc.).

## Usage

```
Cars93.summary
```

**Format**

This data frame contains the following columns:

**Min.passengers** minimum passenger capacity for each class of car

**Max.passengers** maximum passenger capacity for each class of car

**No.of.cars** number of cars in each class

**abbrev** a factor with levels C Compact, L Large, M Mid-Size, Sm Small, Sp Sporty, V Van

**Source**

Lock, R. H. (1993) 1993 New Car Data. Journal of Statistics Education 1(1)

**References**

MASS library

**Examples**

```
type <- Cars93.summary$abbrev
type <- Cars93.summary[,4]
type <- Cars93.summary[, "abbrev"]
type <- Cars93.summary[[4]] # Take the object that is stored
                             # in the fourth list element.

type
pause()

attach(Cars93.summary)
# R can now access the columns of Cars93.summary directly
abbrev
detach("Cars93.summary")
pause()

# To change the name of the \verb!abbrev! variable (the fourth column)
names(Cars93.summary)[4] <- "code"
pause()

# To change all of the names, try
names(Cars93.summary) <- c("minpass", "maxpass", "number", "code")
```

---

cerealsugar

---

*Percentage of Sugar in Breakfast Cereal*


---

**Description**

Measurements of sugar content in frosted flakes breakfast cereal.

Usage

cerealsugar

Format

A vector of 100 measurements.

---

cfseal	<i>Cape Fur Seal Data</i>
--------	---------------------------

---

Description

The cfseal data frame has 30 rows and 11 columns consisting of weight measurements for various organs taken from 30 Cape Fur Seals that died as an unintended consequence of commercial fishing.

Usage

cfseal

Format

This data frame contains the following columns:

- age** a numeric vector
- weight** a numeric vector
- heart** a numeric vector
- lung** a numeric vector
- liver** a numeric vector
- spleen** a numeric vector
- stomach** a numeric vector
- leftkid** a numeric vector
- rightkid** a numeric vector
- kidney** a numeric vector
- intestines** a numeric vector

Source

Stewardson, C.L., Hemsley, S., Meyer, M.A., Canfield, P.J. and Maindonald, J.H. 1999. Gross and microscopic visceral anatomy of the male Cape fur seal, *Arctocephalus pusillus pusillus* (Pinnepedia: Otariidae), with reference to organ size and growth. *Journal of Anatomy* (Cambridge) 195: 235-255. (WWF project ZA-348)

**Examples**

```
print("Allometric Growth - Example 5.7")

cfseal.lm <- lm(log(heart) ~ log(weight), data=cfseal); summary(cfseal.lm)
plot(log(heart) ~ log(weight), data = cfseal, pch=16, xlab = "Heart Weight (g, log scale)",
      ylab = "Body weight (kg, log scale)", axes=FALSE)
heartaxis <- 100*(2^seq(0,3))
bodyaxis <- c(20,40,60,100,180)
axis(1, at = log(bodyaxis), lab = bodyaxis)
axis(2, at = log(heartaxis), lab = heartaxis)
box()
abline(cfseal.lm)
```

cities

*Populations of Major Canadian Cities (1992-96)***Description**

Population estimates for several Canadian cities.

**Usage**

```
cities
```

**Format**

This data frame contains the following columns:

**CITY** a factor, consisting of the city names

**REGION** a factor with 5 levels (ATL=Atlantic, ON=Ontario, QC=Quebec, PR=Prairies, WEST=Alberta and British Columbia) representing the location of the cities

**POP1992** a numeric vector giving population in 1000's for 1992

**POP1993** a numeric vector giving population in 1000's for 1993

**POP1994** a numeric vector giving population in 1000's for 1994

**POP1995** a numeric vector giving population in 1000's for 1995

**POP1996** a numeric vector giving population in 1000's for 1996

**Source**

Statistics Canada

**Examples**

```
cities$have <- factor((cities$REGION=="ON")|(cities$REGION=="WEST"))
plot(POP1996~POP1992, data=cities, col=as.integer(cities$have))
```

---

codling	<i>Dose-mortality data, for fumigation of codling moth with methyl bromide</i>
---------	--

---

## Description

Data are from trials that studied the mortality response of codling moth to fumigation with methyl bromide.

## Usage

```
data(codling)
```

## Format

A data frame with 99 observations on the following 10 variables.

**dose** Injected dose of methyl bromide, in gm per cubic meter

**tot** Number of insects in chamber

**dead** Number of insects dying

**pobs** Proportion dying

**cm** Control mortality, i.e., at dose 0

**ct** Concentration-time sum

**Cultivar** a factor with levels BRAEBURN FUJI GRANNY Gala ROYAL Red Delicious Splendour

**gp** a factor which has a different level for each different combination of Cultivar, year and rep (replicate).

**year** a factor with levels 1988 1989

**numcm** a numeric vector: total number of control insects

## Details

The research that generated these data was in part funded by New Zealand pipfruit growers. The published analysis was funded by New Zealand pipfruit growers. See also sorption.

## Source

Maindonald, J.H.; Waddell, B.C.; Petry, R.J. 2001. Apple cultivar effects on codling moth (Lepidoptera: Tortricidae) egg mortality following fumigation with methyl bromide. *Postharvest Biology and Technology* 22: 99-110.

---

compareTreecalcs	<i>Error rate comparisons for tree-based classification</i>
------------------	---

---

## Description

Compare error rates, between different functions and different selection rules, for an approximately equal random division of the data into a training and test set.

## Usage

```
compareTreecalcs(x = yesno ~ ., data = spam7, cp = 0.00025,
                 fun = c("rpart", "randomForest"))
```

## Arguments

x	model formula
data	an data frame in which to interpret the variables named in the formula
cp	setting for the cost complexity parameter cp, used by rpart()
fun	one or both of "rpart" and "randomForest"

## Details

Data are randomly divided into two subsets, I and II. The function(s) are used in the standard way for calculations on subset I, and error rates returned that come from the calculations carried out by the function(s). Predictions are made for subset II, allowing the calculation of a completely independent set of error rates.

## Value

If rpart is specified in fun, the following:

rpSEcvI	the estimated cross-validation error rate when rpart() is run on the training data (I), and the one-standard error rule is used
rpcvI	the estimated cross-validation error rate when rpart() is run on subset I, and the model used that gives the minimum cross-validated error rate
rpSEtest	the error rate when the model that leads to rpSEcvI is used to make predictions for subset II
rptest	the error rate when the model that leads to rpcvI is used to make predictions for subset II
nSErule	number of splits required by the one standard error rule
nREmin	number of splits to give the minimum error

If rpart is specified in fun, the following:

rfcvI	the out-of-bag (OOB) error rate when randomForest() is run on subset I
rfctest	the error rate when the model that leads to rfcvI is used to make predictions for subset II

**Author(s)**

John Maindonald

---

component.residual	<i>Component + Residual Plot</i>
--------------------	----------------------------------

---

**Description**

Component + Residual plot for a term in a `lm` model.

**Usage**

```
component.residual(lm.obj, which = 1, xlab = "Component",  
  ylab = "C+R")
```

**Arguments**

<code>lm.obj</code>	A <code>lm</code> object
<code>which</code>	numeric code for the term in the <code>lm</code> formula to be plotted
<code>xlab</code>	label for the x-axis
<code>ylab</code>	label for the y-axis

**Value**

A scatterplot with a smooth curve overlaid.

**Author(s)**

J.H. Maindonald

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

```
mice12.lm <- lm(brainwt ~ bodywt + lsize, data=litters)  
oldpar <- par(mfrow = c(1,2))  
component.residual(mice12.lm, 1, xlab = "Body weight", ylab= "t(Body weight) + e")  
component.residual(mice12.lm, 2, xlab = "Litter size", ylab= "t(Litter size) + e")  
par(oldpar)
```



---

confusion	<i>Given actual and predicted group assignments, give the confusion matrix</i>
-----------	--

---

## Description

Given actual and predicted group assignments, give the confusion matrix

## Usage

```
confusion(actual, predicted, gpnames = NULL, rowcol=c("actual", "predicted"),  
printit = c("overall","confusion"), prior = NULL, digits=3)
```

## Arguments

actual	Actual (prior) group assignments
predicted	Predicted group assignments.
gpnames	Names for groups, if different from levels(actual)
rowcol	For predicted categories to appear as rows, specify rowcol="predicted"
printit	Character vector. Print "overall", or "confusion" matrix, or both.
prior	Prior probabilities for groups, if different from the relative group frequencies
digits	Number of decimal digits to display in printed output

## Details

Predicted group assignments should be estimated from cross-validation or from bootstrap out-of-bag data. Better still, work with assignments for test data that are completely separate from the data used to derive the model.

## Value

A list with elements overall (overall accuracy), confusion (confusion matrix) and prior (prior used for calculation of overall accuracy)

## Author(s)

John H Maindonald

## References

Maindonald and Braun: 'Data Analysis and Graphics Using R', 3rd edition 2010, Section 12.2.2

**Examples**

```

library(MASS)
library(DAAG)
cl <- lda(species ~ length+breadth, data=cuckoos, CV=TRUE)$class
confusion(cl, cuckoos$species)

## The function is currently defined as
function (actual, predicted, gpnames = NULL,
          rowcol = c("actual", "predicted"),
          printit = c("overall", "confusion"),
          prior = NULL, digits = 3)
{
  if (is.null(gpnames))
    gpnames <- levels(actual)
  if (is.logical(printit)){
    if(printit)printit <- c("overall", "confusion")
    else printit <- ""
  }
  tab <- table(actual, predicted)
  acctab <- t(apply(tab, 1, function(x) x/sum(x)))
  dimnames(acctab) <- list(Actual = gpnames, `Predicted (cv)` = gpnames)
  if (is.null(prior)) {
    relnum <- table(actual)
    prior <- relnum/sum(relnum)
    acc <- sum(tab[row(tab) == col(tab)])/sum(tab)
  }
  else {
    acc <- sum(prior * diag(acctab))
  }
  names(prior) <- gpnames
  if ("overall"%in%printit) {
    cat("Overall accuracy =", round(acc, digits), "\n")
    if(is.null(prior)){
      cat("This assumes the following prior frequencies:",
          "\n")
      print(round(prior, digits))
    }
  }
  if ("confusion"%in%printit) {
    cat("\nConfusion matrix", "\n")
    print(round(acctab, digits))
  }
  invisible(list(overall=acc, confusion=acctab, prior=prior))
}

```

## Description

Numbers are given in different categories of worker, in each of two investigations. The first source of information is the Board of Trade Census that was conducted on 1886. The second is a relatively informal survey conducted by US Bureau of Labor representatives in 1889, for use in official reports.

## Usage

```
data(cottonworkers)
```

## Format

A data frame with 14 observations on the following 3 variables.

**census1886** Numbers of workers in each of 14 different categories, according to the Board of Trade wage census that was conducted in 1886

**survey1889** Numbers of workers in each of 14 different categories, according to data collected in 1889 by the US Bureau of Labor, for use in a report to the US Congress and House of Representatives

**avwage** Average wage, in pence, as estimated in the US Bureau of Labor survey

## Details

The data in survey1889 were collected in a relatively informal manner, by approaching individuals on the street. Biases might therefore be expected.

## Source

United States congress, House of Representatives, Sixth Annual Report of the Commissioner of Labor, 1890, Part III, Cost of Living (Washington D.C. 1891); idem., Seventh Annual Report of the Commissioner of Labor, 1891, Part III, Cost of Living (Washington D.C. 1892)

Return of wages in the principal textile trades of the United Kingdom, with report therein. (P.P. 1889, LXX). United Kingdom Official Publication.

## References

Boot, H. M. and Maindonald, J. H. 2007. New estimates of age- and sex- specific earnings and the male-female earnings gap in the British cotton industry, 1833-1906. *Economic History Review*. Published online 28-Aug-2007 doi: 10.1111/j.1468-0289.2007.00398.x

## Examples

```
data(cottonworkers)
str(cottonworkers)
plot(survey1889 ~ census1886, data=cottonworkers)
plot(I(avwage*survey1889) ~ I(avwage*census1886), data=cottonworkers)
```

cps1

*Labour Training Evaluation Data***Description**

A non-experimental "control" group, used in various studies of the effect of a labor training program, alternative to the experimental control group in nswdemo.

**Usage**

cps1

**Format**

This data frame contains the following columns:

**trt** a numeric vector identifying the study in which the subjects were enrolled (0 = Control, 1 = treated).

**age** age (in years).

**educ** years of education.

**black** (0 = not black, 1 = black).

**hisp** (0 = not hispanic, 1 = hispanic).

**marr** (0 = not married, 1 = married).

**nodeg** (0 = completed high school, 1 = dropout).

**re74** real earnings in 1974.

**re75** real earnings in 1975.

**re78** real earnings in 1978.

**Details**

The cps1 and psid1 data sets are two non-experimental "control" groups, alternative to that in nswdemo, used in investigating whether use of such a non-experimental control group can be satisfactory. cps2 and cps3 are subsets of cps1, designed to be better matched to the experimental data than cps1. Similarly psid2 and psid3 are subsets of psid1, designed to be better matched to the experimental data than psid1.

**Source**

<http://www.nber.org/~rdehejia/nswdata.html>

## References

- Dehejia, R.H. and Wahba, S. 1999. Causal effects in non-experimental studies: re-evaluating the evaluation of training programs. *Journal of the American Statistical Association* 94: 1053-1062.
- Lalonde, R. 1986. Evaluating the economic evaluations of training programs. *American Economic Review* 76: 604-620.
- Smith, J. A. and Todd, P.E. 2005, "Does Matching overcome. LaLonde's critique of nonexperimental estimators", *Journal of Econometrics* 125: 305-353.
- Dehejia, R.H. 2005. Practical propensity score matching: a reply to Smith and Todd. *Journal of Econometrics* 125: 355-364.

---

 cps2

 Labour Training Evaluation Data
 

---

## Description

A non-experimental "control" group, used in various studies of the effect of a labor training program, alternative to the experimental control group in *nswdemo*.

## Usage

cps2

## Format

This data frame contains the following columns:

- trt** a numeric vector identifying the study in which the subjects were enrolled (0 = Control, 1 = treated).
- age** age (in years).
- educ** years of education.
- black** (0 = not black, 1 = black).
- hisp** (0 = not hispanic, 1 = hispanic).
- marr** (0 = not married, 1 = married).
- nodeg** (0 = completed high school, 1 = dropout).
- re74** real earnings in 1974.
- re75** real earnings in 1975.
- re78** real earnings in 1978.

## Details

The *cps1* and *psid1* data sets are two non-experimental "control" groups, alternative to that in *nswdemo*, used in investigating whether use of such a non-experimental control group can be satisfactory. *cps2* and *cps3* are subsets of *cps1*, designed to be better matched to the experimental data than *cps1*. Similary *psid2* and *psid3* are subsets of *psid1*, designed to be better matched to the experimental data than *psid1*.

## Source

<http://www.nber.org/~rdehejia/nswdata.html>

## References

Dehejia, R.H. and Wahba, S. 1999. Causal effects in non-experimental studies: re-evaluating the evaluation of training programs. *Journal of the American Statistical Association* 94: 1053-1062.

Lalonde, R. 1986. Evaluating the economic evaluations of training programs. *American Economic Review* 76: 604-620.

Smith, J. A. and Todd, P.E. 2005, "Does Matching overcome. LaLonde's critique of nonexperimental estimators", *Journal of Econometrics* 125: 305-353.

Dehejia, R.H. 2005. Practical propensity score matching: a reply to Smith and Todd. *Journal of Econometrics* 125: 355-364.

---

cps3

*Labour Training Evaluation Data*

---

## Description

A non-experimental "control" group, used in various studies of the effect of a labor training program, alternative to the experimental control group in nswdemo.

## Usage

cps3

## Format

This data frame contains the following columns:

**trt** a numeric vector identifying the study in which the subjects were enrolled (0 = Control, 1 = treated).

**age** age (in years).

**educ** years of education.

**black** (0 = not black, 1 = black).

**hisp** (0 = not hispanic, 1 = hispanic).

**marr** (0 = not married, 1 = married).

**nodeg** (0 = completed high school, 1 = dropout).

**re74** real earnings in 1974.

**re75** real earnings in 1975.

**re78** real earnings in 1978.

## Details

The cps1 and psid1 data sets are two non-experimental "control" groups, alternative to that in nswdemo, used in investigating whether use of such a non-experimental control group can be satisfactory. cps2 and cps3 are subsets of cps1, designed to be better matched to the experimental data than cps1. Similary psid2 and psid3 are subsets of psid1, designed to be better matched to the experimental data than psid1.

## Source

<http://www.nber.org/~rdehejia/nswdata.html>

## References

Dehejia, R.H. and Wahba, S. 1999. Causal effects in non-experimental studies: re-evaluating the evaluation of training programs. *Journal of the American Statistical Association* 94: 1053-1062.

Lalonde, R. 1986. Evaluating the economic evaluations of training programs. *American Economic Review* 76: 604-620.

Smith, J. A. and Todd, P.E. 2005, "Does Matching overcome. LaLonde?s critique of nonexperimental estimators", *Journal of Econometrics* 125: 305-353.

Dehejia, R.H. 2005. Practical propensity score matching: a reply to Smith and Todd. *Journal of Econometrics* 125: 355-364.

---

cricketer

*Lifespans of UK 1st class cricketers born 1840-1960*


---

## Description

Year and birth, lifespan, etc, of British first class cricketers, born 1840-1960, whose handedness could be determined from information in the Who's who of cricketers. The status (alive=0, dead=1), and lifetime or lifespan, is for 1992.

## Usage

```
data(cricketer)
```

## Format

A data frame with 5960 observations on the following 8 variables.

left a factor with levels right left

year numeric, year of birth

life numeric, lifetime or lifespan to 1992

dead numeric (0 = alive (censored), 1 = dead, in 1992)

acd numeric (0 = not accidental or not dead, 1 = accidental death)

kia numeric (0 = not killed in action, 1 = killed in action)

inbed numeric (0 = did not die in bed, 1 = died in bed)

cause a factor with levels alive acd (accidental death) inbed (died in bed)

**Details**

Note that those 'killed in action' (mostly during World Wars I and II) form a subset of those who died by accident.

**Source**

John Aggleton, Martin Bland. Data were collated as described in Aggleton et al.

**References**

Aggleton JP, Bland JM, Kentridge RW, Neave NJ 1994. Handedness and longevity: an archival study of cricketers. *British Medical Journal* 309, 1681-1684.

Bailey P, Thorne P, Wynne-Thomas P. 1993. *Who's Who of Cricketers*. 2nd ed, London, Hamlyn.

Bland M and Altman D. 2005. Do the left-handed die young? *Significance* 2, 166-170.

**See Also**

earlycrcktr.

**Examples**

```
data(cricketer)
numLH <- xtabs(~ left+year, data=cricketer)
propLH <- prop.table(numLH, margin=2)[2,]
yr <- as.numeric(colnames(numLH))
plot(propLH ~ yr)
cricketer$lh <- unclass(cricketer$left)-1
left2.hat <- fitted(lm(lh ~ poly(year,2), data=cricketer))
ord <- order(cricketer$year)
lines(left2.hat[ord] ~ cricketer$year[ord])
library(splines)
ns3.hat <- fitted(lm(lh ~ ns(year,3), data=cricketer))
lines(ns3.hat[ord] ~ cricketer$year[ord], col="red")
require(survival)
summary(coxph(Surv(life, kia) ~ bs(year,3) +left, data=cricketer))
cricketer$notacdDead <- with(cricketer, {dead[acd==1]<-0; dead})
summary(coxph(Surv(life, notacdDead) ~ ns(year,2) +left, data=cricketer))
```

---

cuckoohosts

---

*Comparison of cuckoo eggs with host eggs*


---

**Description**

These data compare mean length, mean breadth, and egg color, between cuckoos and their hosts.

**Usage**

cuckoohosts



**Format**

A data frame with 10 observations on the following 12 variables.

**clength** mean length of cuckoo eggs in given host's nest

**cl.sd** standard deviation of cuckoo egg lengths

**cbreadth** mean breadth of cuckoo eggs in given host's nest

**cb.sd** standard deviation of cuckoo egg breadths

**cnum** number of cuckoo eggs

**hlength** length of host eggs

**hl.sd** standard deviation of host egg lengths

**hbreadth** breadth of host eggs

**hb.sd** standard deviation of host egg breadths

**hnum** number of host eggs

**match** number of eggs where color matched

**nomatch** number where color did not match

**Details**

Although from the same study that generated data in the data frame cuckoos, the data do not match precisely. The cuckoo egg lengths and breadths are from the tables on page 168, the host egg lengths and breadths from Appendix IV on page 176, and the color match counts from the table on page 171.

**Source**

Latter, O.H., 1902. The egg of *cuculus canorus*. an inquiry into the dimensions of the cuckoo's egg and the relation of the variations to the size of the eggs of the foster-parent, with notes on coloration, &c. *Biometrika*, 1:164–176.

**Examples**

```
cuckoohosts
str(cuckoohosts)
plot(cuckoohosts)
with(cuckoohosts,
      plot(c(clength,hlength),c(cbreadth,hbreadth),col=rep(1:2,c(6,6))))
```

cuckoos

*Cuckoo Eggs Data***Description**

Length and breadth measurements of 120 eggs lain in the nests of six different species of host bird.

**Usage**

```
cuckoos
```

**Format**

This data frame contains the following columns:

**length** the egg lengths in millimeters

**breadth** the egg breadths in millimeters

**species** a factor with levels hedge.sparrow, meadow.pipit, pied.wagtail, robin, tree.pipit, wren

**id** a numeric vector

**Source**

Latter, O.H. (1902). The eggs of *Cuculus canorus*. An Inquiry into the dimensions of the cuckoo's egg and the relation of the variations to the size of the eggs of the foster-parent, with notes on coloration, &c. *Biometrika* i, 164.

**References**

Tippett, L.H.C. 1931: "The Methods of Statistics". Williams & Norgate, London.

**Examples**

```
print("Strip and Boxplots - Example 2.1.2")

attach(cuckoos)
oldpar <- par(las = 2) # labels at right angle to axis.
stripchart(length ~ species)
boxplot(split(cuckoos$length, cuckoos$species),
        xlab="Length of egg", horizontal=TRUE)
detach(cuckoos)
par(oldpar)
pause()

print("Summaries - Example 2.2.2")
sapply(split(cuckoos$length, cuckoos$species), sd)
pause()
```

```
print("Example 4.1.4")
wren <- split(cuckoos$length, cuckoos$species)$wren
median(wren)
n <- length(wren)
sqrt(pi/2)*sd(wren)/sqrt(n) # this s.e. computation assumes normality
```

---

CVbinary

---

*Cross-Validation for Regression with a Binary Response*


---

## Description

These functions give training (internal) and cross-validation measures of predictive accuracy for regression with a binary response. The data are randomly divided between a number of ‘folds’. Each fold is removed, in turn, while the remaining data are used to re-fit the regression model and to predict at the omitted observations.

## Usage

```
CVbinary(obj, rand=NULL, nfolds=10, print.details=TRUE)
```

```
cv.binary(obj, rand=NULL, nfolds=10, print.details=TRUE)
```

## Arguments

obj	a glm object
rand	a vector which assigns each observation to a fold
nfolds	the number of folds
print.details	logical variable (TRUE = print detailed output, the default)

## Value

cvhat	predicted values from cross-validation
internal	internal or (better) training predicted values
training	training predicted values
acc.cv	cross-validation estimate of accuracy
acc.internal	internal or (better) training estimate of accuracy
acc.training	training estimate of accuracy

## Note

The term ‘training’ seems preferable to the term ‘internal’ in connection with predicted values, and the accuracy measure, that are based on the observations used to derive the model.

**Author(s)**

J.H. Maindonald

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

```
frogs.glm <- glm(pres.abs ~ log(distance) + log(NoOfPools),
                 family=binomial,data=frogs)
CVbinary(frogs.glm)
mifem.glm <- glm(outcome ~ ., family=binomial, data=mifem)
CVbinary(mifem.glm)
```

CVlm

*Cross-Validation for Linear Regression***Description**

This function gives internal and cross-validation measures of predictive accuracy for multiple linear regression. (For binary logistic regression, use the CVbinary function.) The data are randomly assigned to a number of ‘folds’. Each fold is removed, in turn, while the remaining data is used to re-fit the regression model and to predict at the deleted observations.

**Usage**

```
CVlm(df = houseprices, form.lm = formula(sale.price ~ area), m=3, dots =
FALSE, seed=29, plotit = c("Observed","Residual"),
main="Small symbols show cross-validation predicted values",
legend.pos="topleft", printit=TRUE)
cv.lm(df = houseprices, form.lm = formula(sale.price ~ area), m=3, dots =
FALSE, seed=29, plotit = c("Observed","Residual"),
main="Small symbols show cross-validation predicted values",
legend.pos="topleft", printit=TRUE)
```

**Arguments**

df	a data frame
form.lm	a formula or lm call or lm object
m	the number of folds
dots	uses pch=16 for the plotting character
seed	random number generator seed
plotit	This can be one of the text strings "Observed", "Residual", or a logical value. The logical TRUE is equivalent to "Observed", while FALSE is equivalent to "" (no plot)

main	main title for graph
legend.pos	position of legend: one of "bottomright", "bottom", "bottomleft", "left", "topleft", "top", "topright", "right", "center".
printit	if TRUE, output is printed to the screen

### Details

When `plotit="Residual"` and there is more than one explanatory variable, the fitted lines that are shown for the individual folds are approximations.

### Value

ss	the cross-validation residual sum of squares
df	degrees of freedom

### Author(s)

J.H. Maindonald

### See Also

[lm](#), [CVbinary](#)

### Examples

```
CVlm()
## Not run:
CVlm(df=nihills, form.lm=formula(log(time)~log(climb)+log(dist)),
      plotit="Observed")
CVlm(df=nihills, form.lm=formula(log(time)~log(climb)+log(dist)),
      plotit="Residual")

## End(Not run)
```

---

DAAGxdb

*List, each of whose elements hold rows of a file, in character format*

---

### Description

This is the default database for use with the function `datafile`, which uses elements of this list to place files in the working directory.

### Usage

```
data(DAAGxdb)
```

**Format**

Successive elements in this list hold character vectors from which the corresponding files can be generated. The names of the list elements are fuel, fuel.csv, oneBadRow, scan-demo, molclock1, molclock2, and travelbooks.

**Details**

The files fuel.txt and fuel.csv are used in Chapter 1 of DAAGUR, while the files oneBadRow.txt and scan-demo.txt are used in Chapter 14 of DAAGUR.

