# Package 'PASWR'

October 12, 2022

Type Package
Title Probability and Statistics with R
Version 1.3
Date 2022-05-14
Maintainer Alan T. Arnholt <arnholtat@appstate.edu></arnholtat@appstate.edu>
<b>Description</b> Functions and data sets for the text Probability and Statistics with R.
Depends lattice
Imports MASS, e1071
LazyData TRUE
License GPL-2
RoxygenNote 7.1.2
Encoding UTF-8
NeedsCompilation no
Author Alan T. Arnholt [aut, cre]
Repository CRAN
<b>Date/Publication</b> 2022-05-15 11:50:02 UTC
<b>Date:</b> 1 abiteuron 2022 00 13 11130.02 010
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PASWR-package

Probability and Statistics with R

## Description

Data and functions for the book Probability and Statistics with R

#### **Details**

Package: PASWR Type: Package Version: 1.2

Date: 2016-02-24 License: GPL (>=2)

Comprehensive and engineering-oriented, *Probability and Statistics with R* provides a thorough treatment of probability and statistics, clear and accessible real-world examples, and fully detailed proofs. The text provides step-by-step explanations for numerous examples in R and S-PLUS for nearly every topic covered, including both traditional and nonparametric techniques. With a wide range of graphs to illustrate complex material as well as a solutions manual, the book also offers an accompanying website that features supporting information, including datasets, functions, and other downloadable material. It is ideal for undergraduate students and for engineers and scientists who must perform statistical analyses.

#### Author(s)

Alan T. Arnholt

Maintainer: <arnholtat@appstate.edu>

# References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

Aggression

TV and Behavior

# **Description**

Data regarding aggressive behavior in relation to exposure to violent television programs used in Example 10.5

Apple 5

#### **Format**

A data frame with 16 observations on the following 2 variables:

- violence (an integer vector)
- noviolence (an integer vector)

#### **Details**

This is data regarding aggressive behavior in relation to exposure to violent television programs from Gibbons (1997) with the following exposition:

... a group of children are matched as well as possible as regards home environment, genetic factors, intelligence, parental attitudes, and so forth, in an effort to minimize factors other than TV that might influence a tendency for aggressive behavior. In each of the resulting 16 pairs, one child is randomly selected to view the most violent shows on TV, while the other watches cartoons, situation comedies, and the like. The children are then subjected to a series of tests designed to produce an ordinal measure of their aggression factors. (pages 143-144)

#### Source

Gibbons, J. D. (1997) Nonparametric Methods for Quantitative Analysis. American Sciences Press.

#### References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

## **Examples**

```
with(data = Aggression,
wilcox.test(violence, noviolence, paired = TRUE,
alternative = "greater"))
```

Apple

Apple Hardness

## **Description**

An experiment was undertaken where seventeen recently picked (Fresh) apples were randomly selected and measured for hardness. Seventeen apples were also randomly selected from a warehouse (Warehouse) where the apples had been stored for one week. Data are used in Example 8.10.

#### **Format**

A data frame with 17 observations on the following 2 variables: #'

- Fresh (hardness rating measured in kg/meter<sup>2</sup>)
- Warehouse (hardness rating measured in kg/meter<sup>2</sup>)

6 AptSize

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

## **Examples**

```
# Figure 8.5
attach(Apple)
par(pty = "s")
Altblue <- "#A9E2FF"
Adkblue <- "#0080FF"
fresh <- qqnorm(Fresh)</pre>
old <- qqnorm(Warehouse)</pre>
plot(fresh, type = "n",ylab = "Sample Quantiles", xlab = "Theoretical Quantiles")
qqline(Fresh, col = Altblue)
qqline(Warehouse, col = Adkblue)
points(fresh, col = Altblue, pch = 16, cex = 1.2)
points(old, col = Adkblue, pch = 17)
legend(-1.75, 9.45, c("Fresh", "Warehouse"), col = c(Altblue, Adkblue),
text.col = c("black","black"), pch = c(16, 17), lty = c(1, 1), bg = "gray95", cex = 0.75)
title("Q-Q Normal Plots")
detach(Apple)
# Trellis approach
qqmath(\sim c(Fresh, Warehouse), type = c("p","r"), pch = c(16, 17),
cex = 1.2, col=c("#A9E2FF", "#0080FF"),
groups=rep(c("Fresh", "Warehouse"), c(length(Fresh), length(Warehouse))),
data = Apple, ylab = "Sample Quantiles", xlab = "Theoretical Quantiles")
```

AptSize

Apartment Size

# Description

Size of apartments in Mendebaldea, Spain and San Jorge, Spain

## **Format**

A data frame with 8 observations on the following 2 variables:

Mendebaldea Mendebaldea apartment size in square meters

**SanJorge** San Jorge apartment size in square meters

#### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R*. Chapman & Hall/CRC.

Baberuth 7

## **Examples**

```
with(data = AptSize,
boxplot(Mendebaldea, SanJorge) )
```

Baberuth

George Herman Ruth

# **Description**

Baseball statistics for George Herman Ruth (The Bambino or The Sultan Of Swat)

#### **Format**

A data frame with 22 observations on the following 14 variables.

Year year in which the season occurred

Team team he played for Bos-A, Bos-N, or NY-A

G games played

AB at bats

R runs scored

H hits

X2B doubles

X3B triples

HR home runs

RBI runs batted in

SB stolen bases

BB base on balls or walks

BA batting average H/AB

**SLG** slugging percentage (total bases/at bats)

#### **Source**

```
https://www.baseball-reference.com/about/bat_glossary.shtml
```

## References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
with(data = Baberuth,
hist(RBI))
```

8 Battery

Bac

Blood Alcohol Content

# **Description**

Two volunteers each consumed a twelve ounce beer every fifteen minutes for one hour. One hour after the fourth beer was consumed, each volunteer's blood alcohol was measured with a different breathalyzer from the same company. The numbers recorded in data frame Bac are the sorted blood alcohol content values reported with breathalyzers from company X and company Y. Data are used in Example 9.15.

#### **Format**

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

- X blood alcohol content measured in g/L
- Y blood alcohol content measured in g/L

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

# **Examples**

```
with(data = Bac,
var.test(X, Y, alternative = "less"))
```

Battery

Lithium Batteries

#### **Description**

A manufacturer of lithium batteries has two production facilities, A and B. Fifty randomly selected batteries with an advertised life of 180 hours are selected, and tested. The lifetimes are stored in (facilityA). Fifty ramdomly selected batteries with an advertised life of 200 hours are selected, and tested. The lifetimes are stored in (facilityB).

## **Format**

A data frame with 50 observations on the following 2 variables:

facilityA life time measured in hours

facilityB life time measured in hours

bino.gen 9

## **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

# **Examples**

```
with(data = Battery,
qqnorm(facilityA))
with(data = Battery,
qqline(facilityA))
```

bino.gen

Simulating Binomial Distribution

# Description

Function that generates and displays *m* repeated samples of *n* Bernoulli trials with a given probability of success.

# Usage

```
bino.gen(samples, n, pi)
```

# Arguments

samples number of repeated samples to generate

n number of Bernoulli trials

pi probability of success for Bernoulli trial

# Value

simulated.distribution

Simulated binomial distribution

 $theoretical. \\ distribution$ 

Theoretical binomial distribution

## Author(s)

Alan T. Arnholt

```
bino.gen(1000, 20, 0.75)
```

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biomass

Beech Trees

## **Description**

Several measurements of 42 beech trees (Fagus Sylvatica) taken from a forest in Navarra (Spain).

#### **Format**

A data frame with 42 observations on the following 4 variables:

**Dn** diameter of the stem in centimeters

H height of the tree in meters

PST weight of the stem in kilograms

**PSA** aboveground weight in kilograms

#### **Source**

Gobierno de Navarra and Gestion Ambiental Viveros y Repoblaciones de Navarra, 2006. The data were obtained within the European Project FORSEE.

## References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

## **Examples**