**References**

Maindonald, J.H. and Braun, W.J. 2007. Data Analysis and Graphics Using R: An Example-Based Approach. 2nd edn, Cambridge University Press (DAAGUR).

**Examples**

```
data(DAAGxdb)
names(DAAGxdb)
```

---

datafile	<i>Write an ASCII data file to the working directory.</i>
----------	---

---

**Description**

Invoking this function writes one or more nominated files to the working directory. In particular, it may be used to write the files 'fuel.txt' and 'fuel.csv' that are used in Chapter 1 of DAAGUR, and the files 'oneBadRow.txt' and 'scan-demo.txt' that are used in Chapter 14 of DAAGUR.

**Usage**

```
datafile(file = c("fuel", "travelbooks"), datastore = DAAGxdb,
altstore = zzDAAGxdb, showNames = FALSE)
```

**Arguments**

file	character; with the defaults for datastore and altstore the options are "fuel", for fuel.txt; "fuel.csv", for fuel.csv; "oneBadRow", for oneBadRow.txt; "scan-demo", for scan-demo.txt; "molclock1", for molclock1.txt; "molclock2", for molclock2.txt; "travelbooks", for travelbooks.txt; "bestTimes", for bestTimes.txt; "bostonc", for bostonc.txt
datastore	Each element of this list is a character vector that holds the rows of a file.
altstore	An alternative list. The default alternative list is used for the two files that are more than a few lines.
showNames	if TRUE, returns the names of available datasets.

**Value**

An ASCII file is output to the current working directory. The names of all available datasets are returned invisibly.

**Author(s)**

J.H. Maindonald

**Examples**

```
datafile(file="", showNames=TRUE)
```

---

dengue	<i>Dengue prevalence, by administrative region</i>
--------	--

---

**Description**

Data record, for each of 2000 administrative regions, whether or not dengue was recorded at any time between 1961 and 1990.

**Usage**

```
data(dengue)
```

**Format**

A data frame with 2000 observations on the following 13 variables.

**humid** Average vapour density: 1961-1990

**humid90** 90th percentile of humid

**temp** Average temperature: 1961-1990

**temp90** 90th percentile of temp

**h10pix** maximum of humid, within a 10 pixel radius

**h10pix90** maximum of humid90, within a 10 pixel radius

**trees** Percent tree cover, from satellite data

**trees90** 90th percentile of trees

**NoYes** Was dengue observed? (1=yes)

**Xmin** minimum longitude

**Xmax** maximum longitude

**Ymin** minimum latitude

**Ymax** maximum latitude

## Details

This is derived from a data set in which the climate and tree cover information were given for each half degree of latitude by half degree of longitude pixel. The variable NoYes was given by administrative region. The climate data and tree cover data given here are 50th or 90th percentiles, where percentiles were calculated across pixels for an administrative region.

## Source

Simon Hales, Environmental Research New Zealand Ltd.

## References

Hales, S., de Wet, N., Maindonald, J. and Woodward, A. 2002. Potential effect of population and climate change global distribution of dengue fever: an empirical model. *The Lancet* 2002; 360: 830-34.

## Examples

```
str(dengue)
glm(NoYes ~ humid, data=dengue, family=binomial)
glm(NoYes ~ humid90, data=dengue, family=binomial)
```

---

dewpoint

*Dewpoint Data*

---

## Description

The dewpoint data frame has 72 rows and 3 columns. Monthly data were obtained for a number of sites (in Australia) and a number of months.

## Usage

```
dewpoint
```

## Format

This data frame contains the following columns:

**maxtemp** monthly minimum temperatures

**mintemp** monthly maximum temperatures

**dewpt** monthly average dewpoint for each combination of minimum and maximum temperature readings (formerly dewpoint)

## Source

Dr Edward Linacre, visiting fellow in the Australian National University Department of Geography.



**Examples**

```

print("Additive Model - Example 7.5")
require(splines)
attach(dewpoint)
ds.lm <- lm(dewpt ~ bs(maxtemp,5) + bs(mintemp,5), data=dewpoint)
ds.fit <- predict(ds.lm, type="terms", se=TRUE)
oldpar <- par(mfrow=c(1,2))
plot(maxtemp, ds.fit$fit[,1], xlab="Maximum temperature",
      ylab="Change from dewpoint mean", type="n")
lines(maxtemp, ds.fit$fit[,1])
lines(maxtemp, ds.fit$fit[,1]-2*ds.fit$se[,1], lty=2)
lines(maxtemp, ds.fit$fit[,1]+2*ds.fit$se[,1], lty=2)
plot(mintemp, ds.fit$fit[,2], xlab="Minimum temperature",
      ylab="Change from dewpoint mean", type="n")
ord<-order(mintemp)
lines(mintemp[ord], ds.fit$fit[ord,2])
lines(mintemp[ord], ds.fit$fit[ord,2]-2*ds.fit$se[ord,2], lty=2)
lines(mintemp[ord], ds.fit$fit[ord,2]+2*ds.fit$se[ord,2], lty=2)
detach(dewpoint)
par(oldpar)

```

droughts

*Periods Between Rain Events***Description**

Data collected at Winnipeg International Airport (Canada) on periods (in days) between rain events.

**Usage**

```
droughts
```

**Format**

This data frame contains the following columns:

**length** the length of time from the completion of the last rain event to the beginning of the next rain event.

**year** the calendar year.

**Examples**

```

boxplot(length ~ year, data=droughts)
boxplot(log(length) ~ year, data=droughts)
hist(droughts$length, main="Winnipeg Droughts", xlab="length (in days)")
hist(log(droughts$length), main="Winnipeg Droughts", xlab="length (in days, log scale)")

```

---

edcCO2*EPICA Dome C Ice Core 800KYr Carbon Dioxide Data*

---

**Description**

Carbon dioxide record from the EPICA (European Project for Ice Coring in Antarctica) Dome C ice core covering 0 to 800 kyr BP.

**Usage**

```
data(edcCO2)
```

**Format**

A data frame with 1096 observations on the following 2 variables.

age Age in years before present (BP)

co2 CO2 level (ppmv)

**Details**

Data are a composite series.

**Source**

[http://www.ncdc.noaa.gov/paleo/icecore/antarctica/domec/domec\\_epica\\_data.html](http://www.ncdc.noaa.gov/paleo/icecore/antarctica/domec/domec_epica_data.html)

**References**

Luthi, D., M. et al. 2008. High-resolution carbon dioxide concentration record 650,000-800,000 years before present. *Nature*, Vol. 453, pp. 379-382, 15 May 2008. doi:10.1038/nature06949

Indermuhle, A., E. et al, 1999, Atmospheric CO2 concentration from 60 to 20 kyr BP from the Taylor Dome ice core, Antarctica. *Geophysical Research Letters*, 27, 735-738.

Monnin, E., A. et al. 2001. Atmospheric CO2 concentrations over the last glacial termination. *Science*, Vol. 291, pp. 112-114.

Petit, J.R. et al. 1999. Climate and atmospheric history of the past 420,000 years from the Vostok ice core, Antarctica. *Nature* 399: 429-436.

Siegenthaler, U. et al. 2005. Stable Carbon Cycle-Climate Relationship During the Late Pleistocene. *Science*, v. 310 , pp. 1313-1317, 25 November 2005.

**Examples**

```
data(edcCO2)
```

---

edcT*EPICA Dome C Ice Core 800KYr Temperature Estimates*

---

**Description**

Temperature record, using Deuterium as a proxy, from the EPICA (European Project for Ice Coring in Antarctica) Dome C ice core covering 0 to 800 kyr BP.

**Usage**

```
data(edcT)
```

**Format**

A data frame with 5788 observations on the following 5 variables.

Bag Bag number

ztop Top depth (m)

Age Years before 1950

Deuterium Deuterium dD data

dT Temperature difference from the average of the last 1000 years ~ -54.5degC

**Details**

Temperature was estimated from the deuterium data, after making various corrections.

**Source**

[http://www.ncdc.noaa.gov/paleo/icecore/antarctica/domec/domec\\_epica\\_data.html](http://www.ncdc.noaa.gov/paleo/icecore/antarctica/domec/domec_epica_data.html)

**References**

Jouzel, J., et al. 2007. EPICA Dome C Ice Core 800KYr Deuterium Data and Temperature Estimates. IGBP PAGES/World Data Center for Paleoclimatology Data Contribution Series \# 2007-091. NOAA/NCDC Paleoclimatology Program, Boulder CO, USA.

Jouzel, J., et al. 2007. Orbital and Millennial Antarctic Climate Variability over the Past 800,000 Years. Science, Vol. 317, No. 5839, pp.793-797, 10 August 2007.

**Examples**

```
data(edcT)
```

---

`elastic1`*Elastic Band Data Replicated*

---

### Description

The `elastic1` data frame has 7 rows and 2 columns giving, for each amount by which an elastic band is stretched over the end of a ruler, the distance that the band traveled when released.

### Usage

```
elastic1
```

### Format

This data frame contains the following columns:

**stretch** the amount by which the elastic band was stretched

**distance** the distance traveled

### Source

J. H. Maindonald

### Examples

```
plot(elastic1)

print("Inline Functions - Example 12.2.2")
sapply(elastic1, mean)
pause()

sapply(elastic1, function(x)mean(x))
pause()

sapply(elastic1, function(x)sum(log(x)))
pause()

print("Data Output - Example 12.3.2")
write.table(elastic1, file="bandsframe.txt")
```

---

elastic2*Elastic Band Data Replicated Again*

---

### Description

The elastic2 data frame has 9 rows and 2 columns giving, for each amount by which an elastic band is stretched over the end of a ruler, the distance that the band traveled when released.

### Usage

```
elastic2
```

### Format

This data frame contains the following columns:

**stretch** the amount by which the elastic band was stretched

**distance** the distance traveled

### Source

J. H. Maindonald

### Examples

```
plot(elastic2)
pause()

print("Chapter 5 Exercise")

yrange <- range(c(elastic1$distance, elastic2$distance))
xrange <- range(c(elastic1$stretch, elastic2$stretch))
plot(distance ~ stretch, data = elastic1, pch = 16, ylim = yrange, xlim =
xrange)
points(distance ~ stretch, data = elastic2, pch = 15, col = 2)
legend(xrange[1], yrange[2], legend = c("Data set 1", "Data set 2"), pch =
c(16, 15), col = c(1, 2))

elastic1.lm <- lm(distance ~ stretch, data = elastic1)
elastic2.lm <- lm(distance ~ stretch, data = elastic2)
abline(elastic1.lm)
abline(elastic2.lm, col = 2)
summary(elastic1.lm)
summary(elastic2.lm)
pause()

predict(elastic1.lm, se.fit=TRUE)
predict(elastic2.lm, se.fit=TRUE)
```

---

 elasticband

*Elastic Band Data*


---

## Description

The elasticband data frame has 7 rows and 2 columns giving, for each amount by which an elastic band is stretched over the end of a ruler, the distance that the band traveled when released.

## Usage

```
elasticband
```

## Format

This data frame contains the following columns:

**stretch** the amount by which the elastic band was stretched

**distance** the distance traveled

## Source

J. H. Maindonald

## Examples

```
print("Example 1.8.1")

attach(elasticband)    # R now knows where to find stretch and distance
plot(stretch, distance) # Alternative: plot(distance ~ stretch)
detach(elasticband)
pause()

print("Output of Data Frames - Example 12.3.2")

write(t(elasticband),file="bands.txt",ncol=2)

sink("bands2.txt")
elasticband  # NB: No output on screen
sink()

print("Lists - Example 12.7")

elastic.lm <- lm(distance ~ stretch, data=elasticband)
names(elastic.lm)
elastic.lm$coefficients
elastic.lm[["coefficients"]]
pause()

elastic.lm[[1]]
pause()
```

```

elastic.lm[1]
pause()

options(digits=3)
elastic.lm$residuals
pause()

elastic.lm$call
pause()

mode(elastic.lm$call)

```

---

errorsINseveral	<i>Simulation of classical errors in x model, with multiple explanatory variables.</i>
-----------------	--

---

### Description

Simulates  $y$ - $x$  and  $x$ - $x$  values for a classical “errors in  $x$ ” linear regression model. One or more  $x$ - $x$  values are subject to random measurement error, independently of the corresponding covariate values that are measured without error.

### Usage

```

errorsINseveral(n = 1000, a0 = 2.5, beta = c(1.5, 0), mu = 12.5, SDyerr = 0.5,
  default.Vpar = list(SDx = 2, rho = -0.5, timesSDx = 1.5),
  V = with(default.Vpar, matrix(c(1, rho, rho, 1), ncol = 2) * SDx^2),
  xerrV = with(default.Vpar, matrix(c(1, 0, 0, 0), ncol = 2) * (SDx * timesSDx)^2),
  parset = NULL, print.summary = TRUE, plotit = TRUE)

```

### Arguments

n	Number of observations
a0	Intercept in linear regression model
beta	Regression coefficients. If one coefficient only is given, this will be repeated as many times as necessary
mu	Vector of covariate means.
SDyerr	SD of $y$ , conditional on the covariates measured without error
default.Vpar	Parameters for the default model with two explanatory variables,
V	Variance-covariance matrix for the $z$ 's, measured without error. (These are generated from a multivariate normal distribution, mainly as a matter of convenience)
xerrV	Variance-covariance matrix for the added “errors in $x$ ”
parset	Parameter list (theme) in a form suitable for supplying to <code>trellis.par.set()</code> .

`print.summary` If TRUE, print summary details of the regression results from the simulation.

`plotit` If TRUE, plot the fitted values for the model with covariates with error, against the fitted values for covariates without error.

### Details

With default arguments, simulates a model in which two covariates are in contention, the first measured without error, and the second with coefficient 0 in the model that includes both covariates measured without error.

### Value

`ERRfree` Data frame holding covariates without error, plus `$y$`

`addedERR` Data frame holding covariates with error, plus `$y$`

### Author(s)

John Maindonald

### References

Data Analysis and Graphics Using R, 3rd edn, Section 6.8.1

### See Also

[errorsINx](#)

### Examples

```
library(lattice)
function(n=1000, a0=2.5, beta=c(1.5,0), mu=12.5, SDyerr=0.5,
        default.Vpar=list(SDx=2, rho=-0.5, timesSDx=1.5),
        V=with(default.Vpar, matrix(c(1,rho,rho,1), ncol=2)*SDx^2),
        xerrV=with(default.Vpar, matrix(c(1,0,0,0), ncol=2)*(SDx*timesSDx)^2),
        parset=NULL, print.summary=TRUE, plotit=TRUE){
  m <- dim(V)[1]
  if(length(mu)==1)mu <- rep(mu,m)
  ow <- options(warn=-1)
  xxmat <- sweep(matrix(rnorm(m*n, 0, 1), ncol=m) %%% chol(V), 2, mu, "+")
  errxx <- matrix(rnorm(m*n, 0, 1), ncol=m) %%% chol(xerrV, pivot=TRUE)
  options(ow)
  dimnames(xxmat)[[2]] <- paste("z", 1:m, sep="")
  xxWITHerr <- xxmat+errxx
  xxWITHerr <- data.frame(xxWITHerr)
  names(xxWITHerr) <- paste("xWITHerr", 1:m, sep="")
  xxWITHerr[, "y"] <- a0 + xxmat %%% matrix(beta,ncol=1) + rnorm(n, sd=SDyerr)
  err.lm <- lm(y ~ ., data=xxWITHerr)
  xx <- data.frame(xxmat)
  names(xx) <- paste("z", 1:m, sep="")
  xx$y <- xxWITHerr$y
  xx.lm <- lm(y ~ ., data=xx)
```



```

B <- coef(err.lm)
b <- coef(xx.lm)
SE <- summary(err.lm)$coef[,2]
se <- summary(xx.lm)$coef[,2]
if(print.summary){
  beta0 <- c(mean(xx$y)-sum(beta*apply(xx[,1:m],2,mean)), beta)
  tab <- rbind(beta0, b, B)
  dimnames(tab) <- list(c("Values for simulation",
                          "Estimates: no error in x1",
                          "LS Estimates: error in x1"),
                        c("Intercept", paste("b", 1:m, sep="")))
  tabSE <- rbind(rep(NA,m+1),se,SE)
  rownames(tabSE) <- rownames(tab)
  colnames(tabSE) <- c("SE(Int)", paste("SE(", colnames(tab)[-1],")", sep=""))
  tab <- cbind(tab,tabSE)
  print(round(tab,3))
}
if(m==2 & print.summary){
  tau <- default.Vpar$timesSDx
  s1 <- sqrt(V[1,1])
  s2 <- sqrt(V[2,2])
  rho <- default.Vpar$rho
  s12 <- s1*sqrt(1-rho^2)
  lambda <- (1-rho^2)/(1-rho^2+tau^2)
  gam12 <- rho*sqrt(V[1,1]/V[2,2])
  expB2 <- beta[2]+beta[1]*(1-lambda)*gam12
  print(c("Theoretical attenuation of b1" = lambda, "Theoretical b2" = expB2))
}
if(is.null(parset))parset <- simpleTheme(col=c("gray40","gray40"),
                                         col.line=c("black","black"))

if(plotit){
  library(lattice)
  zhat <- fitted(xx.lm)
  xhat <- fitted(err.lm)
  plt <- xyplot(xhat ~ zhat, aspect=1, scales=list(tck=0.5),
               panel=function(x,y,...){
                 panel.xyplot(x,y,type="p",...)
                 panel.abline(lm(y ~ x), lty=2)
                 panel.abline(0,1)
               },
               xlab="Fitted values; regress on exact z",
               ylab="Fitted values; regress on x = xWITHerr",
               key=list(space="top", columns=2,
                        text=list(lab=c("Line y=x", "Regression fit to points")),
                        lines=list(lty=1:2)),
               par.settings=parset
               )
  print(plt)}
invisible(list(ERRfree=xx, addedERR=xxWITHerr))
}

```

errorsINx

*Simulate data for straight line regression, with "errors in x".***Description**

Simulates  $y$ - $x$  and  $x$ - $z$  values for the straight line regression model, but with  $x$ - $z$  values subject to random measurement error, following the classical “errors in  $x$ ” model. Optionally, the  $x$ -values can be split into two groups, with one group shifted relative to the other

**Usage**

```
errorsINx(mu = 12.5, n = 200, a = 15, b = 1.5, SDx=2, SDyerr = 1.5,
  timesSDx=(1:5)/2.5, gpfactor=if(missing(gpdiff))FALSE else TRUE,
  gpdiff=if(gpfactor) 1.5 else 0, layout=NULL,
  parset = simpleTheme(alpha = 0.75, col = c("black","gray45"),
    col.line = c("black","gray45"), lwd=c(1,1.5), pch=c(1,2),
    lty=c(1,2)), print.summary=TRUE, plotit=TRUE, xrelation="same")
```

**Arguments**

mu	Mean of $z$
n	Number of points
a	Intercept in model where $z$ is measured without error
b	Slope in model where $z$ is measured without error
SDx	SD of $z$ -values, measured without error
SDyerr	SD of error term in $y$ where $z$ is measured without error
timesSDx	SD of measurement error is timesSDx, as a multiple of SDx
gpfactor	Should $x$ -values be split into two groups, with one shifted relative to the other?
gpdiff	Amount of shift of one group of $z$ -values relative to the other
layout	Layout for lattice graph, if requested
parset	Parameters to be supplied to the lattice plot, if any
print.summary	Print summary information on fits?
plotit	logical: plot the data?
xrelation	character: sets the $x$ -axis relation component of scales to "same" or "free" or (though this does not make sense here) "sliced".

**Details**

The argument timesSDx can be a numeric vector. One set of  $x$ -values that are contaminated with measurement error is simulated for each element of timesSDx.

**Value**

gph	the trellis graphics object
mat	A matrix, with <code>length(timesSDx)+2</code> columns. Values of <code>\$z\$</code> are in the first column. There is one further column (x with error) for each element of <code>timesSDx</code> , followed by a column for <code>\$y\$</code> . If there is a grouping variable, a further column identifies the groups.

**Author(s)**

John Maindonald

**References**

Data Analysis and Graphics Using R, 3rd edn, Section 6.7

**Examples**

```
library(lattice)
errorsINx()
errorsINx(gpdiff=2, timesSDx=1.25, SDyerr=2.5, n=80)
```

---

excessRisk

*Create and analyze multiway frequency or weighted frequency table*

---

**Description**

This function creates a multi-way table of counts for the response given a set of classifying factors. Output facilitates a check on how the factor specified as `margin` may, after accounting for other classifying factors, affect the response.

**Usage**

```
excessRisk(form = weight ~ seatbelt + airbag, response
= "dead", margin = "airbag", data = nassCDS, decpl = 4,
printResults=TRUE)
```

**Arguments**

form	form is a formula in which classifying factors appear on the right, with an optional weight variable on the left.
response	response is a binary variable or two-level factor such that the response of interest is the relative number in the two levels.
margin	margin is the factor whose effect on the response, after accounting for other classifying factors, is of interest
data	data is a data frame in which variables and factors may be found
decpl	decpl is the number of decimal places in proportions that appear in the output
printResults	if TRUE, a tabular summary is printed.

**Details**

The best way to understand what this function does may be to run it with the default parameters, and/or with examples that appear below.

**Value**

The function returns a data frame, with one row for each combination of levels of factors on the right of the formula, but excluding the factor specified as `margin`

Count for level 2 of response \& level 1 of margin

Total count for level 1 of margin

Count for level 2 of response \& level 2 of margin

Total count for level 2 of margin

Proportion; divide count for level 1 of margin by total

Proportion; divide count for level 2 of margin by total

Excess count for level 2 of response in row; relative to the assumption that, in that row, there is no association between response and margin. This is the observed response (for the default arguments, number of dead) for level 2 (airbag deployed), less the number that would have been expected if the proportion had been that for level 1. (Negative values favor airbags.)

**Author(s)**

John Maindonald

**References**

See `help(nassCDS)`

**See Also**

`xtabs`

**Examples**

```
excessRisk()
excessRisk(weight ~ airbag+seatbelt+dvcat)
UCB <- as.data.frame.table(UCBAdmissions)
excessRisk(Freq~Gender, response="Admit", margin="Gender",data=UCB)
excessRisk(Freq~Gender+Dept, response="Admit", margin="Gender",data=UCB)
```

---

fossilfuel*Fossil Fuel Data*

---

**Description**

Estimates of total worldwide carbon emissions from fossil fuel use.

**Usage**

```
fossilfuel
```

**Format**

This data frame contains the following columns:

**year** a numeric vector giving the year the measurement was taken.

**carbon** a numeric vector giving the total worldwide carbon emissions from fossil fuel use, in millions of tonnes.

**Source**

Marland et al (2003)

**Examples**

```
plot(fossilfuel)
```

---

fossum*Female Possum Measurements*

---

**Description**

The fossum data frame consists of nine morphometric measurements on each of 43 female mountain brushtail possums, trapped at seven sites from Southern Victoria to central Queensland. This is a subset of the possum data frame.

**Usage**

```
fossum
```

**Format**

This data frame contains the following columns:

**case** observation number

**site** one of seven locations where possums were trapped

**Pop** a factor which classifies the sites as Vic Victoria, other New South Wales or Queensland

**sex** a factor with levels f female, m male

**age** age

**hdlngh** head length

**skullw** skull width

**totlngh** total length

**taill** tail length

**footlgh** foot length

**earconch** ear conch length

**eye** distance from medial canthus to lateral canthus of right eye

**chest** chest girth (in cm)

**belly** belly girth (in cm)

**Source**

Lindenmayer, D. B., Viggers, K. L., Cunningham, R. B., and Donnelly, C. F. 1995. Morphological variation among columns of the mountain brushtail possum, *Trichosurus caninus* Ogilby (Phalangeridae: Marsupiala). *Australian Journal of Zoology* 43: 449-458.

**Examples**

```
boxplot(fossum$totlngh)
```

---

frogs

*Frogs Data*

---

**Description**

The frogs data frame has 212 rows and 11 columns. The data are on the distribution of the Southern Corroboree frog, which occurs in the Snowy Mountains area of New South Wales, Australia.

**Usage**

```
frogs
```

## Format

This data frame contains the following columns:

**pres.abs** 0 = frogs were absent, 1 = frogs were present

**northing** reference point

**easting** reference point

**altitude** altitude , in meters

**distance** distance in meters to nearest extant population

**NoOfPools** number of potential breeding pools

**NoOfSites** (number of potential breeding sites within a 2 km radius

**avrain** mean rainfall for Spring period

**meanmin** mean minimum Spring temperature

**meanmax** mean maximum Spring temperature

## Source

Hunter, D. (2000) The conservation and demography of the southern corroboree frog (*Pseudophryne corroboree*). M.Sc. thesis, University of Canberra, Canberra.

## Examples

```
print("Multiple Logistic Regression - Example 8.2")

plot(northing ~ easting, data=frogs, pch=c(1,16)[frogs$pres.abs+1],
     xlab="Meters east of reference point", ylab="Meters north")
pairs(frogs[,4:10])
attach(frogs)
pairs(cbind(altitude,log(distance),log(NoOfPools),NoOfSites),
     panel=panel.smooth, labels=c("altitude","log(distance)",
     "log(NoOfPools)","NoOfSites"))
detach(frogs)

frogs.glm0 <- glm(formula = pres.abs ~ altitude + log(distance) +
  log(NoOfPools) + NoOfSites + avrain + meanmin + meanmax,
  family = binomial, data = frogs)
summary(frogs.glm0)

frogs.glm <- glm(formula = pres.abs ~ log(distance) + log(NoOfPools) +
meanmin +
  meanmax, family = binomial, data = frogs)
oldpar <- par(mfrow=c(2,2))
termplot(frogs.glm, data=frogs)

termplot(frogs.glm, data=frogs, partial.resid=TRUE)

cv.binary(frogs.glm0) # All explanatory variables
pause()
```

```

cv.binary(frogs.glm)    # Reduced set of explanatory variables

for (j in 1:4){
  rand <- sample(1:10, 212, replace=TRUE)
  all.acc <- cv.binary(frogs.glm0, rand=rand, print.details=FALSE)$acc.cv
  reduced.acc <- cv.binary(frogs.glm, rand=rand, print.details=FALSE)$acc.cv
  cat("\nAll:", round(all.acc,3), "   Reduced:", round(reduced.acc,3))
}

```

---

frostedflakes

*Frosted Flakes data*


---

### Description

The frosted flakes data frame has 101 rows and 2 columns giving the sugar concentration (in percent) for 25 g samples of a cereal as measured by 2 methods – high performance liquid chromatography (a slow accurate lab method) and a quick method using the infra-analyzer 400.

### Usage

```
elastic1
```

### Format

This data frame contains the following columns:

**Lab** careful laboratory analysis measurements using high performance liquid chromatography

**IA400** measurements based on the infra-analyzer 400

### Source

W. J. Braun

---

fruitohms

*Electrical Resistance of Kiwi Fruit*


---

### Description

Data are from a study that examined how the electrical resistance of a slab of kiwifruit changed with the apparent juice content.

### Usage

```
fruitohms
```



**Format**

This data frame contains the following columns:

**juice** apparent juice content (percent)

**ohms** electrical resistance (in ohms)

**Source**

Harker, F. R. and Maindonald J.H. 1994. Ripening of nectarine fruit. *Plant Physiology* 106: 165 - 171.

**Examples**

```
plot(ohms ~ juice, xlab="Apparent juice content (%)", ylab="Resistance (ohms)", data=fruitohms)
lines(lowess(fruitohms$juice, fruitohms$ohms), lwd=2)
pause()

require(splines)
attach(fruitohms)
plot(ohms ~ juice, cex=0.8, xlab="Apparent juice content (%)",
     ylab="Resistance (ohms)", type="n")
fruit.lmb4 <- lm(ohms ~ bs(juice,4))
ord <- order(juice)
lines(juice[ord], fitted(fruit.lmb4)[ord], lwd=2)
ci <- predict(fruit.lmb4, interval="confidence")
lines(juice[ord], ci[ord,"lwr"])
lines(juice[ord], ci[ord,"upr"])
```

gaba

*Effect of pentazocine on post-operative pain (average VAS scores)***Description**

The table shows, separately for males and females, the effect of pentazocine on post-operative pain profiles (average VAS scores), with (mbac and fbac) and without (mpl and fpl) preoperatively administered baclofen. Pain scores are recorded every 20 minutes, from 10 minutes to 170 minutes.

**Usage**

gaba

**Format**

A data frame with 9 observations on the following 7 variables.

**min** a numeric vector

**mbac** a numeric vector

**mpl** a numeric vector

fbac a numeric vector  
 fpl a numeric vector  
 avbac a numeric vector  
 avplac a numeric vector

## Details

15 females were given baclofen, as against 3 males. 7 females received the placebo, as against 16 males. Averages for the two treatments (baclofen/placebo), taken over all trial participants and ignoring sex, are misleading.

## Source

Gordon, N. C. et al.(1995): ‘Enhancement of Morphine Analgesia by the GABA<sub>B</sub> against Baclofen’. *Neuroscience* 69: 345-349.

## Examples

```
data(gaba)
mr <- range(gaba$min)
tran <- range(gaba[, c("mbac", "mpl", "fbac", "fpl")])
## Means by treatment and sex
par(mfrow=c(1,2))
plot(mr, tran, xlab = "Time post pentazocine (min)",
      ylab = "Reduction in VAS pain rating",
      type = "n", xlim = c(0, 170), ylim = tran)
points(gaba$min, gaba$fbac, pch = 1, col = 8, lwd = 2, lty = 2,
       type = "b")
points(gaba$min, gaba$fpl, pch = 0, col = 8, lwd = 2, lty = 2,
       type = "b")
points(gaba$min, gaba$mbac, pch = 16, col = 8, lty = 2, type = "b")
points(gaba$min, gaba$mpl, pch = 15, col = 8, lty = 2, type = "b")
box()
## Now plot means, by treatment, averaged over all participants
plot(mr, tran, xlab = "Time post pentazocine (min)",
      ylab = "Reduction in VAS pain rating",
      type = "n", xlim = c(0, 170), ylim = tran)
bac <- (15 * gaba$fbac + 3 * gaba$mbac)/18
plac <- (7 * gaba$fpl + 9 * gaba$mpl)/16
points(gaba$min, plac, pch = 15, lty = 1, col=1, type = "b")
points(gaba$min, bac, pch = 16, lty = 1, col=1, type = "b")
box()
par(mfrow=c(1,1))
```

---

geophones

*Seismic Timing Data*


---

**Description**

The geophones data frame has 56 rows and 2 columns. Thickness of a layer of Alberta substratum as measured by a line of geophones.

**Usage**

```
geophones
```

**Format**

This data frame contains the following columns:

**distance** location of geophone.

**thickness** time for signal to pass through substratum.

**Examples**

```
plot(geophones)
lines(lowess(geophones, f=.25))
```

---

greatLakes

*Yearly averages of Great Lake heights: 1918 - 2009*


---

**Description**

Heights, stored as a multivariate time series, are for the lakes Erie, Michigan/Huron, Ontario and St Clair

**Usage**

```
data(greatLakes)
```

**Format**

The format is: mts [1:92, 1:4] 174 174 174 174 174 ... - attr(\*, "dimnames")=List of 2 ..\$ : NULL ..\$ : chr [1:4] "Erie" "michHuron" "Ontario" "StClair" - attr(\*, "tsp")= num [1:3] 1918 2009 1 - attr(\*, "class")= chr [1:2] "mts" "ts"

**Details**

For more details, go to the website that is the source of the data.

**Source**

<http://www.lre.usace.army.mil/greatlakes/hh/greatlakeswaterlevels/historicdata/>

**Examples**

```
data(greatLakes)
plot(greatLakes)
## maybe str(greatLakes)
```

---

grog

---

*Alcohol consumption in Australia and New Zealand*


---

**Description**

Data are annual apparent alcohol consumption in Australia and New Zealand, in liters of pure alcohol content per annum, separately for beer, wine, and spirits (including spirit-based products).

**Usage**

```
data(grog)
```

**Format**

A data frame with 18 observations on the following 5 variables.

Beer liters per annum

Wine liters per annum

Spirit liters per annum

Country a factor with levels Australia NewZealand

Year Year ending in June of the given year

**Details**

Data are total available pure alcohol content, for the three categories, divided by numbers of persons aged 15 years or more. The source data for New Zealand included quarterly figures from December 1997, and annual data to December for all years. The annual New Zealand figure to June 1998 required an estimate for September 1997 that was obtained by extrapolating back the third quarter trend line from later years.

**Source**

Australian data are from <http://www.abs.gov.au>. New Zealand data are derived from data from <http://www.stats.govt.nz/people/health/alcohol.htm>

**Examples**

```
data(grog)
library(lattice)
xyplot(Beer+Wine+Spirit ~ Year | Country, data=grog)
xyplot(Beer+Wine+Spirit ~ Year, groups=Country, data=grog, outer=TRUE)
```

hardcopy

*Graphical Output for Hardcopy***Description**

This function streamlines graphical output to the screen, pdf or ps files. File names for hard copy devices can be generated automatically from function names of the form g3.2 or fig3.2 (the choice of alphabetic characters prior to 3.2 is immaterial).

**Usage**

```
hardcopy(width = 3.75, height = 3.75, color = FALSE, trellis = FALSE,
         device = c("", "pdf", "ps"), path = getwd(), file =
         NULL, format = c("nn-nn", "name"), split = "\\.",
         pointsize = c(8, 4), fonts=NULL, horiz = FALSE, ...)
```

**Arguments**

width	width of plot in inches (sic!)
height	height of plot in inches (sic!)
color	(lattice plots only) TRUE if plot is not black on white only
trellis	TRUE if plot uses trellis graphics
device	screen "", pdf or ps
path	external path name
file	name of file to hold output, else NULL
format	Alternatives are "nn-nn" and "name".
split	character on which to split function name (file=NULL)
pointsize	Pointsizes. For trellis devices a vector of length 2 giving font sizes for text and for points respectively
fonts	For postscript devices, specify families that will be used in addition to the initial device
horiz	FALSE for landscape mode; applies only to postscript files
...	Other arguments for passing to the pdf or postscript

**Details**

If a file name (file, without extension) is not supplied, the `format` argument determines how the name is constructed. With `format="name"`, the function name is used. With `format="nn-nn"` and `dotsplit` unchanged from the default, a function name of the form `g3.1` leads to the name `03-01`. Here `g` can be replaced by any other non-numeric characters; the result is the same. The relevant extension is in any case added.

**Value**

Graphical output to screen, pdf or ps file.

**Author(s)**

J.H. Maindonald

**See Also**

[postscript](#)

---

head.injury

*Minor Head Injury (Simulated) Data*

---

**Description**

The `head.injury` data frame has 3121 rows and 11 columns. The data were simulated according to a simple logistic regression model to match roughly the clinical characteristics of a sample of individuals who suffered minor head injuries.

**Usage**

`head.injury`

**Format**

This data frame contains the following columns:

**age.65** age factor (0 = under 65, 1 = over 65).

**amnesia.before** amnesia before impact (less than 30 minutes = 0, more than 30 minutes = 1).

**basal.skull.fracture** (0 = no fracture, 1 = fracture).

**GCS.decrease** Glasgow Coma Scale decrease (0 = no deterioration, 1 = deterioration).

**GCS.13** initial Glasgow Coma Scale (0 = not '13', 1 = '13').

**GCS.15.2hours** Glasgow Coma Scale after 2 hours (0 = not '15', 1 = '15').

**high.risk** assessed by clinician as high risk for neurological intervention (0 = not high risk, 1 = high risk).

**loss.of.consciousness** (0 = conscious, 1 = loss of consciousness).

**open.skull.fracture** (0 = no fracture, 1 = fracture)

**vomiting** (0 = no vomiting, 1 = vomiting)

**clinically.important.brain.injury** any acute brain finding revealed on CT (0 = not present, 1 = present).

## References

Stiell, I.G., Wells, G.A., Vandemheen, K., Clement, C., Lesiuk, H., Laupacis, A., McKnight, R.D., Verbee, R., Brison, R., Cass, D., Eisenhauer, M., Greenberg, G.H., and Worthington, J. (2001) The Canadian CT Head Rule for Patients with Minor Head Injury, *The Lancet*. 357: 1391-1396.

---

headInjury	<i>Minor Head Injury (Simulated) Data</i>
------------	---

---

## Description

The headInjury data frame has 3121 rows and 11 columns. The data were simulated according to a simple logistic regression model to match roughly the clinical characteristics of a sample of individuals who suffered minor head injuries.

## Usage

```
headInjury
```

## Format

This data frame contains the following columns:

**age.65** age factor (0 = under 65, 1 = over 65).

**amnesia.before** amnesia before impact (less than 30 minutes = 0, more than 30 minutes = 1).

**basal.skull.fracture** (0 = no fracture, 1 = fracture).

**GCS.decrease** Glasgow Coma Scale decrease (0 = no deterioration, 1 = deterioration).

**GCS.13** initial Glasgow Coma Scale (0 = not '13', 1 = '13').

**GCS.15.2hours** Glasgow Coma Scale after 2 hours (0 = not '15', 1 = '15').

**high.risk** assessed by clinician as high risk for neurological intervention (0 = not high risk, 1 = high risk).

**loss.of.consciousness** (0 = conscious, 1 = loss of consciousness).

**open.skull.fracture** (0 = no fracture, 1 = fracture)

**vomiting** (0 = no vomiting, 1 = vomiting)

**clinically.important.brain.injury** any acute brain finding revealed on CT (0 = not present, 1 = present).

## References

Stiell, I.G., Wells, G.A., Vandemheen, K., Clement, C., Lesiuk, H., Laupacis, A., McKnight, R.D., Verbee, R., Brison, R., Cass, D., Eisenhauer, M., Greenberg, G.H., and Worthington, J. (2001) The Canadian CT Head Rule for Patients with Minor Head Injury, *The Lancet*. 357: 1391-1396.

---

hills	<i>Scottish Hill Races Data</i>
-------	---------------------------------

---

### Description

The record times in 1984 for 35 Scottish hill races.

### Usage

hills

### Format

This data frame contains the following columns:

**dist** distance, in miles (on the map)

**climb** total height gained during the route, in feet

**time** record time in hours

### Source

A.C. Atkinson (1986) Comment: Aspects of diagnostic regression analysis. *Statistical Science* 1, 397-402.

Also, in MASS library, with time in minutes.

### References

A.C. Atkinson (1988) Transformations unmasked. *Technometrics* 30, 311-318. [ "corrects" the time for Knock Hill from 78.65 to 18.65. It is unclear if this based on the original records.]

### Examples

```
print("Transformation - Example 6.4.3")
pairs(hills, labels=c("dist\n\n(miles)", "climb\n\n(feet)",
"time\n\n(hours)"))
pause()

pairs(log(hills), labels=c("dist\n\n(log(miles))", "climb\n\n(log(feet))",
"time\n\n(log(hours))"))
pause()

hills0.loglm <- lm(log(time) ~ log(dist) + log(climb), data = hills)
oldpar <- par(mfrow=c(2,2))
plot(hills0.loglm)
pause()

hills.loglm <- lm(log(time) ~ log(dist) + log(climb), data = hills[-18,])
```



```

summary(hills.loglm)
plot(hills.loglm)
pause()

hills2.loglm <- lm(log(time) ~ log(dist)+log(climb)+log(dist):log(climb),
data=hills[-18,])
anova(hills.loglm, hills2.loglm)
pause()

step(hills2.loglm)
pause()

summary(hills.loglm, corr=TRUE)$coef
pause()

summary(hills2.loglm, corr=TRUE)$coef
par(oldpar)
pause()

print("Nonlinear - Example 6.9.4")
hills.nls0 <- nls(time ~ (dist^alpha)*(climb^beta), start =
  c(alpha = .909, beta = .260), data = hills[-18,])
summary(hills.nls0)
plot(residuals(hills.nls0) ~ predict(hills.nls0)) # residual plot
pause()

hills$climb.mi <- hills$climb/5280
hills.nls <- nls(time ~ alpha + beta*dist + gamma*(climb.mi^delta),
  start=c(alpha = 1, beta = 1, gamma = 1, delta = 1), data=hills[-18,])
summary(hills.nls)
plot(residuals(hills.nls) ~ predict(hills.nls)) # residual plot

```

hills2000

*Scottish Hill Races Data - 2000***Description**

The record times in 2000 for 56 Scottish hill races. We believe the data are, for the most part, trustworthy. This is the subset of `racess2000` for which type is hill.

**Usage**

```
hills2000
```

**Format**

This data frame contains the following columns:

**dist** distance, in miles (on the map)

**climb** total height gained during the route, in feet

**time** record time in hours

**timef** record time in hours for females

### Source

The Scottish Running Resource, <http://www.hillrunning.co.uk>

### Examples

```
pairs(hills2000)
```

---

hotspots

*Hawaian island chain hotspot Potassium-Argon ages*

---

### Description

K-Ar Ages (millions of years) and distances (km) from Kilauea along the trend of the chain of Hawaian volcanic islands and other seamounts that are believed to have been created by a moving "hot spot". The age of Kilauea is given as 0-0.4 Ma.

### Usage

```
data(hotspots)
```

### Format

A data frame with 36 observations on the following 6 variables.

ID Volcano identifier

name Name

distance Distance in kilometers

age K-Ar age in millions of years

error Standard error of estimate?

source Data source; see information on web site below.

### Details

For details of the way that errors were calculated, refer to the original papers. See also the comments under `hotspots2006`. In general, errors do not account for geological uncertainty.

### Source

[http://www.soest.hawaii.edu/GG/HCV/haw\\_formation.html](http://www.soest.hawaii.edu/GG/HCV/haw_formation.html)

**Examples**

```
data(hotspots)
plot(age ~ distance, data=hotspots)
abline(lm(age ~ distance, data=hotspots))
```

hotspots2006

*Hawaiian island chain hotspot Argon-Argon ages***Description**

Ar-Ar Ages (millions of years) and distances (km) from Kilauea along the trend of the chain of Hawaiian volcanic islands and other seamounts that are believed to have been created by a moving "hot spot".

**Usage**

```
data(hotspots2006)
```

**Format**

A data frame with 10 observations on the following 6 variables.

```
age      Ar-Ar age
CI95lim  Measurement error; 95% CI
geoErr   Geological Uncertainty
totplus  Total uncertainty (+)
totminus Total uncertainty (-)
distance Distance in kilometers
```

**Details**

Note that measurement error is small relative to geological uncertainty. Geological uncertainty arises because lavas are likely to have erupted, over a period of up to 2 million years, somewhat after passage over the hot spot's centre. Dredging or drilling will in general have accessed lavas from the younger half of this interval. Hence the asymmetry in the geological uncertainty.

**Source**

Warren D. Sharp and David A. Clague, 50-Ma initiation of Hawaiian-Emperor bend records major change in Pacific Plate motion. *Science* 313: 1281-1284 (2006).

**Examples**

```
data(hotspots2006)
```

---

houseprices

*Aranda House Prices*

---

### Description

The houseprices data frame consists of the floor area, price, and the number of bedrooms for a sample of houses sold in Aranda in 1999. Aranda is a suburb of Canberra, Australia.

### Usage

houseprices

### Format

This data frame contains the following columns:

**area** a numeric vector giving the floor area

**bedrooms** a numeric vector giving the number of bedrooms

**sale.price** a numeric vector giving the sale price in thousands of Australian dollars

### Source

J.H. Maindonald

### Examples

```
plot(sale.price~area, data=houseprices)
pause()

coplot(sale.price~area|bedrooms, data=houseprices)
pause()

print("Cross-Validation - Example 5.5.2")

houseprices.lm <- lm(sale.price ~ area, data=houseprices)
summary(houseprices.lm)$sigma^2
pause()

CVlm()
pause()

print("Bootstrapping - Example 5.5.3")
houseprices.fn <- function (houseprices, index){
  house.resample <- houseprices[index,]
  house.lm <- lm(sale.price ~ area, data=house.resample)
  coef(house.lm)[2]    # slope estimate for resampled data
}
require(boot)          # ensure that the boot package is loaded
houseprices.boot <- boot(houseprices, R=999, statistic=houseprices.fn)
```

```

houseprices1.fn <- function (houseprices, index){
  house.resample <- houseprices[index,]
  house.lm <- lm(sale.price ~ area, data=house.resample)
  predict(house.lm, newdata=data.frame(area=1200))
}

houseprices1.boot <- boot(houseprices, R=999, statistic=houseprices1.fn)
boot.ci(houseprices1.boot, type="perc") # "basic" is an alternative to "perc"
houseprices2.fn <- function (houseprices, index){
  house.resample <- houseprices[index,]
  house.lm <- lm(sale.price ~ area, data=house.resample)
  houseprices$sale.price-predict(house.lm, houseprices) # resampled prediction errors
}

n <- length(houseprices$area)
R <- 200
houseprices2.boot <- boot(houseprices, R=R, statistic=houseprices2.fn)
house.fac <- factor(rep(1:n, rep(R, n)))
plot(house.fac, as.vector(houseprices2.boot$t), ylab="Prediction Errors",
     xlab="House")
pause()

plot(apply(houseprices2.boot$t, 2, sd)/predict.lm(houseprices.lm, se.fit=TRUE)$se.fit,
     ylab="Ratio of Bootstrap SE's to Model-Based SE's", xlab="House", pch=16)
abline(1,0)

```

---

humanpower

*Oxygen uptake versus mechanical power, for humans*


---

## Description

The data set from Daedalus project.

## Usage

```
data(humanpower1)
```

## Format

A data frame with 28 observations on the following 3 variables.

**wattsPerKg** a numeric vector: watts per kilogram of body weight

**o2** a numeric vector: ml/min/kg

**id** a factor with levels 1 - 5 (humanpower1) or 1 - 4 (humanpower2), identifying the different athletes

Details

Data in humanpower1 are from investigations (Bussolari 1987) designed to assess the feasibility of a proposed 119 kilometer human powered flight from the island of Crete – in the initial phase of the Daedalus project. Data are for five athletes – a female hockey player, a male amateur tri-athlete, a female amateur triathlete, a male wrestler and a male cyclist – who were selected from volunteers who were recruited through the news media, Data in humanpower2) are for four out of the 25 applicants who were selected for further testing, in the lead-up to the eventual selection of a pilot for the Daedalus project.

Source

Bussolari, S.R.(1987). Human factors of long-distance human-powered aircraft flights. Human Power 5: 8-12.

Nadel and Bussolari, S.R.(1988). The Daedalus project: physiological problems and solutions. American Scientist 76: 351-360.

References

Nadel and Bussolari, S.R.(1989). The physiological limits of long-duration human-power production – lessons learned from the Daedalus project. Human Power 7: 7-10.

Examples

```
str(humanpower1)
plot(humanpower1)
lm(o2 ~ id + wattsPerKg:id, data=humanpower1)
lm(o2 ~ id + wattsPerKg:id, data=humanpower2)
```

---

intersalt	<i>Blood pressure versus Salt; inter-population data</i>
-----------	--

---

Description

Median blood pressure, as a function of salt intake, for each of 52 human populations.

Usage

```
intersalt
```

Format

- A data frame with 52 observations on the following 4 variables.
- b a numeric vector
  - bp mean diastolic blood pressure (mm Hg)
  - na mean sodium excretion (mmol/24h)
  - country a character vector

## Details

For each population took a sample of 25 males and 25 females from each decade in the age range 20 - 50, i.e. 200 individuals in all.

## Source

Intersalt Cooperative Research Group. 1988. Intersalt: an international study of electrolyte excretion and blood pressure: results for 24 hour urinary sodium and potassium excretion. *British Medical Journal* 297: 319-328.

## References

Maindonald, J.H. *The Design of Research Studies ? A Statistical Perspective*, viii + 109pp. Graduate School Occasional Paper 00/2, Australian National University 2000.

## Examples

```
data(intersalt)
plot(bp ~ na, data=intersalt, xlab="Median sodium excretion (mmol/24h)",
      ylab="Median diastolic blood pressure (mm Hg)")
```

---

ironslag

*Iron Content Measurements*

---

## Description

The ironslag data frame has 53 rows and 2 columns. Two methods for measuring the iron content in samples of slag were compared, a chemical and a magnetic method. The chemical method requires greater effort than the magnetic method.

## Usage

```
ironslag
```

## Format

This data frame contains the following columns:

**chemical** a numeric vector containing the measurements coming from the chemical method

**magnetic** a numeric vector containing the measurements coming from the magnetic method

## Source

Hand, D.J., Daly, F., McConway, K., Lunn, D., and Ostrowski, E. eds (1993) *A Handbook of Small Data Sets*. London: Chapman & Hall.

### Examples

```
iron.lm <- lm(chemical ~ magnetic, data = ironslag)
oldpar <- par(mfrow = c(2,2))
plot(iron.lm)
par(oldpar)
```

---

jobs

*Canadian Labour Force Summary Data (1995-96)*

---

### Description

The number of workers in the Canadian labour force broken down by region (BC, Alberta, Prairies, Ontario, Quebec, Atlantic) for the 24-month period from January, 1995 to December, 1996 (a time when Canada was emerging from a deep economic recession).

### Usage

jobs

### Format

This data frame contains the following columns:

**BC** monthly labour force counts in British Columbia

**Alberta** monthly labour force counts in Alberta

**Prairies** monthly labour force counts in Saskatchewan and Manitoba

**Ontario** monthly labour force counts in Ontario

**Quebec** monthly labour force counts in Quebec

**Atlantic** monthly labour force counts in Newfoundland, Nova Scotia, Prince Edward Island and New Brunswick

**Date** year (in decimal form)

### Details

These data have been seasonally adjusted.

### Source

Statistics Canada



## Examples

```
print("Multiple Variables and Times - Example 2.1.4")
sapply(jobs, range)
pause()

matplot(jobs[,7], jobs[,-7], type="l", xlim=c(95,97.1))
# Notice that we have been able to use a data frame as the second argument to matplot().
# For more information on matplot(), type help(matplot)
text(rep(jobs[24,7], 6), jobs[24,1:6], names(jobs)[1:6], adj=0)
pause()

sapply(log(jobs[,-7]), range)
apply(sapply(log(jobs[,-7]), range), 2, diff)
pause()

oldpar <- par(mfrow=c(2,3))
range.log <- sapply(log(jobs[,-7], 2), range)
maxdiff <- max(apply(range.log, 2, diff))
range.log[2,] <- range.log[1,] + maxdiff
titles <- c("BC Jobs", "Alberta Jobs", "Prairie Jobs",
           "Ontario Jobs", "Quebec Jobs", "Atlantic Jobs")
for (i in 1:6){
  plot(jobs$Date, log(jobs[,i], 2), type = "l", ylim = range.log[,i],
       xlab = "Time", ylab = "Number of jobs", main = titles[i])
}
par(oldpar)
```

---

kiwishade

*Kiwi Shading Data*


---

## Description

The kiwishade data frame has 48 rows and 4 columns. The data are from a designed experiment that compared different kiwifruit shading treatments. There are four vines in each plot, and four plots (one for each of four treatments: none, Aug2Dec, Dec2Feb, and Feb2May) in each of three blocks (locations: west, north, east). Each plot has the same number of vines, each block has the same number of plots, with each treatment occurring the same number of times.

## Usage

```
kiwishade
```

## Format

This data frame contains the following columns:

**yield** Total yield (in kg)

**plot** a factor with levels east.Aug2Dec, east.Dec2Feb, east.Feb2May, east.none, north.Aug2Dec, north.Dec2Feb, north.Feb2May, north.none, west.Aug2Dec, west.Dec2Feb, west.Feb2May, west.none

**block** a factor indicating the location of the plot with levels east, north, west

**shade** a factor representing the period for which the experimenter placed shading over the vines; with levels: none no shading, Aug2Dec August - December, Dec2Feb December - February, Feb2May February - May

## Details

The northernmost plots were grouped together because they were similarly affected by shading from the sun in the north. For the remaining two blocks shelter effects, whether from the west or from the east, were thought more important.

## Source

Snelgar, W.P., Manson, P.J., Martin, P.J. 1992. Influence of time of shading on flowering and yield of kiwifruit vines. *Journal of Horticultural Science* 67: 481-487.

## References

Maindonald J H 1992. Statistical design, analysis and presentation issues. *New Zealand Journal of Agricultural Research* 35: 121-141.