```
plot(log(PSA) \sim log(Dn), data = biomass)
```

Bodyfat

**Body Fat Composition** 

# Description

Values from a study reported in the *American Journal of Clinical Nutrition* that investigated a new method for measuring body composition

## **Format**

A data frame with 18 observations on the following 3 variables:

age age in years

fat body fat composition

sex a factor with levels F for female and M for male

Calculus 11

## Source

Mazess, R. B., Peppler, W. W., and Gibbons, M. (1984) *Total Body Composition by Dual-Photon* (153 Gd) Absorptiometry. American Journal of Clinical Nutrition, **40**, **4**: 834-839.

## References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

# **Examples**

```
boxplot(fat ~ sex, data = Bodyfat)
```

Calculus

Calculus Assessment Scores

# **Description**

Mathematical assessment scores for 36 students enrolled in a biostatistics course according to whether or not the students had successfully completed a calculus course prior to enrolling in the biostatistics course

#### **Format**

A data frame with 18 observations on the following 2 variables:

No.Calculus assessment score for students with no prior calculus

Yes.Calculus assessment score for students with prior calculus

# Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
with(data = Calculus,
z.test(x = Yes.Calculus, y = No.Calculus, sigma.x = 5, sigma.y = 12)$conf
)
```

12 checking.plots

Cars2004EU

Cars in the European Union (2004)

# Description

The numbers of cars per 1000 inhabitants (cars), the total number of known mortal accidents (deaths), and the country population/1000 (population) for the 25 member countries of the European Union for the year 2004

## **Format**

A data frame with 25 observations on the following 4 variables:

country a factor with levels Austria, Belgium, Cyprus, Czech Republic, Denmark, Estonia, Finland, France, Germany, Greece, Hungary, Ireland, Italy, Latvia, Lithuania, Luxembourg, Malta, Netherlands, Poland, Portugal, Slovakia, Slovenia, Spain, Sweden, and United Kingdom

cars numbers of cars per 1000 inhabitants

deaths total number of known mortal accidents

population country population/1000

## Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

# **Examples**

```
plot(deaths ~ cars, data = Cars2004EU)
```

checking.plots

Checking Plots

## **Description**

Function that creates four graphs that can be used to help assess independence, normality, and constant variance

# Usage

```
checking.plots(model, n.id = 3, COL = c("#0080FF", "#A9E2FF"))
```

Chips 13

# **Arguments**

model an aov or lm object
---------------------------

n.id the number of points to identify

COL vector of two colors

# Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

#### See Also

```
twoway.plots, oneway.plots
```

# **Examples**

```
mod.aov <- aov(StopDist ~ tire, data = Tire)
checking.plots(mod.aov)
rm(mod.aov)</pre>
```

Chips

Silicon Chips

# **Description**

Two techniques of splitting chips are randomly assigned to 28 sheets so that each technique is applied to 14 sheets. The values recorded in Chips are the number of usable chips from each silicon sheet.

# **Format**

A data frame with 14 observations on the following 2 variables:

```
techniqueI number of usable chips techniqueII number of usable chips
```

## **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

14 CircuitDesigns

## **Examples**

```
par(mfrow = c(1, 2))
with(data = Chips, qqnorm(techniqueI))
with(data = Chips, qqline(techniqueII))
with(data = Chips, qqline(techniqueII))
with(data = Chips, qqline(techniqueII))
par(mfrow=c(1, 1))
# Trellis Approach
graph1 <- qqmath(~techniqueI, data = Chips, type=c("p", "r"))
graph2 <- qqmath(~techniqueII, data = Chips, type=c("p", "r"))
print(graph1, split=c(1, 1, 2, 1), more = TRUE)
print(graph2, split=c(2, 1, 2, 1), more = FALSE)
rm(graph1, graph2)</pre>
```

CircuitDesigns

Circuit Design Lifetime

## Description

CircuitDesigns contains the results from an accelerated life test used to estimate the lifetime of four different circuit designs (lifetimes in thousands of hours).

#### Format

A data frame with 26 observations on the following 2 variables:

lifetime lifetimes in thousands of hours

design a factor with levels Design1, Design2, Design3, and Design4

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
bwplot(design ~ lifetime, data = CircuitDesigns)
```

CIsim 15

CIsim

Confidence Interval Simulation Program

## **Description**

This program simulates random samples from which it constructs confidence intervals for either the population mean, the population variance, or the population proportion of successes.

# Usage

```
CIsim(
  samples = 100,
  n = 30,
  parameter = 0.5,
  sigma = 1,
  conf.level = 0.95,
  type = c("Mean", "Var", "Pi")
)
```

# **Arguments**

samples the number of samples desired. the size of each sample parameter If constructing confidence intervals for the population mean or the population variance, parameter is the population mean (i.e., type is one of either "Mean" or "Var"). If constructing confidence intervals for the population proportion of successes, the value entered for parameter represents the population proportion of successes (Pi), and as such, must be a number between 0 and 1. is the population standard deviation. sigma is not required if confidence intersigma vals are of type "Pi". conf.level confidence level for the graphed confidence intervals, restricted to lie between zero and one character string, one of "Mean", "Var", or "Pi", or just the initial letter of each, type indicating the type of confidence interval simulation to perform

#### **Details**

Default is to construct confidence intervals for the population mean. Simulated confidence intervals for the population variance or population proportion of successes are possible by selecting the appropriate value in the type argument.

#### Value

Performs specified simulation and draws the resulting confidence intervals on a graphical device.

16 Combinations

## Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

## **Examples**

```
CIsim(samples = 100, n = 30, parameter = 100, sigma = 10, conf.level = 0.90)
# Simulates 100 samples of size 30 from a normal distribution with mean 100
# and a standard deviation of 10. From the 100 simulated samples, 90% confidence
# intervals for the Mean are constructed and depicted in the graph.

CIsim(100, 30, 100, 10, type = "Var")
# Simulates 100 sample of size 30 from a normal distribution with mean 100
# and a standard deviation of 10. From the 100 simulated samples, 95% confidence
# intervals for the variance are constructed and depicted in the graph.

CIsim(100, 50, 0.5, type = "Pi", conf.level = 0.92)
# Simulates 100 samples of size 50 from a binomial distribution where the
# population proportion of successes is 0.5. From the 100 simulated samples,
# 92% confidence intervals for Pi are constructed and depicted in the graph.
```

Combinations

**Combinations** 

# **Description**

Computes all possible combinations of n objects taken k at a time.

# Usage

```
Combinations(n, k)
```

## **Arguments**

n a number

k a number less than or equal to n

## Value

Returns a matrix containing the possible combinations of n objects taken k at a time.

#### See Also

SRS

CosAma 17

## **Examples**

```
Combinations(5,2)
# The columns in the matrix list the values of the 10 possible
# combinations of 5 things taken 2 at a time.
```

CosAma

Cosmed Versus Amatek

# **Description**

The Cosmed is a portable metabolic system. A study at Appalachian State University compared the metabolic values obtained from the Cosmed to those of a reference unit (Amatek) over a range of workloads from easy to maximal to test the validity and reliability of the Cosmed. A small portion of the results for VO2 (ml/kg/min) measurements taken at a 150 watt workload are stored in CosAma.

#### **Format**

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

subject subject number

Cosmed measured VO2 with Cosmed

Amatek measured VO2 with Amatek

## Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
bwplot(~(Cosmed - Amatek), data = CosAma)
```

Depend Depend

Cows

Butterfat of Cows

## **Description**

Random samples of ten mature (five-year-old and older) and ten two-year-old cows were taken from each of five breeds. The average butterfat percentage of these 100 cows is stored in the variable butterfat with the type of cow stored in the variable breed and the age of the cow stored in the variable age.

## **Format**

A data frame with 100 observations on the following 3 variables:

butterfat average butterfat percentage

age a factor with levels 2 years old and Mature

breed a factor with levels Ayrshire, Canadian, Guernsey, Holstein-Friesian, and Jersey

#### Source

Canadian record book of purebreed diary cattle.