## Examples

```
print("Data Summary - Example 2.2.1")
attach(kiwishade)
kiwimeans <- aggregate(yield, by=list(block, shade), mean)
names(kiwimeans) <- c("block", "shade", "meanyield")

kiwimeans[1:4,]
pause()

print("Multilevel Design - Example 9.3")
kiwishade.aov <- aov(yield ~ shade+Error(block/shade), data=kiwishade)
summary(kiwishade.aov)
pause()

sapply(split(yield, shade), mean)

pause()

kiwi.table <- t(sapply(split(yield, plot), as.vector))
kiwi.means <- sapply(split(yield, plot), mean)
kiwi.means.table <- matrix(rep(kiwi.means, 4), nrow=12, ncol=4)
kiwi.summary <- data.frame(kiwi.means, kiwi.table-kiwi.means.table)
names(kiwi.summary) <- c("Mean", "Vine 1", "Vine 2", "Vine 3", "Vine 4")
kiwi.summary
mean(kiwi.means) # the grand mean (only for balanced design)
```

```

if(require(lme4, quietly=TRUE)) {
  kiwishade.lmer <- lmer(yield ~ shade + (1|block) + (1|block:plot),
                        data=kiwishade)
  ## block:shade is an alternative to block:plot

  kiwishade.lmer

  ##                      Residuals and estimated effects
  xyplot(residuals(kiwishade.lmer) ~ fitted(kiwishade.lmer)|block,
        data=kiwishade, groups=shade,
        layout=c(3,1), par.strip.text=list(cex=1.0),
        xlab="Fitted values (Treatment + block + plot effects)",
        ylab="Residuals", pch=1:4, grid=TRUE,
        scales=list(x=list(alternating=FALSE), tck=0.5),
        key=list(space="top", points=list(pch=1:4),
                text=list(labels=levels(kiwishade$shade)),columns=4))
  ploteff <- ranef(kiwishade.lmer, drop=TRUE)[[1]]
  qqmath(ploteff, xlab="Normal quantiles", ylab="Plot effect estimates",
        scales=list(tck=0.5))
}

```

---

leafshape

*Full Leaf Shape Data Set*


---

## Description

Leaf length, width and petiole measurements taken at various sites in Australia.

## Usage

```
leafshape
```

## Format

This data frame contains the following columns:

**bladelen** leaf length (in mm)

**petiole** a numeric vector

**bladewid** leaf width (in mm)

**latitude** latitude

**logwid** natural logarithm of width

**logpet** logarithm of petiole

**loglen** logarithm of length

**arch** leaf architecture (0 = plagiotropic, 1 = orthotropic)

**location** a factor with levels Sabah, Panama, Costa Rica, N Queensland, S Queensland, Tasmania

**Source**

King, D.A. and Maindonald, J.H. 1999. Tree architecture in relation to leaf dimensions and tree stature in temperate and tropical rain forests. *Journal of Ecology* 87: 1012-1024.

---

leafshape17

---

*Subset of Leaf Shape Data Set*


---

**Description**

The leafshape17 data frame has 61 rows and 8 columns. These are leaf length, width and petiole measurements taken at several sites in Australia. This is a subset of the leafshape data frame.

**Usage**

```
leafshape17
```

**Format**

This data frame contains the following columns:

**bladelen** leaf length (in mm)

**petiole** a numeric vector

**bladewid** leaf width (in mm)

**latitude** latitude

**logwid** natural logarithm of width

**logpet** logarithm of petiole measurement

**loglen** logarithm of length

**arch** leaf architecture (0 = orthotropic, 1 = plagiotropic)

**Source**

King, D.A. and Maindonald, J.H. 1999. Tree architecture in relation to leaf dimensions and tree stature in temperate and tropical rain forests. *Journal of Ecology* 87: 1012-1024.

**Examples**

```
print("Discriminant Analysis - Example 11.2")

require(MASS)
leaf17.lda <- lda(arch ~ logwid+loglen, data=leafshape17)
leaf17.hat <- predict(leaf17.lda)
leaf17.lda
  table(leafshape17$arch, leaf17.hat$class)
pause()

tab <- table(leafshape17$arch, leaf17.hat$class)
```

```

    sum(tab[row(tab)==col(tab)])/sum(tab)
leaf17cv.lda <- lda(arch ~ logwid+loglen, data=leafshape17, CV=TRUE)
tab <- table(leafshape17$arch, leaf17cv.lda$class)
pause()

leaf17.glm <- glm(arch ~ logwid + loglen, family=binomial, data=leafshape17)
options(digits=3)
summary(leaf17.glm)$coef
pause()

leaf17.one <- cv.binary(leaf17.glm)
table(leafshape17$arch, round(leaf17.one$internal))      # Resubstitution
pause()

table(leafshape17$arch, round(leaf17.one$cv))            # Cross-validation

```

---

leaftemp	<i>Leaf and Air Temperature Data</i>
----------	--------------------------------------

---

## Description

These data consist of measurements of vapour pressure and of the difference between leaf and air temperature.

## Usage

```
leaftemp
```

## Format

This data frame contains the following columns:

**CO2level** Carbon Dioxide level low, medium, high

**vapPress** Vapour pressure

**tempDiff** Difference between leaf and air temperature

**BtempDiff** a numeric vector

## Source

Katharina Siebke and Susan von Cammerer, Australian National University.

## Examples

```

print("Fitting Multiple Lines - Example 7.3")

leaf.lm1 <- lm(tempDiff ~ 1, data = leaftemp)
leaf.lm2 <- lm(tempDiff ~ vapPress, data = leaftemp)
leaf.lm3 <- lm(tempDiff ~ CO2level + vapPress, data = leaftemp)
leaf.lm4 <- lm(tempDiff ~ CO2level + vapPress + vapPress:CO2level,

```

```

data = leaftemp)

anova(leaf.lm1, leaf.lm2, leaf.lm3, leaf.lm4)

summary(leaf.lm2)
plot(leaf.lm2)

```

---

leaftemp.all

*Full Leaf and Air Temperature Data Set*


---

### Description

The leaftemp.all data frame has 62 rows and 9 columns.

### Usage

```
leaftemp.all
```

### Format

This data frame contains the following columns:

**glasshouse** a factor with levels A, B, C

**CO2level** a factor with Carbon Dioxide Levels: high, low, medium

**day** a factor

**light** a numeric vector

**CO2** a numeric vector

**tempDiff** Difference between Leaf and Air Temperature

**BtempDiff** a numeric vector

**airTemp** Air Temperature

**vapPress** Vapour Pressure

### Source

J.H. Maindonald

litters

*Mouse Litters***Description**

Data on the body and brain weights of 20 mice, together with the size of the litter. Two mice were taken from each litter size.

**Usage**

```
litters
```

**Format**

This data frame contains the following columns:

**lsize** litter size

**bodywt** body weight

**brainwt** brain weight

**Source**

Wainright P, Pelkman C and Wahlsten D 1989. The quantitative relationship between nutritional effects on preweaning growth and behavioral development in mice. *Developmental Psychobiology* 22: 183-193.

**Examples**

```
print("Multiple Regression - Example 6.2")

pairs(litters, labels=c("lsize\n\n(litter size)", "bodywt\n\n(Body Weight)",
                       "brainwt\n\n(Brain Weight)"))
# pairs(litters) gives a scatterplot matrix with less adequate labeling

mice1.lm <- lm(brainwt ~ lsize, data = litters) # Regress on lsize
mice2.lm <- lm(brainwt ~ bodywt, data = litters) #Regress on bodywt
mice12.lm <- lm(brainwt ~ lsize + bodywt, data = litters) # Regress on lsize & bodywt

summary(mice1.lm)$coef # Similarly for other coefficients.
# results are consistent with the biological concept of brain sparing

pause()

hat(model.matrix(mice12.lm)) # hat diagonal
pause()

plot(lm.influence(mice12.lm)$hat, residuals(mice12.lm))

print("Diagnostics - Example 6.3")
```

```

mice12.lm <- lm(brainwt ~ bodywt+lsize, data=litters)
oldpar <-par(mfrow = c(1,2))
bx <- mice12.lm$coef[2]; bz <- mice12.lm$coef[3]
res <- residuals(mice12.lm)
plot(litters$bodywt, bx*litters$bodywt+res, xlab="Body weight",
     ylab="Component + Residual")
panel.smooth(litters$bodywt, bx*litters$bodywt+res) # Overlay
plot(litters$lsize, bz*litters$lsize+res, xlab="Litter size",
     ylab="Component + Residual")
panel.smooth(litters$lsize, bz*litters$lsize+res)
par(oldpar)

```

---

lmdiags

Return data required for diagnostic plots

---

## Description

This extracts the code that provides the major part of the statistical information used by `plot.lm`, leaving out the code used to provide the graphs

## Usage

```
lmdiags(x, which = c(1L:3L, 5L), cook.levels = c(0.5, 1), hii=NULL)
```

## Arguments

<code>x</code>	This must be an object of class <code>lm</code> object, or that inherits from an object of class <code>lm</code> .
<code>which</code>	a subset of the numbers '1:6', indicating the plots for which statistical information is required
<code>cook.levels</code>	Levels for contours of <code>cook.levels</code> , by default <code>c(0.5, 1)</code>
<code>hii</code>	Diagonal elements for the hat matrix. If not supplied ( <code>hii=NULL</code> ), they will be calculated from the argument <code>x</code> .

## Details

See [plot.lm](#) for additional information.

## Value

<code>yh</code>	fitted values
<code>rs</code>	standardized residuals (for <code>glm</code> models standardized deviance residuals)
<code>yhn0</code>	As <code>yh</code> , but omitting observations with zero weight
<code>cook</code>	Cook's statistics
<code>rsp</code>	standardized residuals (for <code>glm</code> models standardized Pearson residuals)



**Note**

This function is designed, in the first place, for use in connection with [plotSimDiags](#), used to give simulations of diagnostic plots for `lm` objects.

**Author(s)**

John Maindonald, using code that John Maindonald, Martin Maechler and others had contributed to [plot.lm](#)

**References**

See references for [plot.lm](#)

**See Also**

[plotSimDiags](#), [plot.lm](#)

**Examples**

```
women.lm <- lm(weight ~ height, data=women)
veclist <- lmdiags(x=women.lm)
## Returns the statistics that are required for graphs 1, 2, 3, and 5
```

---

logisticsim

*Simple Logistic Regression Data Simulator*

---

**Description**

This function simulates simple regression data from a logistic model.

**Usage**

```
logisticsim(x = seq(0, 1, length=101), a = 2, b = -4, seed=NULL)
```

**Arguments**

x	a numeric vector representing the explanatory variable
a	the regression function intercept
b	the regression function slope
seed	numeric constant

**Value**

a list consisting of

x	the explanatory variable vector
y	the Poisson response vector

**Examples**

```
logisticsim()
```

---

Lottario	<i>Ontario Lottery Data</i>
----------	-----------------------------

---

**Description**

The data frame Lottario is a summary of 122 weekly draws of an Ontario lottery, beginning in November, 1978. Each draw consists of 7 numbered balls, drawn without replacement from an urn consisting of balls numbered from 1 through 39.

**Usage**

```
Lottario
```

**Format**

This data frame contains the following columns:

**Number** the integers from 1 to 39, representing the numbered balls

**Frequency** the number of occurrences of each numbered ball

**Source**

The Ontario Lottery Corporation

**References**

Bellhouse, D.R. (1982). Fair is fair: new rules for Canadian lotteries. Canadian Public Policy - Analyse de Politiques 8: 311-320.

**Examples**

```
order(Lottario$Frequency)[33:39] # the 7 most frequently chosen numbers
```

---

lung	<i>Cape Fur Seal Lung Measurements</i>
------	--

---

**Description**

The lung vector consists of weight measurements of lungs taken from 30 Cape Fur Seals that died as an unintended consequence of commercial fishing.

**Usage**

```
lung
```

---

Manitoba.lakes

*The Nine Largest Lakes in Manitoba*


---

**Description**

The `Manitoba.lakes` data frame has 9 rows and 2 columns. The areas and elevations of the nine largest lakes in Manitoba, Canada. The geography of Manitoba (a relatively flat province) can be divided crudely into three main areas: a very flat prairie in the south which is at a relatively high elevation, a middle region consisting of mainly of forest and Precambrian rock, and a northern region which drains more rapidly into Hudson Bay. All water in Manitoba, which does not evaporate, eventually drains into Hudson Bay.

**Usage**

```
Manitoba.lakes
```

**Format**

This data frame contains the following columns:

**elevation** a numeric vector consisting of the elevations of the lakes (in meters)

**area** a numeric vector consisting of the areas of the lakes (in square kilometers)

**Source**

The CANSIM data base at Statistics Canada.

**Examples**

```
plot(Manitoba.lakes)
plot(Manitoba.lakes[-1,])
```

---

measles

*Deaths in London from measles*


---

**Description**

Deaths in London from measles: 1629 – 1939, with gaps.

**Usage**

```
data(measles)
```

**Format**

The format is: Time-Series [1:311] from 1629 to 1939: 42 2 3 80 21 33 27 12 NA NA ...

**Source**

Guy, W. A. 1882. Two hundred and fifty years of small pox in London. Journal of the Royal Statistical Society 399-443.

Stocks, P. 1942. Measles and whooping cough during the dispersal of 1939-1940. Journal of the Royal Statistical Society 105:259-291.

**References**

Lancaster, H. O. 1990. Expectations of Life. Springer.

---

medExpenses	<i>Family Medical Expenses</i>
-------------	--------------------------------

---

**Description**

The medExpenses data frame contains average weekly medical expenses including drugs for 33 families randomly sampled from a community of 600 families which contained 2700 individuals. These data were collected in the 1970's at an unknown location.

**Usage**

```
medExpenses
```

**Format**

**familysize** number of individuals in a family

**expenses** average weekly cost for medical expenses per family member

**Examples**

```
with(medExpenses, weighted.mean(expenses, familysize))
```

---

mifem	<i>Mortality Outcomes for Females Suffering Myocardial Infarction</i>
-------	---

---

**Description**

The mifem data frame has 1295 rows and 10 columns. This is the female subset of the 'monica' data frame

**Usage**

```
mifem
```

**Format**

This data frame contains the following columns:

**outcome** mortality outcome, a factor with levels live, dead  
**age** age at onset  
**yronset** year of onset  
**premi** previous myocardial infarction event, a factor with levels y, n, nk not known  
**smstat** smoking status, a factor with levels c current, x ex-smoker, n non-smoker, nk not known  
**diabetes** a factor with levels y, n, nk not known  
**highbp** high blood pressure, a factor with levels y, n, nk not known  
**hichol** high cholesterol, a factor with levels y, n nk not known  
**angina** a factor with levels y, n, nk not known  
**stroke** a factor with levels y, n, nk not known

**Source**

Newcastle (Australia) centre of the Monica project; see the web site <http://www.ktl.fi/monica>

**Examples**

```
print("CART - Example 10.7")
summary(mifem)
pause()

require(rpart)
mifem.rpart <- rpart(outcome ~ ., data = mifem, cp = 0.0025)
plotcp(mifem.rpart)
printcp(mifem.rpart)
pause()

mifemb.rpart <- prune(mifem.rpart, cp=0.006)
print(mifemb.rpart)
```

---

mignonette

---

*Darwin's Wild Mignonette Data*


---

**Description**

Data which compare the heights of crossed plants with self-fertilized plants. Plants were paired within the pots in which they were grown, with one on one side and one on the other.

**Usage**

```
mignonette
```

**Format**

This data frame contains the following columns:

**cross** heights of the crossed plants

**self** heights of the self-fertilized plants

**Source**

Darwin, Charles. 1877. The Effects of Cross and Self Fertilisation in the Vegetable Kingdom. Appleton and Company, New York.

**Examples**

```
print("Is Pairing Helpful? - Example 4.3.1")

attach(mignonette)
plot(cross ~ self, pch=rep(c(4,1), c(3,12))); abline(0,1)
abline(mean(cross-self), 1, lty=2)
detach(mignonette)
```

---

milk

*Milk Sweetness Study*


---

**Description**

The milk data frame has 17 rows and 2 columns. Each of 17 panelists compared two milk samples for sweetness.

**Usage**

```
milk
```

**Format**

This data frame contains the following columns:

**four** a numeric vector consisting of the assessments for four units of additive

**one** a numeric vector while the is the assessment for one unit of additive

**Source**

J.H. Maindonald

**Examples**

```
print("Rug Plot - Example 1.8.1")
xyrange <- range(milk)
plot(four ~ one, data = milk, xlim = xyrange, ylim = xyrange, pch = 16)
rug(milk$one)
rug(milk$four, side = 2)
abline(0, 1)
```

---

modelcars

---

*Model Car Data*


---

**Description**

The modelcars data frame has 12 rows and 2 columns. The data are for an experiment in which a model car was released three times at each of four different distances up a 20 degree ramp. The experimenter recorded distances traveled from the bottom of the ramp across a concrete floor.

**Usage**

```
modelcars
```

**Format**

This data frame contains the following columns:

**distance.traveled** a numeric vector consisting of the lengths traveled (in cm)

**starting.point** a numeric vector consisting of the distance of the starting point from the top of the ramp (in cm)

**Source**

W.J. Braun

**Examples**

```
plot(modelcars)
modelcars.lm <- lm(distance.traveled ~ starting.point, data=modelcars)
aov(modelcars.lm)
pause()

print("Response Curves - Example 4.6")
attach(modelcars)
stripchart(distance.traveled ~ starting.point, vertical=TRUE, pch=15,
           xlab = "Distance up ramp", ylab="Distance traveled")
detach(modelcars)
```

---

`monica`*WHO Monica Data*

---

## Description

The monica data frame has 6357 rows and 12 columns. Note that `mi fem` is the female subset of this data frame.

## Usage

```
monica
```

## Format

This data frame contains the following columns:

**outcome** mortality outcome, a factor with levels `live`, `dead`

**age** age at onset

**sex** `m` = male, `f` = female

**hosp** `y` = hospitalized, `n` = not hospitalized

**yronset** year of onset

**premi** previous myocardial infarction event, a factor with levels `y`, `n`, `nk` not known

**smstat** smoking status, a factor with levels `c` current, `x` ex-smoker, `n` non-smoker, `nk` not known

**diabetes** a factor with levels `y`, `n`, `nk` not known

**highbp** high blood pressure, a factor with levels `y`, `n`, `nk` not known

**hichol** high cholesterol, a factor with levels `y`, `n`, `nk` not known

**angina** a factor with levels `y`, `n`, `nk` not known

**stroke** a factor with levels `y`, `n`, `nk` not known

## Source

Newcastle (Australia) centre of the Monica project; see the web site <http://www.ktl.fi/monica>

## Examples

```
print("CART - Example 10.7")
summary(monica)
pause()

require(rpart)
monica.rpart <- rpart(outcome ~ ., data = monica, cp = 0.0025)
plotcp(monica.rpart)
printcp(monica.rpart)
pause()

monicab.rpart <- prune(monica.rpart, cp=0.006)
print(monicab.rpart)
```



moths

*Moths Data***Description**

The moths data frame has 41 rows and 4 columns. These data are from a study of the effect of habitat on the densities of two species of moth (A and P). Transects were set across the search area. Within transects, sections were identified according to habitat type.

**Usage**

moths

**Format**

This data frame contains the following columns:

**meters** length of transect

**A** number of type A moths found

**P** number of type P moths found

**habitat** a factor with levels Bank, Disturbed, Lowerside, NEsoak, NWsoak, SEsoak, SWsoak, Upperside

**Source**

Sharyn Wragg, formerly of Australian National University

**Examples**

```
print("Quasi Poisson Regression - Example 8.3")
rbind(table(moths[,4]), sapply(split(moths[, -4], moths$habitat), apply, 2,
sum))
A.glm <- glm(formula = A ~ log(meters) + factor(habitat), family =
quasipoisson, data = moths)
summary(A.glm)
moths$habitat <- relevel(moths$habitat, ref="Lowerside")
A.glm <- glm(A ~ habitat + log(meters), family=quasipoisson, data=moths)
summary(A.glm)$coef
```

---

multilap	<i>Data Filtering Function</i>
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---

**Description**

A subset of data is selected for which the treatment to control ratio of non-binary covariates is never outside a specified range.

**Usage**

```
multilap(df=nsw74psid1, maxf=20, colnames=c("educ", "age", "re74", "re75",
"re78"))
```

**Arguments**

df	a data frame
maxf	filtering parameter
colnames	columns to be compared for filtering

**Author(s)**

J.H. Maindonald

---

nassCDS	<i>Airbag and other influences on accident fatalities</i>
---------	---

---

**Description**

US data, for 1997-2002, from police-reported car crashes in which there is a harmful event (people or property), and from which at least one vehicle was towed. Data are restricted to front-seat occupants, include only a subset of the variables recorded, and are restricted in other ways also.

**Usage**

```
nassCDS
```

**Format**

A data frame with 26217 observations on the following 15 variables.

dvcat ordered factor with levels (estimated impact speeds) 1-9km/h, 10-24, 25-39, 40-54, 55+

weight Observation weights, albeit of uncertain accuracy, designed to account for varying sampling probabilities.

dead factor with levels alive dead

airbag a factor with levels none airbag

seatbelt a factor with levels none belted  
 frontal a numeric vector; 0 = non-frontal, 1=frontal impact  
 sex a factor with levels f m  
 ageOfOcc age of occupant in years  
 yearacc year of accident  
 yearVeh Year of model of vehicle; a numeric vector  
 abcat Did one or more (driver or passenger) airbag(s) deploy? This factor has levels deploy  
 nodeploy unavail  
 occRole a factor with levels driver pass  
 deploy a numeric vector: 0 if an airbag was unavailable or did not deploy; 1 if one or more bags  
 deployed.  
 injSeverity a numeric vector; 0:none, 1:possible injury, 2:no incapacity, 3:incapacity, 4:killed;  
 5:unknown, 6:prior death  
 caseid character, created by pasting together the populations sampling unit, the case number, and  
 the vehicle number. Within each year, use this to uniquely identify the vehicle.

## Details

Data collection used a multi-stage probabilistic sampling scheme. The observation weight, called national inflation factor (national) in the data from NASS, is the inverse of an estimate of the selection probability. These data include a subset of the variables from the NASS dataset. Variables that are coded here as factors are coded as numeric values in that dataset.

## Source

<http://www.stat.uga.edu/~mmeyer/airbags.htm>

<ftp://ftp.nhtsa.dot.gov/nass/>

See also <http://www.maths.anu.edu.au/~johnm/datasets/airbags>

## References

- Meyer, M.C. and Finney, T. (2005): *Who wants airbags?*. Chance 18:3-16.
- Farmer, C.H. 2006. *Another look at Meyer and Finney's 'Who wants airbags?'*. Chance 19:15-22.
- Meyer, M.C. 2006. *Commentary on "Another look at Meyer and Finney's 'Who wants airbags?'"*. Chance 19:23-24.
- For analyses based on the alternative FARS (Fatal Accident Recording System) data, and associated commentary, see:
- Cummings, P; McKnight, B, 2010. *Accounting for vehicle, crash, and occupant characteristics in traffic crash studies*. Injury Prevention 16: 363-366. [The relatively definitive analyses in this paper use a matched cohort design,
- Olson, CM; Cummings, P, Rivara, FP, 2006. *Association of first- and second-generation air bags with front occupant death in car crashes: a matched cohort study*. Am J Epidemiol 164:161-169. [The relatively definitive analyses in this paper use a matched cohort design, using data taken from the FARS (Fatal Accident Recording System) database.]

Braver, ER; Shardell, M; Teoh, ER, 2010. *How have changes in air bag designs affected frontal crash mortality?* Ann Epidemiol 20:499-510.

The web page <http://www-fars.nhtsa.dot.gov/Main/index.aspx> has a menu-based interface into the FARS (Fatality Analysis Recording System) data. The FARS database aims to include every accident in which there was at least one fatality.

## Examples

```
data(nassCDS)
xtabs(weight ~ dead + airbag, data=nassCDS)
xtabs(weight ~ dead + airbag + seatbelt + dvcat, data=nassCDS)
tab <- xtabs(weight ~ dead + abcat, data=nassCDS,
             subset=dvcat=="25-39"&frontal==0)[, c(3,1,2)]
round(tab[2, ]/apply(tab,2,sum)*100,2)
```

---

nasshead

---

*Documentation of names of columns in nass9702cor*


---

## Description

SASname and longname are from the SAS XPT file nass9702cor.XPT that is available from the website noted below. The name shortname is the name used in the data frame nass9702cor, not included in this package, but available from my website that is noted below. It is also used in nassCDS, for columns that nassCDS includes.

## Usage

```
data(nasshead)
```

## Format

A data frame with 56 observations on the following 3 variables.

shortname a character vector

SASname a character vector

longname a character vector

## Details

For full details of the coding of values in columns of nass9702cor, consult one of the SAS format files that can be obtained by following the instructions on Dr Meyer's web site that is noted below.

## Source

<http://www.stat.uga.edu/~mmeyer/airbags.htm> \ <ftp://ftp.nhtsa.dot.gov/nass/> \ Click, e.g., on 1997 and then on SASformats. See also <http://www.maths.anu.edu.au/~johnm/datasets/airbags>

## References

- Meyer, M.C. and Finney, T. (2005): *Who wants airbags?*. Chance 18:3-16.
- Farmer, C.H. 2006. *Another look at Meyer and Finney's 'Who wants airbags?'*. Chance 19:15-22.
- Meyer, M.C. 2006. *Commentary on "Another look at Meyer and Finney's 'Who wants airbags?'"*. Chance 19:23-24.

## Examples

```
data(nasshead)
```

---

nihills

*Record times for Northern Ireland mountain running events*

---

## Description

Data are from the 2007 calendar for the Northern Ireland Mountain Running Association.

## Usage

```
data(nihills)
```

## Format

A data frame with 23 observations on the following 4 variables.

dist distances in miles

climb amount of climb in feet

time record time in hours for males

timef record time in hours for females

## Details

These data make an interesting comparison with the dataset hills2000 in the DAAG package.

## Source

<http://www.nimra.org.uk/calendar.asp>

## Examples

```
data(nihills)
lm(formula = log(time) ~ log(dist) + log(climb), data = nihills)
lm(formula = log(time) ~ log(dist) + log(climb/dist), data = nihills)
```

---

nsw74demo*Labour Training Evaluation Data*

---

**Description**

This data frame contains 445 rows and 10 columns. These data are from an investigation of the effect of training on changes, between 1974-1975 and 1978, in the earnings of individuals who had experienced employment difficulties. Data are for the male experimental control and treatment groups.

**Usage**

nsw74demo

**Format**

This data frame contains the following columns:

**trt** a numeric vector identifying the study in which the subjects were enrolled (0 = PSID, 1 = NSW).

**age** age (in years).

**educ** years of education.

**black** (0 = not black, 1 = black).

**hisp** (0 = not hispanic, 1 = hispanic).

**marr** (0 = not married, 1 = married).

**nodeg** (0 = completed high school, 1 = dropout).

**re74** real earnings in 1974.

**re75** real earnings in 1975.

**re78** real earnings in 1978.

**Source**

<http://www.columbia.edu/~rd247/nswdata.html>

**References**

Dehejia, R.H. and Wahba, S. 1999. Causal effects in non-experimental studies: re-evaluating the evaluation of training programs. *Journal of the American Statistical Association* 94: 1053-1062.