## References

Sokal, R. R. and Rohlf, F. J. (1994) Biometry. W. H. Freeman, New York, third edition.

# **Examples**

```
summary(aov(butterfat ~ breed + age, data = Cows))
```

Depend

Number of Dependent Children for 50 Families

# Description

Number of dependent children for 50 families.

## **Format**

A data frame with 50 observations on the following 4 variables.

C1 a numeric vector

number a numeric vectorCount a numeric vectorPercent a numeric vector

Drosophila 19

## **Source**

Kitchens, L. J. (2003) Basic Statistics and Data Analysis. Duxbury

# **Examples**

```
with(data = Depend, table(C1))
```

Drosophila

Drosophila Melanogaster

# Description

Drosophila contains per diem fecundity (number of eggs laid per female per day for the first 14 days of life) for 25 females from each of three lines of *Drosophila melanogaster*. The three lines are Nonselected (control), Resistant, and Susceptible. Data are used in Example 11.5.

#### **Format**

A data frame with 75 observations on the following 2 variables:

Fecundity number of eggs laid per female per day for the first 14 days of life

Line a factor with levels Nonselected, Resistant, and Susceptible

## **Source**

The original measurements are from an experiment conducted by R. R. Sokal (Sokal and Rohlf, 1994, p. 237).

#### References

Sokal, R. R. and Rohlf, F. J. (1994) Biometry. W. H. Freeman, New York, third edition.

```
summary(aov(Fecundity ~ Line, data = Drosophila))
```

20 EDA

**EDA** 

Exploratory Data Analysis

# Description

Function that produces a histogram, density plot, boxplot, and Q-Q plot

# Usage

```
EDA(x, trim = 0.05, dec = 3)
```

#### **Arguments**

x is a numeric vector where NAs and Infs are allowed but will be removed.

trim is a fraction (between 0 and 0.5, inclusive) of values to be trimmed from each

end of the ordered data such that if trim = 0.5, the result is the median.

dec is a number specifying the number of decimals

## **Details**

The function EDA() will not return console window information on data sets containing more than 5000 observations. It will, however, still produce graphical output for data sets containing more than 5000 observations.

## Value

Function returns various measures of center and location. The values returned for the quartiles are based on the default  $\bf R$  definitions for quartiles. For more information on the definition of the quartiles, type ?quantile and read about the algorithm used by type = 7.

## Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

```
EDA(x = rnorm(100))
# Produces four graphs for the 100 randomly
# generated standard normal variates.
```

Engineer 21

Engineer

Engineer Salaries

## **Description**

Salaries for engineering graduates 10 years after graduation

#### **Format**

A data frame with 51 observations on the following 2 variables:

**salary** salary 10 years after graduation in thousands of dollars **university** one of three different engineering universities

#### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

# **Examples**

```
boxplot(salary ~ university, data = Engineer, horizontal = TRUE)
# Trellis Approach
bwplot(university ~ salary, data = Engineer)
```

**EPIDURAL** 

Traditional Sitting Position Versus Hamstring Stretch Position

# Description

Initial results from a study to determine whether the traditional sitting position or the hamstring stretch position is superior for administering epidural anesthesia to pregnant women in labor as measured by the number of obstructive (needle to bone) contacts (OC)

## **Format**

A data frame with 85 observations on the following 7 variables:

**Doctor** a factor with levels Dr. A, Dr. B, Dr. C, and Dr. D

kg weight in kg of patient

cm height in cm of pateint

**Ease** a factor with levels Difficult, Easy, and Impossible indicating the physician's assessment of how well bone landmarks can be felt in the patient

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**Treatment** a factor with levels Hamstring Stretch and Traditional Sitting

**OC** number of obstructive contacts

**Complications** a factor with levels Failure – person got dizzy, Failure – too many OCs, None, Paresthesia, and Wet Tap

#### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R*. Chapman & Hall/CRC.

# **Examples**

```
EPIDURAL$Teasy <- factor(EPIDURAL$Ease,
levels = c("Easy", "Difficult", "Impossible"))
X <- table(EPIDURAL$Doctor, EPIDURAL$Teasy)
X
par(mfrow = c(2, 2)) # Figure 2.12
barplot(X,
main = "Barplot where Doctor is Stacked \n within Levels of Palpitation")
barplot(t(X),
main = "Barplot where Levels of Palpitation \n is Stacked within Doctor")
barplot(X, beside = TRUE,
main = "Barplot where Doctor is Grouped \n within Levels of Palpitation")
barplot(t(X), beside = TRUE,
main = "Barplot where Levels of Palpitation \n is Grouped within Doctor")
par(mfrow = c(1, 1))
rm(X)</pre>
```

**EPIDURALf** 

Traditional Sitting Position Versus Hamstring Stretch Position

# Description

Intermediate results from a study to determine whether the traditional sitting position or the hamstring stretch position is superior for administering epidural anesthesia to pregnant women in labor as measured by the number of obstructive (needle to bone) contacts (OC)

#### **Format**

A data frame with 342 observations on the following 7 variables:

Doctor a factor with levels Dr. A, Dr. B, Dr. C, and Dr. D

kg weight in kg of patient

cm height in cm of pateint

**Ease** a factor with levels Difficult, Easy, and Impossible indicating the physician's assessment of how well bone landmarks can be felt in the patient

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**Treatment** a factor with levels Hamstring Stretch and Traditional Sitting

**OC** number of obstructive contacts

**Complications** a factor with levels Failure – person got dizzy, Failure – too many OCs, None, Paresthesia, and Wet Tap

#### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

## **Examples**

```
boxplot(OC ~ Treatment, data = EPIDURALf)
```

**EURD** 

European Union Research and Development

# **Description**

A random sample of 15 countries' research and development investments for the years 2002 and 2003 is taken and the results in millions of euros are stored in EURD.

## **Format**

A data frame with 15 observations on the following 3 variables:

**Country** a factor with levels Bulgaria, Croatia, Cyprus, Czech Republic, Estonia, France, Hungary, Latvia, Lithuania, Malta, Portugal, Romania, Slovakia, and Slovenia

RD2002 research and development investments in millions of euros for 2002

RD2003 research and development investments in millions of euros for 2003

# Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
qqmath(\sim(RD2003 - RD2002), data = EURD, type=c("p", "r"))
```

24 FCD

fagus

Retained Carbon in Beech Trees

#### **Description**

The carbon retained by leaves measured in kg/ha is recorded for forty-one different plots of mountainous regions of Navarra (Spain), depending on the forest classification: areas with 90% or more beech trees (*Fagus Sylvatica*) are labeled monospecific, while areas with many species of trees are labeled multispecific.

#### **Format**

A data frame with 41 observations on the following 3 variables:

Plot plot number

carbon carbon retained by leaves measured in kg/ha

type a factor with levels monospecific and multispecific

#### **Source**

Gobierno de Navarra and Gestion Ambiental Viveros y Repoblaciones de Navarra, 2006. The data were obtained within the European Project FORSEE.

#### References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

## **Examples**

boxplot(carbon ~ type, data=fagus)

FCD

Fat Cats

# **Description**

In a weight loss study on obese cats, overweight cats were randomly assigned to one of three groups and boarded in a kennel. In each of the three groups, the cats' total caloric intake was strictly controlled (1 cup of generic cat food) and monitored for 10 days. The difference between the groups was that group A was given 1/4 of a cup of cat food every six hours, group B was given 1/3 a cup of cat food every eight hours, and group C was given 1/2 a cup of cat food every twelve hours. The weight of the cats at the beginning and end of the study was recorded and the difference in weights (grams) is stored in the variable Weight of the data frame FCD. Data are used in Example 11.4.

Fertilize 25

## **Format**

A data frame with 36 observations on the following 2 variables:

```
Weight difference in weights (grams)
Diet a factor with levels A, B, and C
```

#### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

## **Examples**