Lalonde, R. 1986. Evaluating the economic evaluations of training programs. *American Economic Review* 76: 604-620.

---

nsw74psid1*Labour Training Evaluation Data*

---

**Description**

This data frame contains 2675 rows and 10 columns. These data are pertinent to an investigation of the way that earnings changed, between 1974-1975 and 1978, in the absence of training. Data for the experimental treatment group (NSW) were combined with control data results from the Panel Study of Income Dynamics (PSID) study.

**Usage**

nsw74psid1

**Format**

This data frame contains the following columns:

**trt** a numeric vector identifying the study in which the subjects were enrolled (0 = PSID, 1 = NSW).

**age** age (in years).

**educ** years of education.

**black** (0 = not black, 1 = black).

**hisp** (0 = not hispanic, 1 = hispanic).

**marr** (0 = not married, 1 = married).

**nodeg** (0 = completed high school, 1 = dropout).

**re74** real earnings in 1974.

**re75** real earnings in 1975.

**re78** real earnings in 1978.

**Source**

<http://www.columbia.edu/~rd247/nswdata.html>

**References**

Dehejia, R.H. and Wahba, S. 1999. Causal effects in non-experimental studies: re-evaluating the evaluation of training programs. *Journal of the American Statistical Association* 94: 1053-1062.

Lalonde, R. 1986. Evaluating the economic evaluations of training programs. *American Economic Review* 76: 604-620.

## Examples

```
print("Interpretation of Regression Coefficients - Example 6.6")

nsw74psid1.lm <- lm(re78~ trt+ (age + educ + re74 + re75) +
  (black + hisp + marr + nodeg), data = nsw74psid1)
summary(nsw74psid1.lm)$coef
options(digits=4)
sapply(nsw74psid1[, c(2,3,8,9,10)], quantile, prob=c(.25,.5,.75,.95,1))
attach(nsw74psid1)
sapply(nsw74psid1[trt==1, c(2,3,8,9,10)], quantile,
  prob=c(.25,.5,.75,.95,1))
pause()

here <- age <= 40 & re74<=5000 & re75 <= 5000 & re78 < 30000
nsw74psidA <- nsw74psid1[here, ]
detach(nsw74psid1)
table(nsw74psidA$trt)
pause()

A1.lm <- lm(re78 ~ trt + (age + educ + re74 + re75) + (black +
  hisp + marr + nodeg), data = nsw74psidA)
summary(A1.lm)$coef
pause()

A2.lm <- lm(re78 ~ trt + (age + educ + re74 + re75) * (black +
  hisp + marr + nodeg), data = nsw74psidA)
anova(A1.lm, A2.lm)
```

---

nsw74psid3

*Labour Training Evaluation Data*


---

## Description

These data are pertinent to an investigation of the way that earnings changed, between 1974-1975 and 1978, in the absence of training. The data frame combines data for the experimental treatment group (NSW, 185 observations), using as control data results from the PSID (Panel Study of Income Dynamics) study (128 observations). The latter were chosen to mimic the characteristics of the NSW training and control groups. These are a subset of the nsw74psid1 data.

## Usage

```
nsw74psid3
```

## Format

This data frame contains the following columns:

**trt** a numeric vector identifying the study in which the subjects were enrolled (0 = PSID, 1 = NSW)



**age** age (in years)  
**educ** years of education  
**black** (0 = not black, 1 = black)  
**hisp** (0 = not hispanic, 1 = hispanic)  
**marr** (0 = not married, 1 = married)  
**nodeg** (0 = completed high school, 1 = dropout)  
**re74** real earnings in 1974  
**re75** real earnings in 1975  
**re78** real earnings in 1978

### Source

<http://www.columbia.edu/~rd247/nswdata.html>

### References

Dehejia, R.H. and Wahba, S. 1999. Causal effects in non-experimental studies: re-evaluating the evaluation of training programs. *Journal of the American Statistical Association* 94: 1053-1062.

Lalonde, R. 1986. Evaluating the economic evaluations of training programs. *American Economic Review* 76: 604-620.

### Examples

```
print("Contingency Tables - Example 4.4")
table(nsw74psid3$trt, nsw74psid3$nodeg)
chisq.test(table(nsw74psid3$trt, nsw74psid3$nodeg))
```

---

nsw74psidA

*A Subset of the nsw74psid1 Data Set*


---

### Description

The nsw74psidA data frame has 252 rows and 10 columns. See nsw74psid1 for more information.

### Usage

```
nsw74psidA
```

### Format

This data frame contains the following columns:

**trt** a numeric vector  
**age** a numeric vector  
**educ** a numeric vector  
**black** a numeric vector  
**hisp** a numeric vector  
**marr** a numeric vector  
**nodeg** a numeric vector  
**re74** a numeric vector  
**re75** a numeric vector  
**re78** a numeric vector

### Details

This data set was obtained using:

```
here <- age <= 40 & re74<=5000 & re75 <= 5000 & re78 < 30000
nsw74psidA <- nsw74psid1[here, ]
```

### Examples

```
table(nsw74psidA$trt)
A1.lm <- lm(re78 ~ trt + (age + educ + re74 + re75) + (black +
  hisp + marr + nodeg), data = nsw74psidA)
summary(A1.lm)$coef
discA.glm <- glm(formula = trt ~ age + educ + black + hisp +
  marr + nodeg + re74 + re75, family = binomial, data = nsw74psidA)
A.scores <- predict(discA.glm)
options(digits=4)
overlap <- A.scores > -3.5 & A.scores < 3.8
A.lm <- lm(re78 ~ trt + A.scores, data=nsw74psidA, subset = overlap)
summary(A.lm)$coef
```

---

nswdemo

*Labour Training Evaluation Data*


---

### Description

The nswdemo data frame contains 722 rows and 10 columns. These data are pertinent to an investigation of the way that earnings changed, between 1974-1975 and 1978, for an experimental treatment who were given job training as compared with a control group who did not receive such training.

The psid1 data set is an alternative non-experimental "control" group. psid2 and psid3 are subsets of psid1, designed to be better matched to the experimental data than psid1. Note also the cps1, cps2 and cps3 datasets (**DAAGxtras**) that have been proposed as non-experimental controls.

**Usage**

```
data(nswdemo)
```

**Format**

This data frame contains the following columns:

**trt** a numeric vector identifying the study in which the subjects were enrolled (0 = Control, 1 = treated).

**age** age (in years).

**educ** years of education.

**black** (0 = not black, 1 = black).

**hisp** (0 = not hispanic, 1 = hispanic).

**marr** (0 = not married, 1 = married).

**nodeg** (0 = completed high school, 1 = dropout).

**re74** real earnings in 1974.

**re75** real earnings in 1975.

**re78** real earnings in 1978.

**Source**

<http://www.nber.org/~rdehejia/nswdata.html>

**References**

Dehejia, R.H. and Wahba, S. 1999. Causal effects in non-experimental studies: re-evaluating the evaluation of training programs. *Journal of the American Statistical Association* 94: 1053-1062.

Lalonde, R. 1986. Evaluating the economic evaluations of training programs. *American Economic Review* 76: 604-620.

Smith, J. A. and Todd, P.E. 2005, "Does Matching overcome. LaLonde's critique of nonexperimental estimators", *Journal of Econometrics* 125: 305-353.

Dehejia, R.H. 2005. Practical propensity score matching: a reply to Smith and Todd. *Journal of Econometrics* 125: 355-364.

**See Also**

[psid1](#), [psid2](#), [psid3](#)

nswpsid1

*Labour Training Evaluation Data***Description**

This data frame contains 2787 rows and 10 columns. These data are pertinent to an investigation of the way that earnings changed, between 1974-1975 and 1978, in the absence of training. Data for the experimental treatment group in nswdemo are combined with the psid1 control data from the Panel Study of Income Dynamics (PSID) study.

**Usage**

psid1

**Format**

This data frame contains the following columns:

**trt** a numeric vector identifying the study in which the subjects were enrolled (0 = Control, 1 = treated).

**age** age (in years).

**educ** years of education.

**black** (0 = not black, 1 = black).

**hisp** (0 = not hispanic, 1 = hispanic).

**marr** (0 = not married, 1 = married).

**nodeg** (0 = completed high school, 1 = dropout).

**re74** real earnings in 1974.

**re75** real earnings in 1975.

**re78** real earnings in 1978.

**Details**

The cps1 and psid1 data sets are two non-experimental "control" groups, alternative to that in nswdemo, used in investigating whether use of such a non-experimental control group can be satisfactory. cps2 and cps3 are subsets of cps1, designed to be better matched to the experimental data than cps1. Similar psid2 and psid3 are subsets of psid1, designed to be better matched to the experimental data than psid1. nswpsid1 combines data for the experimental treatment group in nswdemo with the psid1 control data from the Panel Study of Income Dynamics (PSID) study.

**Source**

<http://www.nber.org/~rdehejia/nswdata.html>

## References

- Dehejia, R.H. and Wahba, S. 1999. Causal effects in non-experimental studies: re-evaluating the evaluation of training programs. *Journal of the American Statistical Association* 94: 1053-1062.
- Lalonde, R. 1986. Evaluating the economic evaluations of training programs. *American Economic Review* 76: 604-620.
- Smith, J. A. and Todd, P.E. "Does Matching overcome. LaLonde?s critique of nonexperimental estimators", *Journal of Econometrics* 125: 305-353.
- Dehejia, R.H. 2005. Practical propensity score matching: a reply to Smith and Todd. *Journal of Econometrics* 125: 355-364.

---

obounce

---

*Bounce - obsolete*


---

## Description

A utility function for oneway.plot

## Author(s)

J.H. Maindonald

---

oddbooks

---

*Measurements on 12 books*


---

## Description

Data giving thickness (mm), height (cm), width (cm) and weight (g), of 12 books. Books were selected so that thickness decreased as page area increased

## Usage

data(oddbooks)

## Format

A data frame with 12 observations on the following 4 variables.

**thick** a numeric vector

**height** a numeric vector

**breadth** a numeric vector

**weight** a numeric vector

## Source

JM took books from his library.

**Examples**

```
data(oddbooks)
str(oddbooks)
plot(oddbooks)
```

---

onesamp

*Paired Sample t-test*


---

**Description**

This function performs a t-test for the mean difference for paired data, and produces a scatterplot of one column against the other column, showing whether there was any benefit to using the paired design.

**Usage**

```
onesamp(dset, x="unsprayed", y="sprayed", xlab=NULL, ylab=NULL,
        dubious=NULL, conv=NULL, dig=2)
```

**Arguments**

dset	a matrix or dataframe having two columns
x	name of column to play the role of the ‘predictor’
y	name of column to play the role of the ‘response’
xlab	horizontal axis label
ylab	vertical axis label
dubious	vector of logical (FALSE/TRUE) values, specifying points that are to be omitted
conv	scaling factor that should be applied to data
dig	round SE to this number of digits for display on graph

**Value**

A scatterplot of y against x together with estimates of standard errors and standard errors of the difference (y-x).

Also produced is a confidence interval and p-value for the test.

**Author(s)**

J.H. Maindonald

**Examples**

```
onesamp(dset = pair65, x = "ambient", y = "heated", xlab =
        "Amount of stretch (ambient)", ylab =
        "Amount of stretch (heated)")
```

---

onet.permutation	<i>One Sample Permutation t-test</i>
------------------	--------------------------------------

---

**Description**

This function computes the p-value for the one sample t-test using a permutation test. The permutation density can also be plotted.

**Usage**

```
onet.permutation(x=pair65$heated - pair65$ambient, nsim=2000, plotit=TRUE)
```

**Arguments**

x	a numeric vector containing the sample values (centered at the null hypothesis value)
nsim	the number of permutations (randomly selected)
plotit	if TRUE, the permutation density is plotted

**Value**

The p-value for the test of the hypothesis that the mean of x differs from 0

**Author(s)**

J.H. Maindonald

**References**

Good, P. 2000. Permutation Tests. Springer, New York.

**Examples**

```
onet.permutation()
```

---

onetPermutation	<i>One Sample Permutation t-test</i>
-----------------	--------------------------------------

---

**Description**

This function computes the p-value for the one sample t-test using a permutation test. The permutation density can also be plotted.

**Usage**

```
onetPermutation(x=pair65$heated - pair65$ambient, nsim=2000, plotit=TRUE)
```

**Arguments**

x	a numeric vector containing the sample values (centered at the null hypothesis value)
nsim	the number of permutations (randomly selected)
plotit	if TRUE, the permutation density is plotted

**Value**

The p-value for the test of the hypothesis that the mean of x differs from 0

**Author(s)**

J.H. Maindonald

**References**

Good, P. 2000. Permutation Tests. Springer, New York.

**Examples**

```
onetPermutation()
```

---

oneway.plot	<i>Display of One Way Analysis Results</i>
-------------	--

---

**Description**

A line plot of means for unstructured comparison.

**Usage**

```
oneway.plot(obj, axisht = 6, xlim = NULL, xlab = NULL,  
  lsdht = 1.5, hsdht = 0.5, textht = axisht - 2.5, oma = rep(1,  
    4), angle = 80, alpha = 0.05)
```

**Arguments**

obj	One way analysis of variance object (from aov)
axisht	Axis height
xlim	Range on horizontal axis
xlab	Horizontal axis label
lsdht	Height adjustment parameter for LSD comparison plot
hsdht	Height adjustment parameter for Tukey's HSD comparison plot
textht	Height of text
oma	Outer margin area
angle	Text angle (in degrees)
alpha	Test size



**Value**

A line plot

**Author(s)**

J.H. Maindonald

**Examples**

```
rice.aov <- aov(ShootDryMass ~ trt, data=rice)
oneway.plot(obj=rice.aov)
```

---

onewayPlot

*Display of One Way Analysis Results*


---

**Description**

A line plot of estimates for unstructured comparison of factor levels

**Usage**

```
onewayPlot(obj, trtnam = "trt", axisht = 6, xlim = NULL,
  xlab = NULL, lsdht = 1.5, hsdht = 0.5, textht = axisht -
  2.5, oma = rep(1, 4), angle = 80, alpha = 0.05)
```

**Arguments**

obj	One way analysis of variance object (from aov)
trtnam	name of factor for which line plot is required
axisht	Axis height
xlim	Range on horizontal axis
xlab	Horizontal axis label
lsdht	Height adjustment parameter for display of LSD
hsdht	Height adjustment parameter for display of Tukey's HSD
textht	Height of text
oma	Outer margin area
angle	Text angle (in degrees)
alpha	Test size

**Value**

Estimates, labeled with level names, are set out along a line

**Author(s)**

J.H. Maindonald

**Examples**

```
rice.aov <- aov(ShootDryMass ~ trt, data=rice)
onewayPlot(obj=rice.aov)
```

---

orings

---

*Challenger O-rings Data*


---

**Description**

Record of the number and type of O-ring failures prior to the tragic Challenger mission in January, 1986.

**Usage**

```
orings
```

**Format**

This data frame contains the following columns:

**Temperature** O-ring temperature for each test firing or actual launch of the shuttle rocket engine

**Erosion** Number of erosion incidents

**Blowby** Number of blowby incidents

**Total** Total number of incidents

**Source**

Presidential Commission on the Space Shuttle Challenger Accident, Vol. 1, 1986: 129-131.

**References**

Tufte, E. R. 1997. Visual Explanations. Graphics Press, Cheshire, Connecticut, U.S.A.

**Examples**

```
oldpar <- par(mfrow=c(1,2))
plot(Total~Temperature, data = orings[c(1,2,4,11,13,18),]) # the
# observations included in the pre-launch charts
plot(Total~Temperature, data = orings)
par(oldpar)
```

---

overlap.density	<i>Overlapping Density Plots - obsolete</i>
-----------------	---

---

**Description**

Densities for two independent samples are estimated and plotted.

**Usage**

```
overlap.density(x0, x1, ratio=c(0.05, 20), compare.numbers=TRUE,  
plotit=TRUE, gpnames=c("Control", "Treatment"), xlab="Score")
```

**Arguments**

x0	control group measurements
x1	treatment group measurements
ratio	the range within which the relative numbers of observations from the two groups are required to lie. [The relative numbers at any point are estimated from $(\text{density1} * n1) / (\text{density0} * x0)$ ]
compare.numbers	If TRUE (default), then density plots are scaled to have total area equal to the sample size; otherwise total area under each density is 1
plotit	If TRUE, a plot is produced
gpnames	Names of the two samples
xlab	Label for x-axis

**Author(s)**

J.H. Maindonald

**See Also**

t.test

**Examples**

```
attach(two65)  
overlap.density(ambient,heated)  
t.test(ambient,heated)
```

---

overlapDensity	<i>Overlapping Density Plots</i>
----------------	----------------------------------

---

**Description**

Densities for two independent samples are estimated and plotted.

**Usage**

```
overlapDensity(x0, x1, ratio = c(0.05, 20), compare.numbers = FALSE,
               plotit = TRUE, gpnames = c("Control", "Treatment"),
               cutoffs=c(lower=TRUE, upper=TRUE), bw=FALSE,
               xlab = "Score", col=1:2, lty=1:2)
```

**Arguments**

x0	control group measurements
x1	treatment group measurements
ratio	the range within which the relative numbers of observations from the two groups are required to lie. [The relative numbers at any point are estimated from $(\text{density1} * n1) / (\text{density0} * x0)$ ]
compare.numbers	If TRUE (default), then density plots are scaled to have total area equal to the sample size; otherwise total area under each density is 1
plotit	If TRUE, a plot is produced
gpnames	Names of the two samples
cutoffs	logical vector, indicating whether density estimates should be truncated below (lower=TRUE) or above (upper=TRUE)
bw	logical, indicates whether to overwrite with a gray scale plot
xlab	Label for x-axis
col	standard color parameter
lty	standard line type preference

**Author(s)**

J.H. Maindonald

**See Also**

t.test

**Examples**

```
attach(two65)
overlapDensity(ambient,heated)
t.test(ambient,heated)
```

---

ozone

---

*Ozone Data***Description**

Monthly provisional mean total ozone (in Dobson units) at Halley Bay (approximately corrected to Bass-Paur).

**Usage**

ozone

**Format**

This data frame contains the following columns:

**Year** the year

**Aug** August mean total ozone

**Sep** September mean total ozone

**Oct** October mean total ozone

**Nov** November mean total ozone

**Dec** December mean total ozone

**Jan** January mean total ozone

**Feb** February mean total ozone

**Mar** March mean total ozone

**Apr** April mean total ozone

**Annual** Yearly mean total ozone

**Source**

Shanklin, J. (2001) Ozone at Halley, Rothera and Vernadsky/Faraday.

<http://www.antarctica.ac.uk/met/jds/ozone/data/zoz5699.dat>

**References**

Christie, M. (2000) The Ozone Layer: a Philosophy of Science Perspective. Cambridge University Press.

**Examples**

```
AnnualOzone <- ts(ozone$Annual, start=1956)
plot(AnnualOzone)
```

pair65

*Heated Elastic Bands***Description**

The pair65 data frame has 9 rows and 2 columns. Eighteen elastic bands were divided into nine pairs, with bands of similar stretchiness placed in the same pair. One member of each pair was placed in hot water (60-65 degrees C) for four minutes, while the other was left at ambient temperature. After a wait of about ten minutes, the amounts of stretch, under a 1.35 kg weight, were recorded.

**Usage**

pair65

**Format**

This data frame contains the following columns:

**heated** a numeric vector giving the stretch lengths for the heated bands

**ambient** a numeric vector giving the stretch lengths for the unheated bands

**Source**

J.H. Maindonald

**Examples**

```
mean(pair65$heated - pair65$ambient)
sd(pair65$heated - pair65$ambient)
```

panel.corr

*Scatterplot Panel***Description**

This function produces a bivariate scatterplot with the Pearson correlation. This is for use with the function panelplot.

**Usage**

```
panel.corr(data, ...)
```

**Arguments**

**data** A data frame with columns x and y

**...** Additional arguments

**Author(s)**

J.H. Maindonald

**Examples**

```
# correlation between body and brain weights for 20 mice:

weights <- litters[,-1]
names(weights) <- c("x","y")
weights <- list(weights)
weights[[1]]$xlim <- range(litters[,2])
weights[[1]]$ylim <- range(litters[,3])
panelplot(weights, panel.corr, totrows=1, totcols=1)
```

panelCorr

*Scatterplot Panel***Description**

This function produces a bivariate scatterplot with the Pearson correlation. This is for use with the function panelplot.

**Usage**

```
panelCorr(data, ...)
```

**Arguments**

data	A data frame with columns x and y
...	Additional arguments

**Author(s)**

J.H. Maindonald

**Examples**

```
# correlation between body and brain weights for 20 mice:

weights <- litters[,-1]
names(weights) <- c("x","y")
weights <- list(weights)
weights[[1]]$xlim <- range(litters[,2])
weights[[1]]$ylim <- range(litters[,3])
panelplot(weights, panelCorr, totrows=1, totcols=1)
```

panelplot

*Panel Plot***Description**

Panel plots of various types.

**Usage**

```
panelplot(data, panel=points, totrows=3, totcols=2, oma=rep(2.5, 4), par.strip.text=NULL)
```

**Arguments**

data	A list consisting of elements, each of which consists of x, y, xlim and ylim vectors
panel	The panel function to be plotted
totrows	The number of rows in the plot layout
totcols	The number of columns in the plot layout
oma	Outer margin area
par.strip.text	A data frame with column cex

**Author(s)**

J.H. Maindonald

**Examples**

```
x1 <- x2 <- x3 <- (11:30)/5
y1 <- x1 + rnorm(20)/2
y2 <- 2 - 0.05 * x1 + 0.1 * ((x1 - 1.75))^4 + 1.25 * rnorm(20)
r <- round(cor(x1, y2), 3)
rho <- round(cor(rank(x1), rank(y2)), 3)
y3 <- (x1 - 3.85)^2 + 0.015 * rnorm(20)/4
theta <- ((2 * pi) * (1:20))/20
x4 <- 10 + 4 * cos(theta)
y4 <- 10 + 4 * sin(theta) + (0.5 * rnorm(20))
r1 <- cor(x1, y1)
xy <- data.frame(x = c(rep(x1, 3), x4), y = c(y1, y2, y3, y4),
                gp = rep(1:4, rep(20, 4)))
xy <- split(xy, xy$gp)
xlimdf <- lapply(list(x1, x2, x3, x4), range)
ylimdf <- lapply(list(y1, y2, y3, y4), range)
xy <- lapply(1:4, function(i, u, v, w){list(xlim=v[[i]], ylim=w[[i]],
                                           x=u[[i]]$x, y=u[[i]]$y)},
            u=xy, v=xlimdf, w=ylimdf)

panel.corr <- function (data, ...)
```



```

{
  x <- data$x
  y <- data$y
  points(x, y, pch = 16)
  chh <- par()$cxy[2]
  x1 <- min(x)
  y1 <- max(y) - chh/4
  r1 <- cor(x, y)
  text(x1, y1, paste(round(r1, 3)), cex = 0.8, adj = 0)
}

panelplot(xy, panel=panel.corr, totrows=2, totcols=2, oma=rep(1,4))

```

---

pause

*Pause before continuing execution*

---

### Description

If a program produces several plots, insertion of `pause()` between two plots suspends execution until the <Enter> key is pressed, to allow inspection of the current plot.

### Usage

`pause()`

### Author(s)

From the 'sm' package of Bowman and Azzalini (1997)

---

plotSampDist

*Plot(s) of simulated sampling distributions*

---

### Description

Plots are based on the output from `simulateSampDist()`. By default, both density plots and normal probability plots are given, for a sample from the specified population and for samples of the relevant size(s)

### Usage

```

plotSampDist(sampvalues, graph = c("density", "qq"), cex = 0.925,
  titletext = "Empirical sampling distributions of the",
  popsampl=TRUE, ...)

```

**Arguments**

sampvalues	Object output from simulateSampDist()
graph	Either or both of "density" and "qq"
cex	Character size parameter, relative to default
titletext	Title for graph
popsample	If TRUE show distribution of random sample from population
...	Other graphics parameters

**Value**

Plots graph(s), as described above.

**Author(s)**

John Maindonald

**References**

Maindonald, J.H. and Braun, W.J. (3rd edn, 2010) "Data Analysis and Graphics Using R", Sections 3.3 and 3.4.

**See Also**

See Also `help(simulateSampDist)`

**Examples**

```
## By default, sample from normal population
simAvs <- simulateSampDist()
par(pty="s")
plotSampDist(simAvs)
## Sample from empirical distribution
simAvs <- simulateSampDist(rpop=rivers)
plotSampDist(simAvs)

## The function is currently defined as
function(sampvalues, graph=c("density", "qq"), cex=0.925,
         titletext="Empirical sampling distributions of the",
         popsample=TRUE, ...){
  if(length(graph)==2)oldpar <- par(mfrow=c(1,2), mar=c(3.1,4.1,1.6,0.6),
                                   mgp=c(2.5, 0.75, 0), oma=c(0,0,1.5,0), cex=cex)
  values <- sampvalues$values
  numINsamp <- sampvalues$numINsamp
  funtxt <- sampvalues$FUN
  nDists <- length(numINsamp)+1
  nfirst <- 2
  legitems <- paste("Size", numINsamp)
  if(popsample){nfirst <- 1
    legitems <- c("Size 1", legitems)
  }
```

```

    }
    if(match("density", graph)){
      popdens <- density(values[,1], ...)
      avdens <- vector("list", length=nDists)
      maxht <- max(popdens$y)
      ## For each sample size specified in numINSamp, calculate mean
      ## (or other statistic specified by FUN) for numsamp samples
      for(j in nfirst:nDists){
        av <- values[, j]
        avdens[[j]] <- density(av, ...)
        maxht <- max(maxht, avdens[[j]]$y)
      }
    }
    if(length(graph)>0)
      for(graphtype in graph){
        if(graphtype=="density"){
          if(popsample)
            plot(popdens, ylim=c(0, 1.2*maxht), type="l", yaxs="i",
                  main="")
          else plot(avdens[[2]], type="n", ylim=c(0, 1.2*maxht),
                    yaxs="i", main="")
          for(j in 2:nDists)lines(avdens[[j]], col=j)
          legend("topleft",
                 legend=legitems,
                 col=nfirst:nDists, lty=rep(1,nDists-nfirst+1), cex=cex)
        }
        if(graphtype=="qq"){
          if(popsample) qqnorm(values[,1], main="")
          else qqnorm(values[,2], type="n")
          for(j in 2:nDists){
            qqav <- qqnorm(values[, j], plot.it=FALSE)
            points(qqav, col=j, pch=j)
          }
          legend("topleft", legend=legitems,
                 col=nfirst:nDists, pch=nfirst:nDists, cex=cex)
        }
      }
    }
    if(par()$oma[3]>0){
      outer <- TRUE
      line=0
    } else
    {
      outer <- FALSE
      line <- 1.25
    }
    if(!is.null(titletext))
      mtext(side=3, line=line,
            paste(titletext, funtxt),
            cex=1.1, outer=outer)
    if(length(graph)>1)par(oldpar)
  }
}

```

plotSimDiags

*Diagnostic plots for simulated data***Description**

This provides diagnostic plots, closely equivalent to those provided by `plot.lm`, for simulated data. By default, simulated data are for the fitted model. Alternatively, simulated data can be supplied, making it possible to check the effect of fitting, e.g., an AR1 model.