```
# Figure 11.12
FCD.aov <- aov(Weight ~ Diet, data = FCD)
checking.plots(FCD.aov)
rm(FCD.aov)</pre>
```

Fertilize

Cross and Auto Fertilization

# **Description**

Plants' heights in inches obtained from two seeds, one obtained by cross fertilization and the other by auto fertilization, in two opposite but separate locations of a pot are recorded.

# **Format**

A data frame with 15 observations on the following 2 variables:

```
cross height of plant in inches self height of plant in inches
```

# Source

Darwin, C. (1876) The Effect of Cross and Self-Fertilization in the Vegetable Kingdom

#### References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
with(data = Fertilize,
t.test(cross, self))
```

26 Formula1

food

Carrot Shear

# **Description**

Shear measured in kN on frozen carrots from four randomly selected freezers

## **Format**

A data frame with 16 observations on the following 2 variables:

shear carrot shear measured in kN

freezer a factor with levels A, B, C, and D

#### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

# **Examples**

```
summary(aov(shear ~ freezer, data = food))
```

Formula1

Pit Stop Times

## **Description**

Pit stop times for two teams at 10 randomly selected Formula 1 races

# **Format**

A data frame with 10 observations on the following 3 variables:

Race number corresponding to a race site

**Team1** pit stop times for team one

Team2 pit stop times for team two

# Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

GD 27

# **Examples**

```
with(data = Formula1,
boxplot(Team1, Team2))
```

GD

Times Until Failure

# Description

Contains time until failure in hours for a particular electronic component subjected to an accelerated stress test.

# **Format**

A data frame with 100 observations on the following variable:

attf times until failure in hours

# Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
with(data = GD,
hist(attf, prob = TRUE))
with(data = GD,
lines(density(attf)))
# Trellis Approach
histogram(~attf, data = GD, type = "density",
panel = function(x, ...) {
panel.histogram(x, ...)
panel.densityplot(x, col = "blue", plot.points = TRUE, lwd = 2)
} )
```

28 Grades

glucose

Blood Glucose Levels

# Description

Fifteen diabetic patients were randomly selected, and their blood glucose levels were measured in mg/100 ml with two different devices.

## **Format**

A data frame with 15 observations on the following 3 variables:

Patient patient number

Old blood glucose level in mg/100 ml using old device

New blood glucose level in mg/100 ml using new device

## **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

## **Examples**

```
with(data = glucose,
boxplot(Old, New))
```

Grades

GPA and SAT Scores

## **Description**

The admissions committee of a comprehensive state university selected at random the records of 200 second semester freshmen. The results, first semester college GPA and SAT scores, are stored in the data frame Grades. Data are used in Example 12.6.

#### **Format**

A data frame with 200 observations on the following 2 variables:

```
sat SAT scoregpa grade point average
```

Grocery 29

## **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

## **Examples**

```
# traditional scatterplot
plot(gpa ~ sat, data = Grades)
# trellis scatterplot
xyplot(gpa ~ sat, data = Grades, type=c("p", "smooth"))
```

Grocery

**Grocery Spending** 

# Description

The consumer expenditure survey, created by the U.S. Department of Labor, was administered to 30 households in Watauga County, North Carolina, to see how the cost of living in Watauga county with respect to total dollars spent on groceries compares with other counties. The amount of money each household spent per week on groceries is stored in the variable groceries. Data are used in Example 8.3.

# Format

A data frame with 30 observations on the following variable:

groceries total dollars spent on groceries

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
with(data = Grocery,
z.test(x = groceries, sigma.x = 30, conf.level = 0.97)$conf)
```

30 House

HardWater

Mortality and Water Hardness

## **Description**

Mortality and drinking water hardness for 61 cities in England and Wales.

#### **Format**

A data frame with 61 observations on the following 4 variables.

location a factor with levels North South indicating whether the town is as north as Derby

town the name of the town

mortality averaged annual mortality per 100,000 males

hardness calcium concentration (in parts per million)

#### **Details**

These data were collected in an investigation of environmental causes of disease. They show the annual mortality rate per 100,000 for males, averaged over the years 1958-1964, and the calcium concentration (in parts per million) in the drinking water supply for 61 large towns in England and Wales. (The higher the calcium concentration, the harder the water.)

## Source

D. J. Hand, F. Daly, A. D. Lunn, K. J. McConway and E. Ostrowski (1994) *A Handbook of Small Datasets*. Chapman and Hall/CRC, London.

## **Examples**

```
plot(mortality ~ hardness, data = HardWater)
```

House

House Prices

## Description

Random sample of house prices (in thousands of dollars) for three bedroom/two bath houses in Watauga County, NC

HSwrestler 31

## **Format**

A data frame with 14 observations on the following 2 variables:

**Neighborhood** a factor with levels Blowing Rock, Cove Creek, Green Valley, Park Valley, Parkway, and Valley Crucis

Price price of house (in thousands of dollars)

# **Examples**

```
with(data = House,
t.test(Price))
```

**HSwrestler** 

High School Wrestlers

## **Description**

The body fat of 78 high school wrestlers was measured using three separate techniques, and the results are stored in the data frame HSwrestler. The techniques used were hydrostatic weighing (HWFAT), skin fold measurements (SKFAT), and the Tanita body fat scale (TANFAT). Data are used in Examples 10.11, 12.11, and 12.12.

#### **Format**

A data frame with 78 observations on the following 9 variables:

AGE age of wrestler in years

HT height of wrestler in inches

WT weight ofwrestler in pounds

ABS abdominal fat

TRICEPS tricep fat

SUBSCAP subscapular fat

HWFAT hydrostatic fat

TANFAT Tanita fat

SKFAT skin fat

#### Source

Data provided by Dr. Alan Utter, Department of Health Leisure and Exercise Science, Appalachian State University.

32 Hubble

## References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R*. Chapman & Hall/CRC.

## **Examples**

```
FAT <- c(HSwrestler$HWFAT, HSwrestler$SKFAT) GROUP <- factor(rep(c("HWFAT", "TANFAT", "SKFAT"), rep(78, 3))) BLOCK <- factor(rep(1:78, 3)) friedman.test(FAT \sim GROUP | BLOCK)
```

Hubble

Hubble Telescope

## **Description**

The Hubble Space Telescope was put into orbit on April 25, 1990. Unfortunately, on June 25, 1990, a spherical aberration was discovered in Hubble's primary mirror. To correct this, astronauts had to work in space. To prepare for the mission, two teams of astronauts practiced making repairs under simulated space conditions. Each team of astronauts went through 15 identical scenarios. The times to complete each scenario were recorded in days.

## **Format**

A data frame with 15 observations on the following 2 variables:

**Team1** days to complete scenario

**Team2** days to complete scenario

## Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
with(data = Hubble,
qqnorm(Team1 - Team2))
with(data = Hubble,
qqline(Team1 - Team2))
# Trellis Approach
qqmath(~(Team1 - Team2), data = Hubble, type=c("p", "r"))
```

InsurQuotes 33

InsurQuotes

Insurance Quotes

## **Description**

Insurance quotes for two insurers of hazardous waste jobs

#### **Format**

A data frame with 15 observations on the following 2 variables:

```
companyA quotes from company A in euroscompanyB quotes from company B in euros
```

## **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

# **Examples**

```
with(data = InsurQuotes,
t.test(companyA, companyB))
```

interval.plot

Interval Plot

# Description

Function to graph intervals

# Usage

```
interval.plot(ll, ul, parameter = 0)
```

## **Arguments**

vector of lower valuesvector of upper values

parameter value of the desired parameter (used when graphing confidence intervals)

# Value

Draws user-given intervals on a graphical device.

janka janka

## Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

# **Examples**

```
set.seed(385)
samples <- 100
n <- 625
ll <- numeric(samples)
ul <- numeric(samples)
xbar <- numeric(samples)
for (i in 1:samples){
   xbar[i] <- mean(rnorm(n, 80, 25))
   ll[i] <- xbar[i] - qnorm(.975)*25/sqrt(n)
   ul[i] <- xbar[i] + qnorm(.975)*25/sqrt(n)
   }
interval.plot(ll, ul, parameter = 80)</pre>
```

janka

Australian Eucalypt Hardwoods

# **Description**

The dataset consists of density and hardness measurements from 36 Australian Eucalypt hardwoods.

#### **Format**

A data frame with 36 observations on the following 2 variables.

**Density** a measure of density of the timber

Hardness the Janka hardness of the timber

## **Details**

Janka Hardness is an importance rating of Australian hardwood timbers. The test measures the force required to imbed a steel ball into a piece of wood.

#### **Source**

Williams, E.J. (1959) Regression Analysis. John Wiley & Sons, New York.