**Usage**

```
plotSimDiags(obj, simvalues = NULL, seed = NULL,
types = NULL, which = c(1:3, 5), layout = c(4, 1), qqline=TRUE,
cook.levels = c(0.5, 1), caption = list("Residuals vs Fitted",
"Normal Q-Q", "Scale-Location", "Cook's distance", "Residuals vs Leverage",
expression("Cook's dist vs Leverage " * h[ii]/(1 - h[ii]))),
...)
```

**Arguments**

<code>obj</code>	Fitted model object - <code>lm</code> or an object inheriting from <code>lm</code>
<code>simvalues</code>	Optional matrix of simulated data.
<code>seed</code>	Random number seed - set this to make results repeatable.
<code>types</code>	If set, this should be a list with six elements, ordinarily with each list element either "p" or <code>c("p", "smooth")</code> or ( <code>which=2</code> , <code>which=6</code> ) <code>NULL</code> or ( <code>which=4</code> ) "h"
<code>which</code>	Set to be a subset of the numbers 1 to 6, as for <code>plot.lm</code>
<code>layout</code>	Controls the number of simulations and the layout of the plots. For example <code>layout=c(3,4)</code> will give 12 plots in a 3 by 4 layout.
<code>qqline</code>	logical: add line to normal Q-Q plot
<code>cook.levels</code>	Levels of Cook's statistics for which contours are to be plotted.
<code>caption</code>	list: Captions for the six graphs
<code>...</code>	Other parameters to be passed to plotting functions

**Details**

Diagnostic plots from repeated simulations from the fitted model provide a useful indication of the range of variation in the model diagnostics that are consistent with the fitted model.

**Value**

A list of lattice graphics objects is returned, one for each value of `which`. List elements for which a graphics object is not returned are set to `NULL`. Or if `which` is of length 1, a lattice graphics object.

`residVSfitted`    Residuals vs fitted

normalQQ	Normal quantile-quantile plot
scaleVSloc	Scale versus location
CookDist	Cook's distance vs observation number
residVSlev	Standardized residuals (for GLMs, standardized Pearson residuals) vs leverage
CookVSlev	Cook's distance vs leverage

For the default `which=c(1:3,5)`, list items 1, 2, 3 and 5 above contain graphics objects, with list elements 4 and 6 set to `NULL`.

### Note

The graphics objects contained in individual list elements can be extracted for printing, or updating and printing, as required. If the value is returned to the command line, list elements that are not `NULL` will be printed in turn.

### Author(s)

John Maindonald, with some code chunks adapted from `plot.lm`

### References

See [plot.lm](#)

### See Also

`codeplot.lm`, `codeImdiags`

### Examples

```
women.lm <- lm(height ~ weight, data=women)
gphlist <- plotSimDiags(obj=women.lm, which=c(1:3,5))
```

---

plotSimScat	<i>Simulate scatterplots, from lm object with a single explanatory variable.</i>
-------------	--

---

### Description

This plots simulated y-values, or residuals from such simulations, against x-values .

### Usage

```
plotSimScat(obj, sigma = NULL, layout = c(4, 1), type = c("p", "r"),
show = c("points", "residuals"), ...)
```

**Arguments**

obj	An lm object with a single explanatory variable.
sigma	Standard deviation, if different from that for the supplied lm object.
layout	Columns by Rows layout for plots from the simulations.
type	See type as in <a href="#">plot.lm</a> .
show	Specify points or residuals.
...	Other parameters to be passed to plotting functions

**Value**

A lattice graphics object is returned.

**Author(s)**

J H Maindonald

**See Also**

[plotSimDiags](#)

**Examples**

```
nihills.lm <- lm(timef~time, data=nihills)
plotSimDiags(nihills.lm)

## The function is currently defined as
function (obj, sigma = NULL, layout = c(4, 1), type = c("p",
  "r"), show = c("points", "residuals"))
{
  nsim <- prod(layout)
  if (is.null(sigma))
    sigma <- summary(obj)[["sigma"]]
  hat <- fitted(obj)
  xnam <- all.vars(formula(obj))[2]
  ynam <- all.vars(formula(obj))[1]
  df <- data.frame(sapply(1:nsim, function(x) rnorm(length(hat),
    sd = sigma)))
  if (show[1] == "points")
    df <- df + hat
  simnam <- names(df) <- paste("Simulation", 1:nsim, sep = "")
  df[, c(xnam, ynam)] <- model.frame(obj)[, c(xnam, ynam)]
  if (show[1] != "points") {
    df[, "Residuals"] <- df[, ynam] - hat
    ynam <- "Residuals"
    legadd <- "residuals"
  }
  else legadd <- "data"
  leg <- list(text = paste(c("Simulated", "Actual"), legadd),
    columns = 2)
  formula <- formula(paste(paste(simnam, collapse = "+"), "~",
```

```

      xnam))
  parset <- simpleTheme(pch = c(16, 16), lty = 2, col = c("black",
    "gray"))
  gph <- xyplot(formula, data = df, outer = TRUE, par.settings = parset,
    auto.key = leg, lty = 2, layout = layout, type = type)
  formxy <- formula(paste(ynam, "~", xnam))
  addgph <- xyplot(formxy, data = df, pch = 16, col = "gray")
  gph + as.layer(addgph, under = TRUE)
}

```

poissonsim

*Simple Poisson Regression Data Simulator***Description**

This function simulates simple regression data from a Poisson model. It also has the option to create over-dispersed data of a particular type.

**Usage**

```
poissonsim(x = seq(0, 1, length=101), a = 2, b = -4, intcp.sd=NULL,
  slope.sd=NULL, seed=NULL)
```

**Arguments**

x	a numeric vector representing the explanatory variable
a	the regression function intercept
b	the regression function slope
intcp.sd	standard deviation of the (random) intercept
slope.sd	standard deviation of the (random) slope
seed	numeric constant

**Value**

a list consisting of

x	the explanatory variable vector
y	the Poisson response vector

**Examples**

```
poissonsim()
```

possum

*Possum Measurements***Description**

The possum data frame consists of nine morphometric measurements on each of 104 mountain brushtail possums, trapped at seven sites from Southern Victoria to central Queensland.

**Usage**

possum

**Format**

This data frame contains the following columns:

**case** observation number

**site** one of seven locations where possums were trapped

**Pop** a factor which classifies the sites as Vic Victoria, other New South Wales or Queensland

**sex** a factor with levels f female, m male

**age** age

**hdlngth** head length

**skullw** skull width

**totlngth** total length

**taill** tail length

**footlgth** foot length

**earconch** ear conch length

**eye** distance from medial canthus to lateral canthus of right eye

**chest** chest girth (in cm)

**belly** belly girth (in cm)

**Source**

Lindenmayer, D. B., Viggers, K. L., Cunningham, R. B., and Donnelly, C. F. 1995. Morphological variation among columns of the mountain brushtail possum, *Trichosurus caninus* Ogilby (Phalangeridae: Marsupiala). *Australian Journal of Zoology* 43: 449-458.



**Examples**

```

boxplot(earconch~sex, data=possum)
pause()

sex <- as.integer(possum$sex)
oldpar <- par(oma=c(2,4,5,4))
pairs(possum[, c(9:11)], pch=c(0,2:7), col=c("red","blue"),
      labels=c("tail\nlength","foot\nlength","ear conch\nlength"))
chh <- par()$cxy[2]; xleg <- 0.05; yleg <- 1.04
oldpar <- par(xpd=TRUE)
legend(xleg, yleg, c("Cambarville", "Bellbird", "Whian Whian ",
  "Byranger", "Conondale ", "Allyn River", "Bulburin"), pch=c(0,2:7),
  x.intersp=1, y.intersp=0.75, cex=0.8, xjust=0, bty="n", ncol=4)
text(x=0.2, y=yleg - 2.25*chh, "female", col="red", cex=0.8, bty="n")
text(x=0.75, y=yleg - 2.25*chh, "male", col="blue", cex=0.8, bty="n")
par(oldpar)
pause()

sapply(possum[,6:14], function(x)max(x,na.rm=TRUE)/min(x,na.rm=TRUE))
pause()

here <- na.omit(possum$footlgth)
possum.prc <- princomp(possum[here, 6:14])
pause()

plot(possum.prc$scores[,1] ~ possum.prc$scores[,2],
     col=c("red","blue")[as.numeric(possum$sex[here])],
     pch=c(0,2:7)[possum$site[here]], xlab = "PC1", ylab = "PC2")
# NB: We have abbreviated the axis titles
chh <- par()$cxy[2]; xleg <- -15; yleg <- 20.5
oldpar <- par(xpd=TRUE)
legend(xleg, yleg, c("Cambarville", "Bellbird", "Whian Whian ",
  "Byranger", "Conondale ", "Allyn River", "Bulburin"), pch=c(0,2:7),
  x.intersp=1, y.intersp=0.75, cex=0.8, xjust=0, bty="n", ncol=4)
text(x=-9, y=yleg - 2.25*chh, "female", col="red", cex=0.8, bty="n")
summary(possum.prc, loadings=TRUE, digits=2)
par(oldpar)
pause()

require(MASS)
here <- !is.na(possum$footlgth)
possum.lda <- lda(site ~ hdlngth+skullw+totlngth+ taill+footlgth+
  earconch+eye+chest+belly, data=possum, subset=here)
options(digits=4)
possum.lda$svd # Examine the singular values
plot(possum.lda, dimen=3)
# Scatterplot matrix - scores on 1st 3 canonical variates (Figure 11.4)
possum.lda

```

**Description**

The possumsites data frame consists of Longitudes, Latitudes, and altitudes for the seven sites from Southern Victoria to central Queensland where the possum observations were made.

**Usage**

possumsites

**Format**

This data frame contains the following columns:

- Longitude** a numeric vector
- Latitude** a numeric vector
- altitude** in meters

**Source**

Lindenmayer, D. B., Viggers, K. L., Cunningham, R. B., and Donnelly, C. F. 1995. Morphological variation among columns of the mountain brushtail possum, *Trichosurus caninus* Ogilby (Phalangeridae: Marsupiala). Australian Journal of Zoology 43: 449-458.

**Examples**

```
require(oz)
oz(sections=c(3:5, 11:16))
attach(possumsites)
points(Longitude, Latitude, pch=16, col=2)
chw <- par()$cxy[1]
chh <- par()$cxy[2]
posval <- c(2,4,2,2,4,2,2)
text(Longitude+(3-posval)*chw/4, Latitude, row.names(possumsites), pos=posval)
```

---

powerplot	<i>Plot of Power Functions</i>
-----------	--------------------------------

---

**Description**

This function plots powers of a variable on the interval [0,10].

**Usage**

powerplot(expr="x^2", xlab="x", ylab="y")

**Arguments**

- |      |                               |
|------|-------------------------------|
| expr | Functional form to be plotted |
| xlab | x-axis label                  |
| ylab | y-axis label                  |

**Details**

Other expressions such as "sin(x)" and "cos(x)", etc. could also be plotted with this function, but results are not guaranteed.

**Value**

A plot of the given expression on the interval [0,10].

**Author(s)**

J.H. Maindonald

**Examples**

```
oldpar <- par(mfrow = c(2, 3), mar = par()$mar - c(
  1, 1, 1.0, 1), mgp = c(1.5, 0.5, 0), oma=c(0,1,0,1))
# on.exit(par(oldpar))
powerplot(expr="sqrt(x)", xlab="")
powerplot(expr="x^0.25", xlab="", ylab="")
powerplot(expr="log(x)", xlab="", ylab="")
powerplot(expr="x^2")
powerplot(expr="x^4", ylab="")
powerplot(expr="exp(x)", ylab="")
par(oldpar)
```

---

poxetc

*Deaths from various causes, in London from 1629 till 1881, with gaps*

---

**Description**

Deaths from "flux" or smallpox, measles, all causes, and ratios of the the first two categories to total deaths.

**Usage**

```
data(poxetc)
```

**Format**

This is a multiple time series consisting of 5 series: fpox, measles, all, fpox2all, measles2all.

**Source**

Guy, W. A. 1882. Two hundred and fifty years of small pox in London. Journal of the Royal Statistical Society 399-443.

**References**

Lancaster, H. O. 1990. Expectations of Life. Springer.

**Examples**

```
data(poxetc)
str(poxetc)
plot(poxetc)
```

---

press

*Predictive Error Sum of Squares*

---

**Description**

Allen's PRESS statistic is computed for a fitted model.

**Usage**

```
press(obj)
```

**Arguments**

obj                    A lm object

**Value**

A single numeric value.

**Author(s)**

W.J. Braun

**See Also**

lm

**Examples**

```
litters.lm <- lm(brainwt ~ bodywt + lsize, data = litters)
press(litters.lm)
litters.lm0 <- lm(brainwt ~ bodywt + lsize -1, data=litters)
press(litters.lm0) # no intercept
litters.lm1 <- lm(brainwt ~ bodywt, data=litters)
press(litters.lm1) # bodywt only
litters.lm2 <- lm(brainwt ~ bodywt + lsize + lsize:bodywt, data=litters)
press(litters.lm2) # include an interaction term
```

---

 primates

*Primate Body and Brain Weights*


---

### Description

A subset of `Animals` data frame from the MASS library. It contains the average body and brain measurements of five primates.

### Usage

```
primates
```

### Format

This data frame contains the following columns:

**Bodywt** a numeric vector consisting of the body weights (in kg) of five different primates

**Brainwt** a numeric vector consisting of the corresponding brain weights (in g)

### Source

P. J. Rousseeuw and A. M. Leroy (1987) *Robust Regression and Outlier Detection*. Wiley, p. 57.

### Examples

```
attach(primates)
plot(x=Bodywt, y=Brainwt, pch=16,
     xlab="Body weight (kg)", ylab="Brain weight (g)",
     xlim=c(5,300), ylim=c(0,1500))
chw <- par()$cxy[1]
chh <- par()$cxy[2]
text(x=Bodywt+chw, y=Brainwt+c(-.1,0,0,.1,0)*chh,
     labels=row.names(primates), adj=0)
detach(primates)
```

---

 progression

*Progression of Record times for track races, 1912 - 2008*


---

### Description

Progression in world record times for track and road races.

### Usage

```
data(progression)
```

**Format**

A data frame with 227 observations on the following 4 columns.

- year Year that time was first recorded
- Distance distance in kilometers
- Time time in minutes
- race character; descriptor for event (100m, mile, ...)

**Details**

Record times for men’s track events, from 1912 onwards. The series starts with times that were recognized as record times in 1912, where available.

**Source**

Links to sources for the data are at  
[http://en.wikipedia.org/wiki/Athletics\\_world\\_record](http://en.wikipedia.org/wiki/Athletics_world_record)

**Examples**

```
data(progression)
plot(log(Time) ~ log(Distance), data=progression)
xyplot(log(Time) ~ log(Distance), data=progression, type=c("p","r"))
xyplot(log(Time) ~ log(Distance), data=progression,
       type=c("p","smooth"))
res <- resid(lm(log(Time) ~ log(Distance), data=progression))
plot(res ~ log(Distance), data=progression,
     ylab="Residuals from regression line on log scales")
```

---

psid1	<i>Labour Training Evaluation Data</i>
-------	--

---

**Description**

A non-experimental "control" group, used in various studies of the effect of a labor training program, alternative to the experimental control group in nswdemo.

**Usage**

psid1

**Format**

This data frame contains the following columns:

**trt** a numeric vector identifying the study in which the subjects were enrolled (0 = Control, 1 = treated).

**age** age (in years).

**educ** years of education.

**black** (0 = not black, 1 = black).

**hisp** (0 = not hispanic, 1 = hispanic).

**marr** (0 = not married, 1 = married).

**nodeg** (0 = completed high school, 1 = dropout).

**re74** real earnings in 1974.

**re75** real earnings in 1975.

**re78** real earnings in 1978.

**Details**

The cps1 and psid1 data sets are two non-experimental "control" groups, alternative to that in nswdemo, used in investigating whether use of such a non-experimental control group can be satisfactory. cps2 and cps3 are subsets of cps1, designed to be better matched to the experimental data than cps1. Similarly psid2 and psid3 are subsets of psid1, designed to be better matched to the experimental data than psid1.

**Source**

<http://www.nber.org/~rdehejia/nswddata.html>

**References**

Dehejia, R.H. and Wahba, S. 1999. Causal effects in non-experimental studies: re-evaluating the evaluation of training programs. *Journal of the American Statistical Association* 94: 1053-1062.

Lalonde, R. 1986. Evaluating the economic evaluations of training programs. *American Economic Review* 76: 604-620.

Smith, J. A. and Todd, P.E. "Does Matching overcome. LaLonde's critique of nonexperimental estimators", *Journal of Econometrics* 125: 305-353.

Dehejia, R.H. 2005. Practical propensity score matching: a reply to Smith and Todd. *Journal of Econometrics* 125: 355-364.

psid2

*Labour Training Evaluation Data***Description**

A non-experimental "control" group, used in various studies of the effect of a labor training program, alternative to the experimental control group in nswdemo.

**Usage**

psid2

**Format**

This data frame contains the following columns:

**trt** a numeric vector identifying the study in which the subjects were enrolled (0 = Control, 1 = treated).

**age** age (in years).

**educ** years of education.

**black** (0 = not black, 1 = black).

**hisp** (0 = not hispanic, 1 = hispanic).

**marr** (0 = not married, 1 = married).

**nodeg** (0 = completed high school, 1 = dropout).

**re74** real earnings in 1974.

**re75** real earnings in 1975.

**re78** real earnings in 1978.

**Details**

The cps1 and psid1 data sets are two non-experimental "control" groups, alternative to that in nswdemo, used in investigating whether use of such a non-experimental control group can be satisfactory. cps2 and cps3 are subsets of cps1, designed to be better matched to the experimental data than cps1. Similarly psid2 and psid3 are subsets of psid1, designed to be better matched to the experimental data than psid1.

**Source**

<http://www.nber.org/~rdehejia/nswdata.html>



## References

- Dehejia, R.H. and Wahba, S. 1999. Causal effects in non-experimental studies: re-evaluating the evaluation of training programs. *Journal of the American Statistical Association* 94: 1053-1062.
- Lalonde, R. 1986. Evaluating the economic evaluations of training programs. *American Economic Review* 76: 604-620.
- Smith, J. A. and Todd, P.E. "Does Matching overcome. LaLonde's critique of nonexperimental estimators", *Journal of Econometrics* 125: 305-353.
- Dehejia, R.H. 2005. Practical propensity score matching: a reply to Smith and Todd. *Journal of Econometrics* 125: 355-364.

---

 psid3

---

*Labour Training Evaluation Data*


---

## Description

A non-experimental "control" group, used in various studies of the effect of a labor training program, alternative to the experimental control group in nswdemo.

## Usage

psid3

## Format

This data frame contains the following columns:

- trt** a numeric vector identifying the study in which the subjects were enrolled (0 = Control, 1 = treated).
- age** age (in years).
- educ** years of education.
- black** (0 = not black, 1 = black).
- hisp** (0 = not hispanic, 1 = hispanic).
- marr** (0 = not married, 1 = married).
- nodeg** (0 = completed high school, 1 = dropout).
- re74** real earnings in 1974.
- re75** real earnings in 1975.
- re78** real earnings in 1978.

## Details

The cps1 and psid1 data sets are two non-experimental "control" groups, alternative to that in nswdemo, used in investigating whether use of such a non-experimental control group can be satisfactory. cps2 and cps3 are subsets of cps1, designed to be better matched to the experimental data than cps1. Similary psid2 and psid3 are subsets of psid1, designed to be better matched to the experimental data than psid1.

**Source**

<http://www.nber.org/~rdehejia/nswdata.html>

**References**

- Dehejia, R.H. and Wahba, S. 1999. Causal effects in non-experimental studies: re-evaluating the evaluation of training programs. *Journal of the American Statistical Association* 94: 1053-1062.
- Lalonde, R. 1986. Evaluating the economic evaluations of training programs. *American Economic Review* 76: 604-620.
- Smith, J. A. and Todd, P.E. "Does Matching overcome. LaLonde's critique of nonexperimental estimators", *Journal of Econometrics* 125: 305-353.
- Dehejia, R.H. 2005. Practical propensity score matching: a reply to Smith and Todd. *Journal of Econometrics* 125: 355-364.

---

qreference

Normal QQ Reference Plot

---

**Description**

This function computes the normal QQ plot for given data and allows for comparison with normal QQ plots of simulated data.

**Usage**

```
qreference(test = NULL, m = 50, nrep = 6, distribution = function(x) qnorm(x,
  mean = ifelse(is.null(test), 0, mean(test)), sd = ifelse(is.null(test),
  1, sd(test))), seed = NULL, nrows = NULL, cex.strip = 0.75,
  xlab = NULL, ylab = NULL)
```

**Arguments**

test	a vector containing a sample to be tested; if not supplied, all qq-plots are of the reference distribution
m	the sample size for the reference samples; default is test sample size if test sample is supplied
nrep	the total number of samples, including reference samples and test sample if any
distribution	reference distribution; default is standard normal
seed	the random number generator seed
nrows	number of rows in the plot layout
cex.strip	character expansion factor for labels
xlab	label for x-axis
ylab	label for y-axis

**Value**

QQ plots of the sample (if test is non-null) and all reference samples

**Author(s)**

J.H. Maindonald

**Examples**

```
# qreference(rt(180,1))

# qreference(rt(180,1), distribution=function(x) qt(x, df=1))

# qreference(rexp(180), nrep = 4)

# toycars.lm <- lm(distance ~ angle + factor(car), data = toycars)
# qreference(residuals(toycars.lm), nrep = 9)
```

---

races2000

*Scottish Hill Races Data - 2000*


---

**Description**

The record times in 2000 for 77 Scottish long distance races. We believe the data are, for the most part, trustworthy. However, the `dist` variable for Caerketton (record 58) seems to have been variously recorded as 1.5 mi and 2.5 mi.

**Usage**

```
races2000
```

**Format**

This data frame contains the following columns:

**dist** distance, in miles (on the map)

**climb** total height gained during the route, in feet

**time** record time in hours

**timef** record time in hours for females

**type** a factor, with levels indicating type of race, i.e. hill, marathon, relay, uphill or other

**Source**

The Scottish Running Resource, <http://www.hillrunning.co.uk>

**Examples**

```
pairs(races2000[,-5])
```

---

`rainforest`*Rainforest Data*

---

**Description**

The rainforest data frame has 65 rows and 7 columns.

**Usage**

```
rainforest
```

**Format**

This data frame contains the following columns:

**dbh** a numeric vector

**wood** a numeric vector

**bark** a numeric vector

**root** a numeric vector

**rootsk** a numeric vector

**branch** a numeric vector

**species** a factor with levels *Acacia mabellae*, *C. fraseri*, *Acmena smithii*, *B. myrtifolia*

**Source**

J. Ash, Australian National University

**References**

Ash, J. and Helman, C. (1990) Floristics and vegetation biomass of a forest catchment, Kioloa, south coastal N.S.W. *Cunninghamia*, 2: 167-182.

**Examples**

```
table(rainforest$species)
```

---

`rareplants`*Rare and Endangered Plant Species*

---

**Description**

These data were taken from species lists for South Australia, Victoria and Tasmania. Species were classified as CC, CR, RC and RR, with C denoting common and R denoting rare. The first code relates to South Australia and Victoria, and the second to Tasmania. They were further classified by habitat according to the Victorian register, where D = dry only, W = wet only, and WD = wet or dry.

**Usage**`rareplants`**Format**

The format is: chr "rareplants"

**Source**

Jasmyn Lynch, Department of Botany and Zoology at Australian National University

**Examples**

```
chisq.test(rareplants)
```

---

`rice`*Genetically Modified and Wild Type Rice Data*

---

**Description**

The rice data frame has 72 rows and 7 columns. The data are from an experiment that compared wild type (wt) and genetically modified rice plants (ANU843), each with three different chemical treatments (F10, NH4Cl, and NH4NO3).

**Usage**`rice`

**Format**

This data frame contains the following columns:

**PlantNo** a numeric vector

**Block** a numeric vector

**RootDryMass** a numeric vector

**ShootDryMass** a numeric vector

**trt** a factor with levels F10, NH4Cl, NH4NO3, F10 +ANU843, NH4Cl +ANU843, NH4NO3 +ANU843

**fert** a factor with levels F10 NH4Cl NH4NO3

**variety** a factor with levels wt ANU843

**Source**

Perrine, F.M., Prayitno, J., Weinman, J.J., Dazzo, F.B. and Rolfe, B. 2001. Rhizobium plasmids are involved in the inhibition or stimulation of rice growth and development. Australian Journal of Plant Physiology 28: 923-927.

**Examples**

```
print("One and Two-Way Comparisons - Example 4.5")
attach(rice)
oldpar <- par(las = 2)
stripchart(ShootDryMass ~ trt, pch=1, cex=1, xlab="Level of factor 1")
detach(rice)
pause()

rice.aov <- aov(ShootDryMass ~ trt, data=rice); anova(rice.aov)
anova(rice.aov)
pause()

summary.lm(rice.aov)$coef
pause()

rice$trt <- relevel(rice$trt, ref="NH4Cl")
# Set NH4Cl as the baseline

fac1 <- factor(sapply(strsplit(as.character(rice$trt), " \\+"), function(x)x[1]))
anu843 <- sapply(strsplit(as.character(rice$trt), " \\+"),
function(x)c("wt", "ANU843")[length(x)])
anu843 <- factor(anu843, levels=c("wt", "ANU843"))
attach(rice)
interaction.plot(fac1, anu843, ShootDryMass)
detach(rice)
par(oldpar)
```

---

rockArt	<i>Pacific Rock Art features</i>
---------	----------------------------------

---

**Description**

Data characterise rock art at 103 sites in the Pacific.