```
with(data = janka, plot(Hardness ~ Density, col = "blue"))
```

Kinder 35

Kinder

Kindergarten Class

## **Description**

The data frame Kinder contains the height in inches and weight in pounds of 20 children from a kindergarten class. Data are used in Example 12.17.

# **Format**

A data frame with 20 observations on the following 2 variables:

```
ht height in inches of child
```

wt weight in pounds of child

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

# **Examples**

```
# Figure 12.10
with(data = Kinder,
plot(wt, ht))
# Trellis Approach
xyplot(ht ~ wt, data = Kinder)
```

ksdist

Simulated Distribution of  $D_n$  (Kolmogorov-Smirnov)

## **Description**

Function to visualize the sampling distribution of  $\mathcal{D}_n$  (the Kolmogorov-Smirnov one sample statistic) and to find simulated critical values.

# Usage

```
ksdist(n = 10, sims = 10000, alpha = 0.05)
```

#### **Arguments**

n sample size

sims number of simulations to perform

alpha desired  $\alpha$  level

36 ksLdist

## Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

#### See Also

ksLdist

# **Examples**

```
ksdist(n = 10, sims = 15000, alpha =0.05)
```

ksLdist

Simulated Lilliefors' Test of Normality Values

# Description

Function to visualize the sampling distribution of  $\mathcal{D}_n$  (the Kolmogorov-Smirnov one sample statistic) for simple and composite hypotheses

# Usage

```
ksLdist(n = 10, sims = 10000, alpha = 0.05)
```

# **Arguments**

n sample size

sims number of simulations to perform

alpha desired  $\alpha$  level

# Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

# See Also

ksdist

```
ksLdist(n = 10, sims = 1500, alpha = 0.05)
```

Leddiode 37

Leddiode

LED Diodes

### **Description**

The diameter in millimeters for a random sample of 15 diodes from each of the two suppliers is stored in the data frame Leddiode.

#### **Format**

A data frame with 15 observations on the following 2 variables:

supplierA diameter in millimeters of diodes from supplier AsupplierB diameter in millimeters of diodes from supplier B

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

## **Examples**

```
with(data = Leddiode,
boxplot(supplierA, supplierB, col = c("red", "blue")))
```

LostR

Lost Revenue Due to Worker Illness

### **Description**

Data set containing the lost revenue in dollars/day and number of workers absent due to illness for a metallurgic company

#### **Format**

A data frame with 25 observations on the following 2 variables:

NumberSick number of absent workers due to illness

LostRevenue lost revenue in dollars

#### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

38 normarea

### **Examples**

```
xyplot(LostRevenue ~ NumberSick, data = LostR, type=c("p", "r"))
```

MilkCarton

Milk Carton Drying Times

### **Description**

A plastics manufacturer makes two sizes of milk containers: half gallon and gallon sizes. The time required for each size to dry is recorded in seconds in the data frame MilkCarton.

#### **Format**

A data frame with 40 observations on the following 2 variables:

**Hgallon** drying time in seconds for half gallon containers

Wgallon drying time in seconds for whole gallon containers

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

### **Examples**

```
with(data = MilkCarton,
boxplot(Hgallon, Wgallon))
```

normarea

Normal Area

### **Description**

Function that computes and draws the area between two user specified values in a user specified normal distribution with a given mean and standard deviation

### Usage

```
normarea(lower = -Inf, upper = Inf, m = 0, sig = 1)
```

nsize 39

## **Arguments**

lower	the desired lower value
upper	the desired upper value
m	the mean for the population (default is the standard normal with $m=0$ )
sig	the standard deviation of the population (default is the standard normal with sig = 1)

# Value

Draws the specified area in a graphics device

### Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

## **Examples**

```
# Finds and graphically illustrates P(70 < X < 130) given X is N(100, 15) normarea(lower = 70, upper = 130, m = 100, sig = 15)
```

nsize	Required Sample Size	

## Description

Function to determine required sample size to be within a given margin of error

## Usage

```
nsize(b, sigma = NULL, p = 0.5, conf.level = 0.95, type = c("mu", "pi"))
```

## Arguments

b	the desired bound
sigma	population standard deviation; not required if using type "pi"
p	estimate for the population proportion of successes; not required if using type " $mu$ "
conf.level	confidence level for the problem, restricted to lie between zero and one
type	character string, one of "mu" or "pi", or just the initial letter of each, indicating the appropriate parameter; default value is "mu"

## **Details**

Answer is based on a normal approximation when using type "pi".

40 ntester

#### Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

### **Examples**

```
nsize(b = 0.015, p = 0.5, conf.level = 0.95, type = "pi") # Returns the required sample size (n) to estimate the population # proportion of successes with a 0.95 confidence interval # so that the margin of error is no more than 0.015 when the # estimate of the population propotion of successes is 0.5. nsize(b = 0.02, sigma = 0.1, conf.level = 0.95, type = "mu") # Returns the required sample size (n) to estimate the population # mean with a 0.95 confidence interval so that the margin # of error is no more than 0.02.
```

ntester

Normality Tester

### Description

Q-Q plots of randomly generated normal data of the same sample size as the tested data are generated and plotted on the perimeter of the graph while a Q-Q plot of the actual data is depicted in the center of the graph.

### Usage

```
ntester(actual.data)
```

#### **Arguments**

actual.data

is a numeric vector. Missing and infinite values are allowed, but are ignored in the calculation. The length of actual.data must be less than 5000 after dropping nonfinite values.

#### **Details**

Q-Q plots of randomly generated normal data of the same size as the tested data are generated and plotted on the perimeter of the graph sheet while a Q-Q plot of the actual data is depicted in the center of the graph. The p-values are calculated based on the Shapiro-Wilk W-statistic. Function will only work on numeric vectors containing less than or equal to 5000 observations. Best used for moderate sized samples (n < 50).

### Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

oneway.plots 41

#### References

Shapiro, S.S. and Wilk, M.B. 1965. *An analysis of variance test for normality (complete samples)*. Biometrika **52**: 591-611.

### **Examples**

```
ntester(actual.data = rexp(40, 1))
# Q-Q plot of random exponential data in center plot
# surrounded by 8 Q-Q plots of randomly generated
# standard normal data of size 40.
```

oneway.plots

Exploratory Graphs for Single Factor Designs

## Description

Function to create dotplots, boxplots, and design plot (means) for single factor designs

### Usage

```
oneway.plots(Y, fac1, COL = c("#A9E2FF", "#0080FF"))
```

## Arguments

Y response variable for a single factor design

fac1 predictor variable (factor)
COL a vector with two colors

### Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

#### See Also

```
twoway.plots, checking.plots
```

```
with(data = Tire, oneway.plots(StopDist, tire))
```

42 Phenyl

Phenyl

Phenylketonuria

## Description

The data frame Phenyl records the level of Q10 at four different times for 46 patients diagnosed with phenylketonuria. The variable Q10.1 contains the level of Q10 measured in micromoles for the 46 patients. Q10.2, Q10.3, and Q10.4 are the values recorded at later times respectively for the 46 patients.

#### **Format**

A data frame with 46 observations on the following 4 variables.

Q10.1 level of Q10 at time 1 in micromoles

Q10.2 level of Q10 at time 2 in micromoles

Q10.3 level of Q10 at time 3 in micromoles

**Q10.4** level of Q10 at time 4 in micromoles

### **Details**

Phenylketonuria (PKU) is a genetic disorder that is characterized by an inability of the body to utilize the essential amino acid, phenylalanine. Research suggests patients with phenylketonuria have deficiencies in coenzyme Q10.

### Source

Artuch, R., et. al. (2004) "Study of Antioxidant Status in Phenylketonuric Patients." Clinical Biochemistry, 37: 198-203.

#### References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R*. Chapman & Hall/CRC.