**Usage**

rockArt

**Format**

A data frame with 103 observations on the following 641 variables.

Site.No. a numeric vector

Site.Name a character vector

Site.Code a character vector

District a character vector

Island a character vector

Country a character vector

Technique a character vector

Engtech a character vector

red a numeric vector

black a numeric vector

yellow a numeric vector

white a numeric vector

green a numeric vector

red.blk a numeric vector

red.wh a numeric vector

red.yell a numeric vector

r.w.y a numeric vector

black.white a numeric vector

blue a numeric vector

Geology a character vector

Topography a character vector

Location a character vector

Proxhab.km. a character vector

Proxcoast.km. a numeric vector

Maxheight.m. a numeric vector

Language a character vector

No.motif a character vector

Ca1 a numeric vector

Ca2 a numeric vector

Ca3 a numeric vector

Ca4 a numeric vector

Cb5 a numeric vector

Cb6 a numeric vector

Cc7 a numeric vector

Cc8 a numeric vector

Cc9 a numeric vector

Cc10 a numeric vector

Cc11 a numeric vector

Cc12 a numeric vector

Cc13 a numeric vector

Cc14 a numeric vector

Cc15 a numeric vector

Cc16 a numeric vector

Cc17 a numeric vector

Cc18 a numeric vector

Cc19 a numeric vector

Cc20 a numeric vector

Cd21 a numeric vector

Cd22 a numeric vector

Cd23 a numeric vector

Cd24 a numeric vector

Cd25 a numeric vector

Cd26 a numeric vector

Cd27 a numeric vector

Ce28 a numeric vector

Ce29 a numeric vector

Cf30 a numeric vector

Cf31 a numeric vector

Cf32 a numeric vector

Cf33 a numeric vector

Cf34 a numeric vector

Cf35 a numeric vector



Cf36 a numeric vector  
Cf37 a numeric vector  
Cf38 a numeric vector  
Cg39 a numeric vector  
Cg40 a numeric vector  
Ch41 a numeric vector  
Ch42 a numeric vector  
Ci43 a numeric vector  
Ci44 a numeric vector  
Cj45 a numeric vector  
Ck46 a numeric vector  
Ck47 a numeric vector  
Cl48 a numeric vector  
Cm49 a numeric vector  
Cm50 a numeric vector  
Cm51 a numeric vector  
Cm52 a numeric vector  
Cm53 a numeric vector  
Cm54 a numeric vector  
Cm55 a numeric vector  
Cm56 a numeric vector  
Cm57 a numeric vector  
Cm58 a numeric vector  
Cn59 a numeric vector  
Cn60 a numeric vector  
Cn61 a numeric vector  
Cn62 a numeric vector  
Cn63 a numeric vector  
Cn64 a numeric vector  
Cn65 a numeric vector  
Cn66 a numeric vector  
Cn67 a numeric vector  
Cn68 a numeric vector  
Cn69 a numeric vector  
Cn70 a numeric vector  
Cn71 a numeric vector  
Co72 a numeric vector

Co73 a numeric vector  
Co74 a numeric vector  
Co75 a numeric vector  
Co76 a numeric vector  
Co77 a numeric vector  
Co78 a numeric vector  
Co79 a numeric vector  
Cp80 a numeric vector  
Cq81 a numeric vector  
Cq82 a numeric vector  
Cq83 a numeric vector  
Cq84 a numeric vector  
Cq85 a numeric vector  
Cq86 a numeric vector  
Cq87 a numeric vector  
Cq88 a numeric vector  
Cq89 a numeric vector  
Cq90 a numeric vector  
Cq91 a numeric vector  
Cq92 a numeric vector  
Cq93 a numeric vector  
Cq94 a numeric vector  
Cq95 a numeric vector  
Cq96 a numeric vector  
Cq97 a numeric vector  
Cr98 a numeric vector  
Cr99 a numeric vector  
Cr100 a numeric vector  
Cr101 a numeric vector  
Cs102 a numeric vector  
Cs103 a numeric vector  
Cs104 a numeric vector  
Cs105 a numeric vector  
Cs106 a numeric vector  
Ct107 a numeric vector  
C108 a numeric vector  
C109 a numeric vector

C110 a numeric vector  
C111 a numeric vector  
SSa1 a numeric vector  
SSd2 a numeric vector  
SSd3 a numeric vector  
SSd4 a numeric vector  
SSd5 a numeric vector  
SSd6 a numeric vector  
SSd7 a numeric vector  
SSd8 a numeric vector  
SSf9 a numeric vector  
SSg10 a numeric vector  
SSj11 a numeric vector  
SSj12 a numeric vector  
SSj13 a numeric vector  
SSl14 a numeric vector  
SSm15 a numeric vector  
SSm16 a numeric vector  
SSn17 a numeric vector  
SSn18 a numeric vector  
SSn19 a numeric vector  
SSn20 a numeric vector  
SSn21 a numeric vector  
SSn22 a numeric vector  
SSn23 a numeric vector  
SSn24 a numeric vector  
SSn25 a numeric vector  
SSn26 a numeric vector  
SSn27 a numeric vector  
SSn28 a numeric vector  
SSn29 a numeric vector  
SSn30 a numeric vector  
SSn31 a numeric vector  
SSn32 a numeric vector  
SSn33 a numeric vector  
SSn34 a numeric vector  
SSn35 a numeric vector

SSo36 a numeric vector  
SSo37 a numeric vector  
SSp38 a numeric vector  
SSq39 a numeric vector  
SSq40 a numeric vector  
SSt41 a numeric vector  
SSu42 a numeric vector  
Oa1 a numeric vector  
Oc2 a numeric vector  
Od3 a numeric vector  
Od4 a numeric vector  
Oe5 a numeric vector  
Of6 a numeric vector  
Of7 a numeric vector  
Of8 a numeric vector  
Of9 a numeric vector  
Og10 a numeric vector  
Og11 a numeric vector  
Og12 a numeric vector  
Og13 a numeric vector  
Og14 a numeric vector  
Og15 a numeric vector  
Oi16 a numeric vector  
Om17 a numeric vector  
Om18 a numeric vector  
Om19 a numeric vector  
Om20 a numeric vector  
Om21 a numeric vector  
On22 a numeric vector  
On23 a numeric vector  
On24 a numeric vector  
Oq25 a numeric vector  
Oq26 a numeric vector  
Oq27 a numeric vector  
.u28 a numeric vector  
Ov29 a numeric vector  
Ov30 a numeric vector

031 a numeric vector  
032 a numeric vector  
033 a numeric vector  
Sa1 a numeric vector  
Sb2 a numeric vector  
Sb3 a numeric vector  
Sd4 a numeric vector  
Sd5 a numeric vector  
Sd6 a numeric vector  
Sd7 a numeric vector  
Se8 a numeric vector  
Si9 a numeric vector  
Sm10 a numeric vector  
Sm11 a numeric vector  
S12 a numeric vector  
S13 a numeric vector  
Sx14 a numeric vector  
Sx15 a numeric vector  
Sx16 a numeric vector  
Sx17 a numeric vector  
Sy18 a numeric vector  
Sz19 a numeric vector  
S20 a numeric vector  
S21 a numeric vector  
S22 a numeric vector  
S23 a numeric vector  
S24 a numeric vector  
S25 a numeric vector  
SCd1 a numeric vector  
SCd2 a numeric vector  
SCd3 a numeric vector  
SCd4 a numeric vector  
SCd5 a numeric vector  
SCd6 a numeric vector  
SCd7 a numeric vector  
SCm8 a numeric vector  
SCn9 a numeric vector

SCn10 a numeric vector  
SCw11 a numeric vector  
SCx12 a numeric vector  
SCx13 a numeric vector  
SCx14 a numeric vector  
SCx15 a numeric vector  
SCx16 a numeric vector  
SCy17 a numeric vector  
SCy18 a numeric vector  
SC19 a numeric vector  
SC20 a numeric vector  
SC21 a numeric vector  
SC22 a numeric vector  
SC23 a numeric vector  
SC24 a numeric vector  
SC25 a numeric vector  
SC26 a numeric vector  
SRd1 a numeric vector  
SRd2 a numeric vector  
SRd3 a numeric vector  
SRd4 a numeric vector  
SRf5 a numeric vector  
SRf6 a numeric vector  
SRf7 a numeric vector  
SRj8 a numeric vector  
SR9 a numeric vector  
SR10 a numeric vector  
Bd1 a numeric vector  
Bn2 a numeric vector  
Bn3 a numeric vector  
Bn4 a numeric vector  
Bt5 a numeric vector  
Bx6 a numeric vector  
Ha1 a numeric vector  
Hg2 a numeric vector  
Hn3 a numeric vector  
Hq4 a numeric vector

Hq5 a numeric vector  
TDd1 a numeric vector  
TDf2 a numeric vector  
TDj3 a numeric vector  
TDn4 a numeric vector  
TDq5 a numeric vector  
TD6 a numeric vector  
TD7 a numeric vector  
TD8 a numeric vector  
TD9 a numeric vector  
Dc1 a numeric vector  
Dg2 a numeric vector  
Dh3 a numeric vector  
Dk4 a numeric vector  
Dm5 a numeric vector  
Dm6 a numeric vector  
D7 a numeric vector  
D8 a numeric vector  
D9 a numeric vector  
D10 a numeric vector  
D11 a numeric vector  
D12 a numeric vector  
D13 a numeric vector  
Ta1 a numeric vector  
Tc2 a numeric vector  
Tc3 a numeric vector  
Tc4 a numeric vector  
Td5 a numeric vector  
Tf6 a numeric vector  
Tf7 a numeric vector  
Tg8 a numeric vector  
Th9 a numeric vector  
To10 a numeric vector  
T11 a numeric vector  
T12 a numeric vector  
T13 a numeric vector  
T14 a numeric vector

T15 a numeric vector  
T16 a numeric vector  
CNg1 a numeric vector  
CN2 a numeric vector  
CN3 a numeric vector  
CN4 a numeric vector  
CN5 a numeric vector  
CN6 a numeric vector  
CN7 a numeric vector  
CN8 a numeric vector  
Ld1 a numeric vector  
Lf2 a numeric vector  
Lg3 a numeric vector  
Lp4 a numeric vector  
L5 a numeric vector  
L6 a numeric vector  
L7 a numeric vector  
L8 a numeric vector  
L9 a numeric vector  
L10 a numeric vector  
L11 a numeric vector  
LS1 a numeric vector  
LS2 a numeric vector  
LL1 a numeric vector  
LL2 a numeric vector  
LL3 a numeric vector  
LL4 a numeric vector  
LL5 a numeric vector  
EGd1 a numeric vector  
EGf2 a numeric vector  
CCd1 a numeric vector  
CCn2 a numeric vector  
CCn3 a numeric vector  
EMc1 a numeric vector  
EMd2 a numeric vector  
EMd3 a numeric vector  
EMf4 a numeric vector



EMf5 a numeric vector  
EMn6 a numeric vector  
EMx7 a numeric vector  
EM8 a numeric vector  
EM9 a numeric vector  
EM10 a numeric vector  
EM11 a numeric vector  
EM12 a numeric vector  
TE1 a numeric vector  
TE2 a numeric vector  
TE3 a numeric vector  
TE4 a numeric vector  
TE5 a numeric vector  
BWe1 a numeric vector  
BWn2 a numeric vector  
BWn3 a numeric vector  
TS1 a numeric vector  
TS2 a numeric vector  
TS3 a numeric vector  
TS4 a numeric vector  
TS5 a numeric vector  
TS6 a numeric vector  
TS7 a numeric vector  
TS8 a numeric vector  
TS9 a numeric vector  
Pg1 a numeric vector  
Pg2 a numeric vector  
Pg3 a numeric vector  
DUaa1 a numeric vector  
DUw2 a numeric vector  
DU3 a numeric vector  
CP1 a numeric vector  
CP2 a numeric vector  
CP3 a numeric vector  
CP4 a numeric vector  
CP5 a numeric vector  
CP6 a numeric vector

CP7 a numeric vector  
CP8 a numeric vector  
CP9 a numeric vector  
CP10 a numeric vector  
CP11 a numeric vector  
CP12 a numeric vector  
STd1 a numeric vector  
STd2 a numeric vector  
STd3 a numeric vector  
STg4 a numeric vector  
STaa5 a numeric vector  
STaa6 a numeric vector  
STaa7 a numeric vector  
STaa8 a numeric vector  
ST9 a numeric vector  
ST10 a numeric vector  
ST11 a numeric vector  
ST12 a numeric vector  
Wd1 a numeric vector  
Wd2 a numeric vector  
Wd3 a numeric vector  
Wd4 a numeric vector  
Wn5 a numeric vector  
Waa6 a numeric vector  
Waa7 a numeric vector  
W8 a numeric vector  
W9 a numeric vector  
W10 a numeric vector  
W11 a numeric vector  
W12 a numeric vector  
W13 a numeric vector  
Zd1 a numeric vector  
Zd2 a numeric vector  
Zn3 a numeric vector  
Zw4 a numeric vector  
Zw5 a numeric vector  
Zaa6 a numeric vector

Z7 a numeric vector  
Z8 a numeric vector  
Z9 a numeric vector  
Z10 a numeric vector  
Z11 a numeric vector  
Z12 a numeric vector  
CLd1 a numeric vector  
CLd2 a numeric vector  
CLd3 a numeric vector  
CLd4 a numeric vector  
CLd5 a numeric vector  
CLd6 a numeric vector  
CLd7 a numeric vector  
CLd8 a numeric vector  
CLd9 a numeric vector  
CLd10 a numeric vector  
CLd11 a numeric vector  
CLd12 a numeric vector  
CLd13 a numeric vector  
CLd14 a numeric vector  
CLd15 a numeric vector  
CLd16 a numeric vector  
CLd17 a numeric vector  
CLd18 a numeric vector  
CLd19 a numeric vector  
CLd20 a numeric vector  
CLd21 a numeric vector  
CLd22 a numeric vector  
CLd23 a numeric vector  
CLd24 a numeric vector  
CLd25 a numeric vector  
CLd26 a numeric vector  
CLd27 a numeric vector  
CLd28 a numeric vector  
CLd29 a numeric vector  
CLd30 a numeric vector  
CLd31 a numeric vector

CLd32 a numeric vector  
CLd33 a numeric vector  
CLd34 a numeric vector  
CLd35 a numeric vector  
CLd36 a numeric vector  
CLd37 a numeric vector  
CLd38 a numeric vector  
CLn39 a numeric vector  
CLn40 a numeric vector  
CLn41 a numeric vector  
CLn42 a numeric vector  
CLn43 a numeric vector  
CLn44 a numeric vector  
CLn45 a numeric vector  
CLn46 a numeric vector  
CLn47 a numeric vector  
CLn48 a numeric vector  
CLw49 a numeric vector  
CL50 a numeric vector  
CL51 a numeric vector  
CL52 a numeric vector  
CL53 a numeric vector  
CL54 a numeric vector  
CL55 a numeric vector  
CL56 a numeric vector  
CL57 a numeric vector  
CL58 a numeric vector  
CL59 a numeric vector  
Xd1 a numeric vector  
Xd2 a numeric vector  
Xd3 a numeric vector  
Xd4 a numeric vector  
Xd5 a numeric vector  
Xd6 a numeric vector  
Xd7 a numeric vector  
Xd8 a numeric vector  
Xd9 a numeric vector

Xd10 a numeric vector  
Xd11 a numeric vector  
Xd12 a numeric vector  
Xd13 a numeric vector  
Xf14 a numeric vector  
Xk15 a numeric vector  
Xn16 a numeric vector  
Xn17 a numeric vector  
Xn18 a numeric vector  
Xn19 a numeric vector  
Xn20 a numeric vector  
Xn21 a numeric vector  
Xn22 a numeric vector  
Xn23 a numeric vector  
Xn24 a numeric vector  
Xn25 a numeric vector  
Xn26 a numeric vector  
Xn27 a numeric vector  
Xn28 a numeric vector  
Xn29 a numeric vector  
Xn30 a numeric vector  
Xn31 a numeric vector  
Xn32 a numeric vector  
Xp33 a numeric vector  
Xp34 a numeric vector  
Xp35 a numeric vector  
Xq36 a numeric vector  
Xq37 a numeric vector  
Xq38 a numeric vector  
X39 a numeric vector  
X40 a numeric vector  
X41 a numeric vector  
X42 a numeric vector  
X43 a numeric vector  
X44 a numeric vector  
X45 a numeric vector  
X46 a numeric vector

X47 a numeric vector  
X48 a numeric vector  
X49 a numeric vector  
X50 a numeric vector  
Qd1 a numeric vector  
Qe2 a numeric vector  
Qe3 a numeric vector  
Qh4 a numeric vector  
Qh5 a numeric vector  
Qh6 a numeric vector  
Qh7 a numeric vector  
Qh8 a numeric vector  
Qh9 a numeric vector  
Qn10 a numeric vector  
Qn11 a numeric vector  
Qt12 a numeric vector  
Q13 a numeric vector  
Q14 a numeric vector  
Q15 a numeric vector  
Q16 a numeric vector  
Q17 a numeric vector  
Q18 a numeric vector  
Q19 a numeric vector  
Q20 a numeric vector  
Q21 a numeric vector  
Q22 a numeric vector  
TZd1 a numeric vector  
TZf2 a numeric vector  
TZh3 a numeric vector  
TZ4 a numeric vector  
CRd1 a numeric vector  
CR2 a numeric vector  
CR3 a numeric vector  
EUd1 a numeric vector  
EUd2 a numeric vector  
EUg3 a numeric vector  
EUm4 a numeric vector

EUw5 a numeric vector  
EU6 a numeric vector  
Ud1 a numeric vector  
Ud2 a numeric vector  
Ud3 a numeric vector  
Uaa4 a numeric vector  
U5 a numeric vector  
Vd1 a numeric vector  
V2 a numeric vector  
V3 a numeric vector  
V4 a numeric vector  
V5 a numeric vector  
LWE1 a numeric vector  
LWE2 a numeric vector  
Ad1 a numeric vector  
A12 a numeric vector  
Am3 a numeric vector  
An4 a numeric vector  
Aw5 a numeric vector  
Aaa6 a numeric vector  
A7 a numeric vector  
A8 a numeric vector  
A9 a numeric vector  
EVd1 a numeric vector  
EVg2 a numeric vector  
TK1 a numeric vector  
ECL1 a numeric vector  
EFe1 a numeric vector  
EFm2 a numeric vector  
EFm3 a numeric vector  
EF4 a numeric vector  
LPo1 a numeric vector  
LPq2 a numeric vector  
LP3 a numeric vector  
LP4 a numeric vector  
LP5 a numeric vector  
PT1 a numeric vector

CSC a numeric vector  
CSR a numeric vector  
CCRC a numeric vector  
SA a numeric vector  
Anthrop a numeric vector  
Turtle a numeric vector  
Boat a numeric vector  
Canoe a numeric vector  
Hand a numeric vector  
Foot a numeric vector  
Lizard a numeric vector  
Crocodile a numeric vector  
Jellyfish a numeric vector  
Bird a numeric vector  
Anthrobird a numeric vector  
Axe a numeric vector  
Marine a numeric vector  
Face a numeric vector  
Zoo1 a numeric vector  
Zoo2 a numeric vector  
Zoo3 a numeric vector  
Zoo4 a numeric vector  
Zoo5 a numeric vector  
Zoo6 a numeric vector

### Details

Note the vignette **rockArt**.

### Source

Meredith Wilson: *Picturing Pacific Pre-History* (PhD thesis), 2002, Australian National University.

### References

Meredith Wilson: Rethinking regional analyses of Western Pacific rock-art. *Records of the Australian Museum*, Supplement 29: 173-186.



## Examples

```
data(rockArt)
rockart.dist <- dist(x = as.matrix(rockArt[, 28:641]), method = "binary")
sum(rockart.dist==1)/length(rockart.dist)
plot(density(rockart.dist, to = 1))
rockart.cmd <- cmdscale(rockart.dist)
tab <- table(rockArt$District)
district <- as.character(rockArt$District)
district[!(rockArt$District %in% names(tab)[tab>5])] <- "other"
## Not run:
xyplot(rockart.cmd[,2] ~ rockart.cmd[,1], groups=district,
       auto.key=list(columns=5),
       par.settings=list(superpose.symbol=list(pch=16)))
library(MASS)
## For sammon, need to avoid zero distances
omit <- c(47, 54, 60, 63, 92)
rockart.dist <- dist(x = as.matrix(rockArt[-omit, 28:641]), method = "binary")
rockart.cmd <- cmdscale(rockart.dist)
rockart.sam <- sammon(rockart.dist, rockart.cmd)
xyplot(rockart.sam$points[,2] ~ rockart.sam$points[,1],
       groups=district[-omit], auto.key=list(columns=5),
       par.settings=list(superpose.symbol=list(pch=16)))
## Notice the very different appearance of the Sammon plot

## End(Not run)
```

---

roller

*Lawn Roller Data*


---

## Description

The roller data frame has 10 rows and 2 columns. Different weights of roller were rolled over different parts of a lawn, and the depression was recorded.

## Usage

```
roller
```

## Format

This data frame contains the following columns:

**weight** a numeric vector consisting of the roller weights

**depression** the depth of the depression made in the grass under the roller

## Source

Stewart, K.M., Van Toor, R.F., Crosbie, S.F. 1988. Control of grass grub (Coleoptera: Scarabaeidae) with rollers of different design. N.Z. Journal of Experimental Agriculture 16: 141-150.

**Examples**

```
plot(roller)
roller.lm <- lm(depression ~ weight, data = roller)
plot(roller.lm, which = 4)
```

sampdist

*Plot sampling distribution of mean or other sample statistic.***Description**

The function `sampvals` generates the data. A density plot of a normal probability plot is provided, for one or more sample sizes. For a density plot, the density estimate for the population is superimposed in gray. For the normal probability plot, the population plot is a dashed gray line. Default arguments give the sampling distribution of the mean, for a distribution that is mildly positively skewed.

**Usage**

```
sampdist(sampsize = c(3, 9, 30), seed = NULL, nsamp = 1000, FUN = mean,
        sampvals = function(n) exp(rnorm(n, mean = 0.5, sd = 0.3)),
        tck = NULL, plot.type = c("density", "qq"), layout = c(3, 1))
```

**Arguments**

<code>sampvals</code>	Function that generates the data. For sampling from existing data values, this might be function that generates bootstrap samples.
<code>sampsize</code>	One or more sample sizes. A plot will be provided for each different sample size.
<code>seed</code>	Specify a seed if it is required to make the exact set(s) of sample values reproducible.
<code>nsamp</code>	Number of samples.
<code>FUN</code>	Function that calculates the sample statistic.
<code>plot.type</code>	Specify density, or qq. Or if no plot is required, specify "".
<code>tck</code>	Tick size on lattice plots, by default 1, but 0.5 may be suitable for plots that are, for example, 50% of the default dimensions in each direction.
<code>layout</code>	Layout on page, e.g. <code>c(3, 1)</code> for a 3 columns by one row layout.

**Value**

Data frame

**Author(s)**

John Maindonald.

## Examples

```
sampdist(plot.type="density")
sampdist(plot.type="qq")

## The function is currently defined as
function (sampsizes = c(3, 9, 30), seed = NULL, nsamp = 1000, FUN = mean,
          sampvals = function(n) exp(rnorm(n, mean = 0.5, sd = 0.3)),
          tck = NULL, plot.type = c("density", "qq"), layout = c(3,
                                                                    1))
{
  if (!is.null(seed))
    set.seed(seed)
  ncases <- length(sampsizes)
  y <- sampvals(nsamp)
  xlim = quantile(y, c(0.01, 0.99))
  xlim <- xlim + c(-1, 1) * diff(xlim) * 0.1
  samplingDist <- function(sampsizes=3, nsamp=1000, FUN=mean)
    apply(matrix(sampvals(sampsizes*nsamp), ncol=sampsizes), 1, FUN)
  df <- data.frame(sapply(sampsizes, function(x)samplingDist(x, nsamp=nsamp)))
  names(df) <- paste("y", sampsizes, sep="")
  form <- formula(paste("~", paste(names(df), collapse="+")))
  lab <- lapply(sampsizes, function(x) substitute(A, list(A = paste(x))))
  if (plot.type[1] == "density")
    gph <- densityplot(form, data=df, layout = layout, outer=TRUE,
                       plot.points = FALSE, panel = function(x, ...) {
                         panel.densityplot(x, ..., col = "black")
                         panel.densityplot(y, col = "gray40", lty = 2,
                                             ...)
                       }, xlim = xlim, xlab = "", scales = list(tck = tck),
                       between = list(x = 0.5), strip = strip.custom(strip.names = TRUE,
                               factor.levels = as.expression(lab), var.name = "Sample size",
                               sep = expression(" = ")))
  else if (plot.type[1] == "qq")
    gph <- qqmath(form, data = df, layout = layout, plot.points = FALSE,
                  outer=TRUE,
                  panel = function(x, ...) {
                    panel.qqmath(x, ..., col = "black", alpha=0.5)
                    panel.qqmath(y, col = "gray40", lty = 2, type = "l",
                                  ...)
                  }, xlab = "", xlim = c(-3, 3), ylab = "", scales = list(tck = tck),
                  between = list(x = 0.5), strip = strip.custom(strip.names = TRUE,
                          factor.levels = as.expression(lab), var.name = "Sample size",
                          sep = expression(" = ")))
  if (plot.type[1] %in% c("density", "qq"))
    print(gph)
  invisible(df)
}
```

## Description

The science data frame has 1385 rows and 7 columns.

The data are on attitudes to science, from a survey where there were results from 20 classes in private schools and 46 classes in public schools.

## Usage

```
science
```

## Format

This data frame contains the following columns:

**State** a factor with levels ACT Australian Capital Territory, NSW New South Wales

**PrivPub** a factor with levels private school, public school

**school** a factor, coded to identify the school

**class** a factor, coded to identify the class

**sex** a factor with levels f, m

**like** a summary score based on two of the questions, on a scale from 1 (dislike) to 12 (like)

**Class** a factor with levels corresponding to each class

## Source

Francine Adams, Rosemary Martin and Murali Nayadu, Australian National University

## Examples

```
classmeans <- with(science, aggregate(like, by=list(PrivPub, Class), mean))
names(classmeans) <- c("PrivPub", "Class", "like")
dim(classmeans)

attach(classmeans)
boxplot(split(like, PrivPub), ylab = "Class average of attitude to science score", boxwex = 0.4)
rug(like[PrivPub == "private"], side = 2)
rug(like[PrivPub == "public"], side = 4)
detach(classmeans)
if(require(lme4, quietly=TRUE)) {
  science.lmer <- lmer(like ~ sex + PrivPub + (1 | school) +
    (1 | school:class), data = science,
    na.action=na.exclude)
  summary(science.lmer)
  science1.lmer <- lmer(like ~ sex + PrivPub + (1 | school:class),
    data = science, na.action=na.exclude)
  summary(science1.lmer)
  ranf <- ranef(obj = science1.lmer, drop=TRUE)[["school:class"]]
  flist <- science1.lmer@flist[["school:class"]]
  privpub <- science[match(names(ranf), flist), "PrivPub"]
  num <- unclass(table(flist)); numlabs <- pretty(num)
  ## Plot effect estimates vs numbers
```

```

plot(sqrt(num), ranf, xaxt="n", pch=c(1,3)[as.numeric(privpub)],
     xlab="# in class (square root scale)",
     ylab="Estimate of class effect")
lines(lowess(sqrt(num[privpub=="private"]),
             ranf[privpub=="private"], f=1.1), lty=2)
lines(lowess(sqrt(num[privpub=="public"]),
             ranf[privpub=="public"], f=1.1), lty=3)
axis(1, at=sqrt(numlabs), labels=paste(numlabs))
}

```

seedrates

*Barley Seeding Rate Data*

## Description

The seedrates data frame has 5 rows and 2 columns on the effect of seeding rate of barley on yield.

## Usage

```
seedrates
```

## Format

This data frame contains the following columns:

**rate** the seeding rate

**grain** the number of grain per head of barley

## Source

McLeod, C.C. 1982. Effect of rates of seeding on barley grown for grain. New Zealand Journal of Agriculture 10: 133-136.

## References

Maindonald J H 1992. Statistical design, analysis and presentation issues. New Zealand Journal of Agricultural Research 35: 121-141.

## Examples

```

plot(grain~rate,data=seedrates,xlim=c(50,180),ylim=c(15.5,22),axes=FALSE)
new.df<-data.frame(rate=(2:8)*25)
seedrates.lm1<-lm(grain~rate,data=seedrates)
seedrates.lm2<-lm(grain~rate+I(rate^2),data=seedrates)
hat1<-predict(seedrates.lm1,newdata=new.df,interval="confidence")
hat2<-predict(seedrates.lm2,newdata=new.df,interval="confidence")
axis(1,at=new.df$rate); axis(2); box()
z1<-spline(new.df$rate, hat1[, "fit"]); z2<-spline(new.df$rate,

```

```

hat2[, "fit"])
rate<-new.df$rate; lines(z1$x, z1$y)
lines(spline(rate, hat1[, "lwr"]), lty=1, col=3)
lines(spline(rate, hat1[, "upr"]), lty=1, col=3)
lines(z2$x, z2$y, lty=4)
lines(spline(rate, hat2[, "lwr"]), lty=4, col=3)
lines(spline(rate, hat2[, "upr"]), lty=4, col=3)

```

---

show.colors

*Show R's Colors*


---

## Description

This function displays the built-in colors.

## Usage

```
show.colors(type=c("singles", "shades", "gray"), order.cols=TRUE)
```

## Arguments

type	type of display - single, multiple or gray shades
order.cols	Arrange colors in order

## Value

A plot of colors for which there is a single shade (type = "single"), multiple shades (type = "multiple"), or gray shades (type = "gray")

## Author(s)

J.H. Maindonald

## Examples

```

require(MASS)
show.colors()

```

---

simulateLinear	<i>Simulation of Linear Models for ANOVA vs. Regression Comparison</i>
----------------	--

---

**Description**

This function simulates a number of bivariate data sets in which there are replicates at each level of the predictor. The p-values for ANOVA and for the regression slope are compared.

**Usage**

```
simulateLinear(sd=2, npoints=5, nrep=4, nsets=200, type="xy", seed=21)
```

**Arguments**

sd	The error standard deviation
npoints	Number of distinct predictor levels
nrep	Number of replications at each level
nsets	Number of simulation runs
type	Type of data
seed	Random Number generator seed

**Value**

The proportion of regression p-values that are less than the ANOVA p-values is printed

**Author(s)**

J.H. Maindonald

**Examples**

```
simulateLinear()
```

---

simulateSampDist	<i>Simulated sampling distribution of mean or other statistic</i>
------------------	---

---

**Description**

Simulates the sample distribution of the specified statistic, for samples of the size(s) specified in numINSamp. Additionally a with replacement sample is drawn from the specified population.

**Usage**

```
simulateSampDist(rpop = rnorm, numsamp = 100, numINSamp = c(4, 16),
                 FUN = mean, seed=NULL
                )
```

**Arguments**

rpop	Either a function that generates random samples from the specified distribution, or a vector of values that define the population (i.e., an empirical distribution)
numsamp	Number of samples that should be taken. For close approximation of the asymptotic distribution (e.g., for the mean) this number should be large
numINSamp	Size(s) of each of the numsamp sample(s)
FUN	Function to calculate the statistic whose sampling distribution is to be simulated
seed	Optional seed for random number generation

**Value**

List, with elements values, numINSamp and FUN

values	Matrix, with dimensions numsamp by numINSamp + 1. The first column has a random with replacement sample from the population, while the remaining length(numINSamp) columns hold simulated values from sampling distributions with samples of the specified size(s)
numINSamp	Input value of numINSamp
numsamp	Input value of numsamp

**Author(s)**

John Maindonald

**References**

Maindonald, J.H. and Braun, W.J. (3rd edn, 2010) *Data Analysis and Graphics Using R*, 3rd edn, Sections 3.3 and 3.4

**See Also**

help(plotSampDist)

**Examples**

```
## By default, sample from normal population
simAvs <- simulateSampDist()
par(pty="s")
plotSampDist(simAvs)
## Sample from empirical distribution
simAvs <- simulateSampDist(rpop=rivers)
plotSampDist(simAvs)

## The function is currently defined as
function(rpop=rnorm, numsamp=100, numINSamp=c(4,16), FUN=mean,
seed=NULL){
  if(!is.null(seed))set.seed(seed)
  funtxt <- deparse(substitute(FUN))
```



```

nDists <- length(numINSamp)+1
values <- matrix(0, nrow=numsamp, ncol=nDists)
if(!is.function(rpop)) {
  x <- rpop
  rpop <- function(n)sample(x, n, replace=TRUE)
}
values[,1] <- rpop(numsamp)
for(j in 2:nDists){
  n <- numINSamp[j-1]
  for(i in 1:numsamp)values[i, j] <- FUN(rpop(n))
}
colnames(values) <- paste("Size", c(1, numINSamp))
invisible(list(values=values, numINSamp=numINSamp, FUN=funtxt))
}

```

socsupport

*Social Support Data***Description**

Data from a survey on social and other kinds of support.

**Usage**

```
socsupport
```

**Format**

This data frame contains the following columns:

**gender** a factor with levels female, male

**age** age, in years, with levels 18-20, 21-24, 25-30, 31-40, 40+

**country** a factor with levels australia, other

**marital** a factor with levels married, other, single

**livewith** a factor with levels alone, friends, other, parents, partner, residences

**employment** a factor with levels employed fulltime, employed part-time, govt assistance, other, parental support

**firstyr** a factor with levels first year, other

**enrolment** a factor with levels full-time, part-time, <NA>

**emotional** summary of 5 questions on emotional support availability

**emotionalsat** summary of 5 questions on emotional support satisfaction

**tangible** summary of 4 questions on availability of tangible support

**tangiblesat** summary of 4 questions on satisfaction with tangible support

**affect** summary of 3 questions on availability of affectionate support sources

**affectsat** summary of 3 questions on satisfaction with affectionate support sources

**psi** summary of 3 questions on availability of positive social interaction  
**psisat** summary of 3 questions on satisfaction with positive social interaction  
**esupport** summary of 4 questions on extent of emotional support sources  
**psupport** summary of 4 questions on extent of practical support sources  
**supsources** summary of 4 questions on extent of social support sources (formerly, socsupport)  
**BDI** Score on the Beck depression index (summary of 21 questions)

### Source

Melissa Manning, Psychology, Australian National University

### Examples

```
attach(socsupport)

not.na <- apply(socsupport[,9:19], 1, function(x)!any(is.na(x)))
ss.pr1 <- princomp(as.matrix(socsupport[not.na, 9:19]), cor=TRUE)
pairs(ss.pr1$scores[,1:3])
sort(-ss.pr1$scores[,1])      # Minus the largest value appears first
pause()

not.na[36] <- FALSE
ss.pr <- princomp(as.matrix(socsupport[not.na, 9:19]), cor=TRUE)
summary(ss.pr)               # Examine the contribution of the components
pause()

# We now regress BDI on the first six principal components:
ss.lm <- lm(BDI[not.na] ~ ss.pr$scores[, 1:6], data=socsupport)
summary(ss.lm)$coef
pause()

ss.pr$loadings[,1]
plot(BDI[not.na] ~ ss.pr$scores[, 1], col=as.numeric(gender),
     pch=as.numeric(gender), xlab="1st principal component", ylab="BDI")
topleft <- par()$usr[c(1,4)]
legend(topleft[1], topleft[2], col=1:2, pch=1:2, legend=levels(gender))
```

---

softbacks

*Measurements on a Selection of Paperback Books*

---

### Description

This is a subset of the allbacks data frame which gives measurements on the volume and weight of 8 paperback books.

### Usage

softbacks

**Format**

This data frame contains the following columns:

**volume** a numeric vector giving the book volumes in cubic centimeters

**weight** a numeric vector giving the weights in grams

**Source**

The bookshelf of J. H. Maindonald.

**Examples**

```
print("Outliers in Simple Regression - Example 5.2")
paperback.lm <- lm(weight ~ volume, data=softbacks)
summary(paperback.lm)
plot(paperback.lm)
```

---

sorption

*sorption data set*


---

**Description**

Concentration-time measurements on different varieties of apples under methyl bromide injection.

**Usage**

```
data(sorption)
```

**Format**

A data frame with 192 observations on the following 14 variables.

**m5** a numeric vector

**m10** a numeric vector

**m30** a numeric vector

**m60** a numeric vector

**m90** a numeric vector

**m120** a numeric vector

**ct** concentration-time

**Cultivar** a factor with levels Pacific Rose BRAEBURN Fuji GRANNY Gala ROYAL Red Delicious Splendour

**Dose** injected dose of methyl bromide

**rep** replicate number, within Cultivar and year

**year** a factor with levels 1988 1989 1998 1999

**year.rep** a factor with levels 1988:1 1988:2 1988:3 1989:1 1989:2 1998:1 1998:2 1998:3 1999:1 1999:2

**gp** a factor with levels BRAEBURN1 BRAEBURN2 Fuji1 Fuji10 Fuji2 Fuji6 Fuji7 Fuji8 Fuji9 GRANNY1 GRANNY2 Gala4 Gala5 Pacific Rose10 Pacific Rose6 Pacific Rose7 Pacific Rose8 Pacific Rose9 ROYAL1 ROYAL2 Red Del10 Red Del9 Red Delicious1 Red Delicious2 Red Delicious3 Red Delicious4 Red Delicious5 Red Delicious6 Red Delicious7 Red Delicious8 Splendour4 Splendour5

**inyear** a factor with levels 1 2 3 4 5 6

---

SP500close	<i>Closing Numbers for S and P 500 Index</i>
------------	--

---

**Description**

Closing numbers for S and P 500 Index, Jan. 1, 1990 through early 2000.

**Usage**

SP500close

**Source**

Derived from SP500 in the MASS library.

**Examples**

ts.plot(SP500close)

---

SP500W90	<i>Closing Numbers for S and P 500 Index - First 100 Days of 1990</i>
----------	---

---

**Description**

Closing numbers for S and P 500 Index, Jan. 1, 1990 through early 2000.

**Usage**

SP500W90

**Source**

Derived from SP500 in the MASS library.

**Examples**

ts.plot(SP500W90)

---

spam7*Spam E-mail Data*

---

**Description**

The data consist of 4601 email items, of which 1813 items were identified as spam.

**Usage**

spam7

**Format**

This data frame contains the following columns:

**crl.tot** total length of words in capitals

**dollar** number of occurrences of the \\$ symbol

**bang** number of occurrences of the ! symbol

**money** number of occurrences of the word 'money'

**n000** number of occurrences of the string '000'

**make** number of occurrences of the word 'make'

**yesno** outcome variable, a factor with levels n not spam, y spam

**Source**

George Forman, Hewlett-Packard Laboratories

These data are available from the University of California at Irvine Repository of Machine Learning Databases and Domain Theories. The address is: <http://www.ics.uci.edu/~Here>

**Examples**

```
require(rpart)
spam.rpart <- rpart(formula = yesno ~ crl.tot + dollar + bang +
  money + n000 + make, data=spam7)
plot(spam.rpart)
text(spam.rpart)
```

---

stVincent	<i>Averages by block of yields for the St. Vincent Corn data</i>
-----------	--

---

**Description**

These data frames have yield averages by blocks (parcels).

**Usage**

stVincent

**Format**

A data frame with 324 observations on 8 variables.

- code** a numeric vector
- island** a numeric vector
- id** a numeric vector
- site** a factor with 8 levels.
- block** a factor with levels I II III IV
- plot** a numeric vector
- trt** a factor consisting of 12 levels
- harvwt** a numeric vector; the average yield

**Source**

Andrews DF; Herzberg AM, 1985. Data. A Collection of Problems from Many Fields for the Student and Research Worker. Springer-Verlag. (pp. 339-353)

---

sugar	<i>Sugar Data</i>
-------	-------------------

---

**Description**

The sugar data frame has 12 rows and 2 columns. They are from an experiment that compared an unmodified wild type plant with three different genetically modified forms. The measurements are weights of sugar that were obtained by breaking down the cellulose.

**Usage**

sugar

**Format**

This data frame contains the following columns:

**weight** weight, in mg

**trt** a factor with levels Control i.e. unmodified Wild form, A Modified 1, B Modified 2, C Modified 3

**Source**

Anonymous

**Examples**

```
sugar.aov <- aov(weight ~ trt, data=sugar)
fitted.values(sugar.aov)
summary.lm(sugar.aov)
sugar.aov <- aov(formula = weight ~ trt, data = sugar)
summary.lm(sugar.aov)
```

---

tinting

---

*Car Window Tinting Experiment Data*


---

**Description**

These data are from an experiment that aimed to model the effects of the tinting of car windows on visual performance. The authors were mainly interested in effects on side window vision, and hence in visual recognition tasks that would be performed when looking through side windows.

**Usage**

```
tinting
```

**Format**

This data frame contains the following columns:

**case** observation number

**id** subject identifier code (1-26)

**age** age (in years)

**sex** a factor with levels f female, m male

**tint** an ordered factor with levels representing degree of tinting: no < lo < hi

**target** a factor with levels locon: low contrast, hicon: high contrast

**it** the inspection time, the time required to perform a simple discrimination task (in milliseconds)

**csoa** critical stimulus onset asynchrony, the time to recognize an alphanumeric target (in milliseconds)

**agegp** a factor with levels younger, 21-27, older, 70-78

## Details

Visual light transmittance (VLT) levels were 100% (tint=none), 81.3% (tint=lo), and 35.1% (tint=hi). Based on these and other data, Burns et al. argue that road safety may be compromised if the front side windows of cars are tinted to 35

## Source

Burns, N.R., Nettlebeck, T., White, M. and Willson, J., 1999. Effects of car window tinting on visual performance: a comparison of younger and older drivers. *Ergonomics* 42: 428-443.

## Examples

```
levels(tinting$agegp) <- capstring(levels(tinting$agegp))
xyplot(csoa ~ it | sex * agegp, data=tinting) # Simple use of xyplot()
pause()

xyplot(csoa ~ it|sex*agegp, data=tinting, panel=panel.superpose, groups=target)
pause()

xyplot(csoa ~ it|sex*agegp, data=tinting, panel=panel.superpose, col=1:2,
       groups=target, key=list(x=0.14, y=0.84, points=list(pch=rep(1,2),
        col=1:2), text=list(levels(tinting$target), col=1:2), border=TRUE))
pause()

xyplot(csoa ~ it|sex*agegp, data=tinting, panel=panel.superpose,
       groups=tint, type=c("p","smooth"), span=0.8, col=1:3,
       key=list(x=0.14, y=0.84, points=list(pch=rep(1,2), col=1:3),
        text=list(levels(tinting$tint), col=1:3), border=TRUE))
```

---

tomato

*Root weights of tomato plants exposed to 4 different treatments*

---

## Description

The tomato data frame has 24 rows and 2 columns. They are from an experiment that exposed tomato plants to four different 'nutrients'.

## Usage

```
data(tomato)
```

## Format

This data frame contains the following columns:

**weight** weight, in g

**trt** a factor with levels water only, conc nutrient, 2-4-D + conc nutrient, 3x conc nutrient



**Source**

Dr Ron Balham, Victoria University of Wellington NZ, sometime in 1971 - 1976.

**Examples**

```
tomato.aov <- aov(log(weight) ~ trt, data=tomato)
fitted.values(tomato.aov)
summary.lm(tomato.aov)
tomato.aov <- aov(formula = weight ~ trt, data = tomato)
summary.lm(tomato.aov)
```

---

toycars

*Toy Cars Data*

---

**Description**

The toycars data frame has 27 rows and 3 columns. Observations are on the distance traveled by one of three different toy cars on a smooth surface, starting from rest at the top of a 16 inch long ramp tilted at varying angles.

**Usage**

```
toycars
```

**Format**

This data frame contains the following columns:

**angle** tilt of ramp, in degrees

**distance** distance traveled, in meters

**car** a numeric code (1 = first car, 2 = second car, 3 = third car)

**Examples**

```
toycars.lm <- lm(distance ~ angle + factor(car), data=toycars)
summary(toycars.lm)
```

two65

*Unpaired Heated Elastic Bands***Description**

Twenty-one elastic bands were divided into two groups.

One of the sets was placed in hot water (60-65 degrees C) for four minutes, while the other was left at ambient temperature. After a wait of about ten minutes, the amounts of stretch, under a 1.35 kg weight, were recorded.

**Usage**

```
pair65
```

**Format**

This list contains the following elements:

**heated** a numeric vector giving the stretch lengths for the heated bands

**ambient** a numeric vector giving the stretch lengths for the unheated bands

**Source**

J.H. Maindonald

**Examples**

```
twot.permutation(two65$ambient,two65$heated) # two sample permutation test
```

twot.permutation

*Two Sample Permutation Test - Obsolete***Description**

This function computes the p-value for the two sample t-test using a permutation test. The permutation density can also be plotted.

**Usage**

```
twot.permutation(x1=two65$ambient, x2=two65$heated, nsim=2000, plotit=TRUE)
```

**Arguments**

x1	Sample 1
x2	Sample 2
nsim	Number of simulations
plotit	If TRUE, the permutation density will be plotted

**Details**

Suppose we have  $n_1$  values in one group and  $n_2$  in a second, with  $n = n_1 + n_2$ . The permutation distribution results from taking all possible samples of  $n_2$  values from the total of  $n$  values.

**Value**

The p-value for the test of the hypothesis that the mean of  $x_1$  differs from  $x_2$

**Author(s)**

J.H. Maindonald

**References**

Good, P. 2000. Permutation Tests. Springer, New York.

**Examples**

```
twot.permutation()
```

---

twotPermutation	<i>Two Sample Permutation Test</i>
-----------------	------------------------------------

---

**Description**

This function computes the p-value for the two sample t-test using a permutation test. The permutation density can also be plotted.

**Usage**

```
twotPermutation(x1=two65$ambient, x2=two65$heated, nsim=2000, plotit=TRUE)
```

**Arguments**

x1	Sample 1
x2	Sample 2
nsim	Number of simulations
plotit	If TRUE, the permutation density will be plotted

**Details**

Suppose we have  $n_1$  values in one group and  $n_2$  in a second, with  $n = n_1 + n_2$ . The permutation distribution results from taking all possible samples of  $n_2$  values from the total of  $n$  values.

**Value**

The p-value for the test of the hypothesis that the mean of  $x_1$  differs from  $x_2$

**Author(s)**

J.H. Maindonald

**References**

Good, P. 2000. Permutation Tests. Springer, New York.

**Examples**

```
twotPermutation()
```

---

vif

*Variance Inflation Factors*

---

**Description**

Variance inflation factors are computed for the standard errors of linear model coefficient estimates.

**Usage**

```
vif(obj, digits=5)
```

**Arguments**

obj	A lm object
digits	Number of digits

**Value**

A vector of variance inflation factors corresponding to the coefficient estimates given in the lm object.

**Author(s)**

J.H. Maindonald

**See Also**

lm

**Examples**

```
litters.lm <- lm(brainwt ~ bodywt + lsize, data = litters)
vif(litters.lm)

carprice1.lm <- lm(gpm100 ~ Type+Min.Price+Price+Max.Price+Range.Price,
  data=carprice)
vif(carprice1.lm)

carprice.lm <- lm(gpm100 ~ Type + Price, data = carprice)
vif(carprice.lm)
```

vince111b

*Averages by block of corn yields, for treatment 111 only***Description**

These data frames have averages by blocks (parcels) for the treatment 111.

**Usage**

```
vince111b
```

**Format**

A data frame with 36 observations on 8 variables.

**site** a factor with levels AGSV CASV CPSV LPSV MPSV OOSV OTSV SSSV UISV

**parcel** a factor with levels I II III IV

**code** a numeric vector

**island** a numeric vector

**id** a numeric vector

**plot** a numeric vector

**trt** a numeric vector

**harvwt** a numeric vector

**Source**

Andrews DF; Herzberg AM, 1985. Data. A Collection of Problems from Many Fields for the Student and Research Worker. Springer-Verlag. (pp. 339-353)

vlt

*Video Lottery Terminal Data***Description**

Data on objects appearing in three windows on a video lottery terminal, together with the prize payout (usually 0). Observations were taken on two successive days in late 1994 at a hotel lounge north of Winnipeg, Manitoba. Each observation cost 25 cents (Canadian). The game played was 'Double Diamond'.

**Usage**

vlt

**Format**

This data frame contains the following columns:

**window1** object appearing in the first window.

**window2** object appearing in the second window.

**window3** object appearing in the third window.

**prize** cash prize awarded (in Canadian dollars).

**night** 1, if observation was taken on day 1; 2, if observation was taken on day 2.

**Details**

At each play, each of three windows shows one of 7 possible objects. Apparently, the three windows are independent of each other, and the objects should appear with equal probability across the three windows. The objects are coded as follows: blank (0), single bar (1), double bar (2), triple bar (3), double diamond (5), cherries (6), and the numeral "7" (7).

Prizes (in quarters) are awarded according to the following scheme: 800 (5-5-5), 80 (7-7-7), 40 (3-3-3), 25 (2-2-2), 10 (1-1-1), 10 (6-6-6), 5 (2 "6"s), 2 (1 "6") and 5 (any combination of "1", "2" and "3"). In addition, a "5" doubles any winning combination, e.g. (5-3-3) pays 80 and (5-3-5) pays 160.

**Source**

Braun, W. J. (1995) An illustration of bootstrapping using video lottery terminal data. Journal of Statistics Education <http://www.amstat.org/publications/jse/v3n2/datasets.braun.html>

**Examples**

```
vlt.stk <- stack(vlt[,1:3])
table(vlt.stk)
```

wages1833

*Wages of Lancashire Cotton Factory Workers in 1833***Description**

The wages1833 data frame gives the wages of Lancashire cotton factory workers in 1833.

**Usage**

```
wages1833
```

**Format**

This data frame contains the following columns:

**age** age in years

**mnum** number of male workers

**mwage** average wage of male workers

**fnum** number of female workers

**fwage** average wage of female workers

**Source**

Boot, H.M. 1995. How Skilled Were the Lancashire Cotton Factory Workers in 1833? *Economic History Review* 48: 283-303.

**Examples**

```
attach(wages1833)
plot(mwage~age,ylim=range(c(mwage, fwage[fwage>0])))
points(fwage[fwage>0]~age[fwage>0],pch=15,col="red")
lines(lowess(age,mwage))
lines(lowess(age[fwage>0],fwage[fwage>0]),col="red")
```

whoops

*Deaths from whooping cough, in London***Description**

Deaths from whooping cough, in London from 1740 to 1881.

**Usage**

```
data(whoops)
```

**Format**

This is a multiple time series consisting of 3 series: wcough, ratio, and alldeaths.

**Source**

Guy, W. A. 1882. Two hundred and fifty years of small pox in London. Journal of the Royal Statistical Society 399-443.

**References**

Lancaster, H. O. 1990. Expectations of Life. Springer.

**Examples**

```
data(whoops)
str(whoops)
plot(whoops)
```

---

worldRecords

---

*Record times for track and road races, at August 9th 2006*


---

**Description**

Record times for track and road races, at August 9th 2006

**Usage**

```
data(worldRecords)
```

**Format**

A data frame with 40 observations on the following 9 variables.

Distance distance in kilometers

roadOrtrack a factor with levels road track

Place place; a character vector

Time time in minutes

Date a Date

**Details**

For further details, and some additional details, see the web site that is the source of the data.

**Source**

<http://www.gbrathletics.com/wrec.htm>



**Examples**

```
data(worldRecords)
xyplot(log(Time) ~ log(Distance), groups=roadORtrack, data=worldRecords)
xyplot(log(Time) ~ log(Distance), groups=roadORtrack, data=worldRecords,
       type=c("p", "r"))
xyplot(log(Time) ~ log(Distance), groups=roadORtrack, data=worldRecords,
       type=c("p", "smooth"))
```

---

zzDAAGxdb

---

*List, each of whose elements hold rows of a file, in character format*


---

**Description**

This is the default alternative database for use with the function `datafile`, which uses elements of this list to place files in the working directory. The names of the list elements are `bestTimes` and `bostonc`.

**Usage**

```
data(zzDAAGxdb)
```

**Format**

Successive elements in this list hold character vectors from which the corresponding files can be readily generated.

**Details**

The web site given as the source of the data has additional information on the `bestTimes` data. Records are as at August 7 2006.

**Source**

<http://www.gbrathletics.com/wrec.htm> (`bestTimes`)

<http://lib.stat.cmu.edu/datasets/> (`bostonc`)

**References**

Harrison, D. and Rubinfeld, D.L. 'Hedonic prices and the demand for clean air', J. Environ. Economics & Management, vol.5, 81-102, 1978. corrected by Kelley Pace ([kpace@unix1.sncc.lsu.edu](mailto:kpace@unix1.sncc.lsu.edu))

**Examples**

```
data(zzDAAGxdb)
names(zzDAAGxdb)
```

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