```
with(data = Phenyl,
t.test(Q10.1, conf.level = 0.99))
```

Phone 43

Phone

Telephone Call Times

### **Description**

Phone contains times in minutes of long distance telephone calls during a one month period for a small business. Data are used in Example 10.1.

#### **Format**

A data frame with 23 observations on the following variable:

call.time time spent on long distance calls in minutes

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R*. Chapman & Hall/CRC.

## **Examples**

```
with(data = Phone,
SIGN.test(call.time, md = 2.1))
```

Rat

Rat Survival Time

### **Description**

The survival time in weeks of 20 male rats exposed to high levels of radiation.

#### **Format**

A data frame with 20 observations on the following variable:

survival.time number of weeks survived

#### **Source**

Lawless, J. (1982) Statistical Models and Methods for Lifetime Data. John Wiley, New York.

#### References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

Ratbp

### **Examples**

```
with(data = Rat,
EDA(survival.time))
```

Ratbp

Rat Blood Pressure

## Description

Twelve rats were chosen, and a drug was administered to six rats, the treatment group, chosen at random. The other six rats, the control group, received a placebo. The drops in blood pressure (mmHg) for the treatment group (with probability distribution F) and the control group (with probability distribution G) are stored in the variables Treat and Cont, respectively. Data are used in Example 10.18.

#### **Format**

A data frame with 6 observations on the following 2 variables:

Treat drops in blood pressure in mmHg for treatment group

Cont drops in blood pressure in mmHg for control group

#### Source

The data is originally from Ott and Mendenhall (1985, problem 8.17).

### References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
with(data = Ratbp,
boxplot(Treat, Cont))
```

Refrigerator 45

Refrigerator	Refrigerator Energy Consumption	

### **Description**

Thirty 18 cubic feet refrigerators were randomly selected from a company's warehouse. The first fifteen had their motors modified while the last fifteen were left intact. The energy consumption (kilowatts) for a 24 hour period for each refrigerator was recorded and stored in the data frame Refrigerator. The refrigerators with the design modification are stored in the variable modelA, and those without the design modification are stored in the variable modelB.

#### **Format**

A data frame with 30 observations on the following 2 variables.

```
modelA energy consumption in kilowatts for a 24 hour period modelB energy consumption in kilowatts for a 24 hour period
```

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

#### **Examples**

```
with(data = Refrigerator,
boxplot(modelA, modelB))
```

|--|

### **Description**

A laboratory is interested in testing a new child friendly pesticide on *Blatta orientalis* (oriental cockroaches). Scientists apply the new pesticide to 81 randomly selected Blatta orientalis oothecae (eggs). The results from the experiment are stored in the data frame Roacheggs in the variable eggs. A zero in the variable eggs indicates that nothing hatched from the egg while a 1 indicates the birth of a cockroach. Data is used in Example 7.16.

## **Format**

A data frame with 81 observations on the following variable:

**eggs** numeric vector where a 0 indicates nothing hatched while a 1 indicates the birth of a cockroach.

46 Salinity

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

### **Examples**

```
p <- seq(0.1, 0.9, 0.001)
negloglike <- function(p){
  -(sum(Roacheggs$eggs)*log(p) + sum(1 - Roacheggs$eggs)*log(1 - p))
}
nlm(negloglike, 0.2)
rm(negloglike)</pre>
```

Salinity

Surface Water Salinity

## Description

Surface-water salinity measurements were taken in a bottom-sampling project in Whitewater Bay, Florida. These data are stored in the data frame Salinity.

### **Format**

A data frame with 48 observations on the following variable:

salinity surface-water salinity measurements

## Source

Davis, J. (1986) Statistics and Data Analysis in Geology. John Wiley, New York.

### References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R*. Chapman & Hall/CRC.

```
with(data = Salinity,
EDA(salinity))
```

satfruit 47

satfruit Fruit Trees

#### **Description**

To estimate the total surface occupied by fruit trees in 3 small areas (R63, R67, and R68) of Navarra (Spain) in 2001, a sample of 47 square segments has been taken. The experimental units are square segments or quadrats of 4 hectares, obtained by random sampling after overlaying a square grid on the study domain. Data are used in Case Study: Fruit Trees, Chapter 12.

#### **Format**

A data frame with 47 observations on the following 17 variables:

QUADRAT number of the sampled segment or quadrat

SArea the small area, a factor with levels R63, R67, and R68

WH area classified as wheat in sampled segment

BA area classified as barley in sampled segment

NAR area classified as non arable in sampled segment

COR area classified as corn in sampled segment

SF area classified as sunflower in sampled segment

VI area classified as vineyard in sampled segment

PS area classified as grass in sampled segment

ES area classified as asparagus in sampled segment

AF area classified as lucerne in sampled segment

**CO** area classified as rape (*Brassica Napus*) in sampled segment

AR area classified as rice in sampled segment

AL area classified as almonds in sampled segment

OL area classified as olives in sampled segment

FR area classified as fruit trees in sampled segment

**OBS** the observed area of fruit trees in sampled segment

### Source

Militino, A. F., et. al. (2006) "Using Small Area Models to Estimate the Total Area Occupied by Olive Trees." Journal of Agricultural, Biological and Environmental Statistics, 11: 450-461.

### References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

48 SBIQ

### **Examples**

```
with(data = satfruit,
pairs(satfruit[ , 15:17]))
# Trellis Approach
splom(~data.frame(satfruit[ , 15:17]), data = satfruit)
```

SBIQ

County IQ

### **Description**

A school psychologist administered the Stanford-Binet intelligence quotient (IQ) test in two counties. Forty randomly selected, gifted and talented students were selected from each county. The Stanford-Binet IQ test is said to follow a normal distribution with a mean of 100 and standard deviation of 16.

#### **Format**

A data frame with 40 observations on the following 2 variables:

County1 IQ scores for county one

County2 IQ scores for county two

### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R*. Chapman & Hall/CRC.

```
with(data = SBIQ, qqnorm(County1))
with(data = SBIQ, qqline(County1))
# Trellis Approach
qqmath(~County1, data = SBIQ, type=c("p", "r"))
```

Schizo 49

Schizo

Dopamine Activity

### **Description**

Twenty-five patients with schizophrenia were classified as psychotic or nonpsychotic after being treated with an antipsychotic drug. Samples of cerebral fluid were taken from each patient and assayed for dopamine b-hydroxylase (DBH) activity. The dopamine measurements for the two groups are in nmol/(ml)(h)/(mg) of protein.

#### **Format**

A data frame with 15 observations on the following 2 variables:

**nonpsychotic** dopamine activity level for patients classified nonpsychotic **psychotic** dopamine activity level for patients classified psychotic

#### **Source**

Sternberg, D. E., Van Kammen, D. P., and Bunney, W. E. (1982) "Schizophrenia: Dopamine b-Hydroxylase Activity and Treatment Response." *Science*, **216**: 1423-1425.

#### References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

#### **Examples**

```
with(data = Schizo, boxplot(nonpsychotic, psychotic,
names = c("nonpsychotic", "psychotic"), col = c("green", "red")))
```

Score

Standardized Test Scores

### **Description**

Standardized test scores from a random sample of twenty college freshmen.

#### **Format**

A data frame with 20 observations on the following variable:

scores standardized test score

50 SDS4

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

### **Examples**

```
qqmath(~scores, data = Score, type=c("p", "r"))
```

SDS4

M1 Motorspeedway Times

### **Description**

The times recorded are those for 41 successive vehicles travelling northwards along the M1 motorway in England when passing a fixed point near Junction 13 in Bedfordshire on Saturday, March 23, 1985. After subtracting the times, the following 40 interarrival times reported to the nearest second are stored in SDS4 under the variable Times. Data are used in Example 10.17.

### **Format**

A data frame with 40 observations on the following variable:

Times interarrival times to the nearest second

### **Source**

Hand, D. J., et. al. (1994) A Handbook of Small Data Sets. Chapman & Hall, London.

### References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
with(data = SDS4, hist(Times))
```

SIGN.test 51

### **Description**

This function will test a hypothesis based on the sign test and reports linearly interpolated confidence intervals for one sample problems.

### Usage

```
SIGN.test(
   x,
   y = NULL,
   md = 0,
   alternative = "two.sided",
   conf.level = 0.95,
   ...
)
```

## Arguments

x	numeric vector; NAs and Infs are allowed but will be removed.
у	optional numeric vector; NAs and Infs are allowed but will be removed.
md	a single number representing the value of the population median specified by the null hypothesis
alternative	is a character string, one of "greater", "less", or "two.sided", or the initial letter of each, indicating the specification of the alternative hypothesis. For one-sample tests, alternative refers to the true median of the parent population in relation to the hypothesized value of the median.
conf.level	confidence level for the returned confidence interval, restricted to lie between zero and one
	further arguments to be passed to or from methods

## Details

Computes a "Dependent-samples Sign-Test" if both x and y are provided. If only x is provided, computes the "Sign-Test."

## Value

A list of class htest\_S, containing the following components:

statistic the S-statistic (the number of positive differences between the data and the hypothesized median), with names attribute "S".

p.value the p-value for the test

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conf.int	is a confidence interval (vector of length 2) for the true median based on linear interpolation. The confidence level is recorded in the attribute conf.level. When the alternative is not "two.sided", the confidence interval will be half-infinite, to reflect the interpretation of a confidence interval as the set of all values k for which one would not reject the null hypothesis that the true mean
	or difference in means is k. Here infinity will be represented by Inf.
estimate	is avector of length 1, giving the sample median; this estimates the corresponding population parameter. Component estimate has a names attribute describing its elements.
null.value	is the value of the median specified by the null hypothesis. This equals the input argument md. Component null.value has a names attribute describing its elements.
alternative	records the value of the input argument alternative: "greater", "less", or "two.sided" $$
data.name	a character string (vector of length 1) containing the actual name of the input vector x
Confidence.Int	ervals
	a 3 by 3 matrix containing the lower achieved confidence interval, the interpolated confidence interval, and the upper achived confidence interval

### **Null Hypothesis**

For the one-sample sign-test, the null hypothesis is that the median of the population from which x is drawn is md. For the two-sample dependent case, the null hypothesis is that the median for the differences of the populations from which x and y are drawn is md. The alternative hypothesis indicates the direction of divergence of the population median for x from md (i.e., "greater", "less", "two.sided".)

#### **Assumptions**

The median test assumes the parent population is continuous.

### Note

The reported confidence interval is based on linear interpolation. The lower and upper confidence levels are exact.

#### Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

#### References

- Gibbons, J.D. and Chakraborti, S. 1992. *Nonparametric Statistical Inference*. Marcel Dekker Inc., New York.
- Kitchens, L.J. 2003. Basic Statistics and Data Analysis. Duxbury.
- Conover, W. J. 1980. Practical Nonparametric Statistics, 2nd ed. Wiley, New York.
- Lehmann, E. L. 1975. *Nonparametrics: Statistical Methods Based on Ranks*. Holden and Day, San Francisco.

SimDataST 53

### See Also

```
z.test, zsum.test, tsum.test
```

### **Examples**

```
with(data = Phone, SIGN.test(call.time, md = 2.1))
# Computes two-sided sign-test for the null hypothesis
# that the population median is 2.1. The alternative
# hypothesis is that the median is not 2.1. An interpolated
# upper 95% upper bound for the population median will be computed.
```

SimDataST

Simulated Data (Predictors)

### **Description**

Simulated data for five variables. Data are used with Example 12.21.

#### **Format**

A data frame with 200 observations on the following 5 variables:

- Y1 a numeric vector
- Y2 a numeric vector
- x1 a numeric vector
- x2 a numeric vector
- x3 a numeric vector

### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R*. Chapman & Hall/CRC.

```
xyplot(Y1 ~ x1, data = SimDataST, type=c("p", "smooth"))
```

54 Soccer

SimDataXT

Simulated Data (Logarithms)

### **Description**

Simulated data for four varaibles. Data are used with Example 12.18.

#### **Format**

A data frame with 200 observations on the following 4 variables:

Y a numeric vector

x1 a numeric vector

x2 a numeric vector

x3 a numeric vector

#### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

### **Examples**

```
xyplot(Y ~ x1, data = SimDataXT, type=c("p", "smooth"))
```

Soccer

World Cup Soccer

### **Description**

Soccer contains how many goals were scored in the regulation 90 minute periods of World Cup soccer matches from 1990 to 2002. Data are used in Example 4.4.

## **Format**

A data frame with 575 observations on the following 3 variables:

CGT cumulative goal time in minutes

Game game in which goals were scored

Goals number of goals scored in regulation period

SRS 55

#### **Details**

The World Cup is played once every four years. National teams from all over the world compete. In 2002 and in 1998, thirty-six teams were invited; whereas, in 1994 and in 1990, only 24 teams participated. The data frame Soccer contains three columns: CGT, Game, and Goals. All of the information contained in Soccer is indirectly available from the FIFA World Cup website, located at https://www.fifa.com/.

#### Source

Chu, S. (2003) "Using Soccer Goals to Motivate the Poisson Process." *INFORMS* Transaction on Education, **3**, **2**: 62-68.

#### References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R*. Chapman & Hall/CRC.

### **Examples**

```
with(data = Soccer,
table(Goals))
```

SRS

Simple Random Sample

## **Description**

Computes all possible samples from a given population using simple random sampling

## Usage

```
SRS(popvalues, n)
```

### **Arguments**

popvalues are values of the population. NAs and Infs are allowed but will be removed from

the population.

n the sample size

#### **Details**

If non-finite values are entered as part of the population, they are removed; and the returned simple random sample computed is based on the remaining finite values.

StatTemps

#### Value

The function srs() returns a matrix containing the possible simple random samples of size n taken from a population of finite values population.

### Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

#### See Also

combn

#### **Examples**

```
SRS(popvalues = c(5, 8, 3, NA, Inf), n = 2)
```

StatTemps

Student Temperature

## **Description**

In a study conducted at Appalachian State University, students used digital oral thermometers to record their temperatures each day they came to class. A randomly selected day of student temperatures is provided in StatTemps. Information is also provided with regard to subject gender and the hour of the day when the students' temperatures were measured.

### **Format**

A data frame with 34 observations on the following 3 variables:

```
temperature temperature in farenheit gender a factor with levels Female and Male class a factor with levels 8 a.m. and 9 a.m.
```

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
bwplot(gender ~ temperature, data = StatTemps)
```

Stschool 57

Stschool

School Satisfaction

#### **Description**

A questionnaire is randomly administered to 11 students from State School X and to 15 students from State School Y (the results have been ordered and stored in the data frame Stschool). Data are used in Example 9.11.

#### **Format**

A data frame with 26 observations on the following 4 variables:

X satisfaction score

Y satisfaction score

Satisfaction combined satisfaction scores

**School** a factor with levels X and Y

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R*. Chapman & Hall/CRC.

## Examples

```
with(data = Stschool,
t.test(X, Y, var.equal = TRUE))
```

Sundig

Workstation Comparison

#### **Description**

To compare the speed differences between two different brands of workstations (Sun and Digital), the times each brand took to complete complex simulations was recorded. Five complex simulations were selected, and the five selected simulations were run on both workstations. The resulting times in minutes for the five simulations are stored in data frame Sundig.

#### **Format**

A data frame with 5 observations on the following 3 variables:

SUN time in seconds for a Sun workstation to complete a simulation

**DIGITAL** time in seconds for a Digital workstation to complete a simulation

d difference between Sun and Digital

58 sunflower

#### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

#### **Examples**

```
with(data = Sundig,
t.test(SUN, DIGITAL, paired = TRUE)$conf)
```

sunflower

Sunflower Defoliation

### **Description**

Seventy-two field trials were conducted by applying four defoliation treatments (non defoliated control, 33%, 66%, and 100%) at different growth stages (stage) ranging from pre-flowering (1) to physiological maturity (5) in four different locations of Navarra, Spain: Carcastillo (1), Melida (2), Murillo (3), and Unciti (4). There are two response variables: yield in kg/ha of the sunflower and numseed, the number of seeds per sunflower head. Data are stored in the data frame sunflower. Data used in Case Study: Sunflower defoliation from Chapter 11.

#### **Format**

A data frame with 72 observations on the following 5 variables:

**location** a factor with levels A, B, C, and D for locations Carcastillo, Melida, Murillo, and Unciti respectively

```
stage a factor with levels stage1, stage2, stage3, stage4, and stage5
```

defoli a factor with levels control, treat1, treat2, and treat3

yield sunflower yield in kg/ha

numseed number of seeds per sunflower head

#### Source

Muro, J., et. al. (2001) "Defoliation Effects on Sunflower Yield Reduction." Agronomy Journal, 93: 634-637.

#### References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
summary(aov(yield ~ stage + defoli + stage:defoli, data = sunflower))
```

SurfaceSpain 59

SurfaceSpain	Surface Area for Spanish Communities	

### **Description**

Surface area (km<sup>2</sup>) for seventeen autonomous Spanish communities.

#### **Format**

A data frame with 17 observations on the following 2 variables:

```
community a factor with levels Andalucia, Aragon, Asturias, Baleares, C. Valenciana, Canarias,
    Cantabria, Castilla-La Mancha, Castilla-Leon, Cataluna, Extremadura, Galicia, La
    Rioja, Madrid, Murcia, Navarra, and P. Vasco
```

surface surface area in km<sup>2</sup>

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

### **Examples**

```
with(data = SurfaceSpain,
barplot(surface, names.arg = community, las = 2))
# Trellis Approach
barchart(community ~ surface, data = SurfaceSpain)
```

Swimtimes Swim Times
----------------------

### **Description**

Swimmers' improvements in seconds for two diets are stored in the data frame Swimtimes. The values in highfat represent the time improvement in seconds for swimmers on a high fat diet, and the values in lowfat represent the time improvement in seconds for swimmers on a low fat diet. Data are used in Example 10.9.

#### **Format**

A data frame with 14 observations on the following 2 variables:

```
lowfat time improvement in seconds highfat time improvement in seconds
```

60 Tennis

#### **Details**

Times for the thirty-two swimmers for the 200 yard individual medley were taken right after the swimmers' conference meet. The swimmers were randomly assigned to follow one of the diets. The group on diet 1 followed a low fat diet the entire year but lost two swimmers along the way. The group on diet 2 followed the high fat diet the entire year and also lost two swimmers.

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

### **Examples**

```
with(data = Swimtimes,
wilcox.test(highfat, lowfat))
```

Tennis

Speed Detector

### Description

The Yonalasee tennis club has two systems to measure the speed of a tennis ball. The local tennis pro suspects one system (Speed1) consistently records faster speeds. To test her suspicions, she sets up both systems and records the speeds of 12 serves (three serves from each side of the court). The values are stored in the data frame Tennis in the variables Speed1 and Speed2. The recorded speeds are in kilometers per hour.

#### **Format**

A data frame with 12 observations on the following 2 variables:

```
Speed1 speed in kilometers per hour Speed2 speed in kilometers per hour
```

#### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
with(data = Tennis,
boxplot(Speed1, Speed2))
```

TestScores 61

TestScores

Statistics Grades

### Description

Test grades of 29 students taking a basic statistics course

#### **Format**

A data frame with 29 observations on the following variable:

```
grade test score
```

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

### **Examples**

```
with(data = TestScores,
EDA(grade))
```

Tire

Tire Stopping Distances

### **Description**

The data frame Tire has the stopping distances measured to the nearest foot for a standard sized car to come to a complete stop from a speed of sixty miles per hour. There are six measurements of the stopping distance for four different tread patterns labeled A, B, C, and D. The same driver and car were used for all twenty-four measurements. Data are used in Example 11.1 and 11.2.

#### **Format**

A data frame with 24 observations on the following 2 variables:

**StopDist** stopping distance measured to the nearest foot **tire** a factor with levels A, B, C, and D

#### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

62 TireWear

## **Examples**

```
summary(aov(StopDist ~ tire, data = Tire))
```

TireWear

Tire Wear

## Description

The data frame TireWear contains measurements for the amount of tread loss after 10,000 miles of driving in thousandths of an inch. Data are used in Example 11.8.

#### **Format**

A data frame with 16 observations on the following 3 variables:

Wear tread loss measured in thousandths of an inch

Treat a factor with levels A, B, C, and D

Block a factor with levels Car1, Car2, Car3, and Car4

### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
par(mfrow = c(1, 2), cex = 0.8)
with(data = TireWear,
interaction.plot(Treat, Block, Wear, type = "b", legend = FALSE))
with(data = TireWear,
interaction.plot(Block, Treat, Wear, type = "b", legend = FALSE))
par(mfrow = c(1, 1), cex = 1)
```

titanic3 63

titanic3

Titanic Survival Status

#### **Description**

The titanic3 data frame describes the survival status of individual passengers on the Titanic. The titanic3 data frame does not contain information for the crew, but it does contain actual and estimated ages for almost 80% of the passengers.

#### **Format**

A data frame with 1309 observations on the following 14 variables:

pclass a factor with levels 1st, 2nd, and 3rd

**survived** Survival (0 = No; 1 = Yes)

name Name

sex a factor with levels female and male

age age in years

sibsp Number of Siblings/Spouses Aboard

parch Number of Parents/Children Aboard

ticket Ticket Number

fare Passenger Fare

cabin Cabin

embarked a factor with levels Cherbourg, Queenstown, and Southampton

boat Lifeboat

body Body IdentificationNumber

home.dest Home/Destination

#### **Details**

Thomas Cason of UVa has greatly updated and improved the titanic data frame using the *Encyclopedia Titanica* and created a new dataset called titanic3. This dataset reflects the state of data available as of August 2, 1999. Some duplicate passengers have been dropped, many errors have been corrected, many missing ages have been filled in, and new variables have been created.

#### Source

https://hbiostat.org/data/repo/titanic.html

### References

Harrell, F. E. (2001) Regression Modeling Strategies with Applications to Linear Models, Logistic Regression, and Survival Analysis. Springer.

Top20

#### **Examples**

```
with(titanic3,
table(pclass, sex))
```

TOE

Nuclear Energy

### **Description**

Nuclear energy (in TOE, tons of oil equivalent) produced in 12 randomly selected European countries during 2003

#### **Format**

A data frame with 12 observations on the following variable:

energy nuclear energy measured in tons of oil equivalent

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

## **Examples**

```
with(TOE,
plot(density(energy)))
```

Top20

Tennis Income

### **Description**

Top20 contains data (in millions of dollars) corresponding to the earnings of 15 randomly selected tennis players whose earnings fall somewhere in positions 20 through 100 of ranked earnings.

### **Format**

A data frame with 15 observations on the following variable:

income yearly income in millions of dollars

tsum.test 65

#### **Source**

```
https://www.atptour.com/
```

#### References

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

## **Examples**

```
with(data = Top20,
EDA(income))
```

tsum.test

Summarized t-test

### **Description**

Performs a one-sample, two-sample, or a Welch modified two-sample t-test based on user supplied summary information. Output is identical to that produced with t.test.

### Usage

```
tsum.test(
  mean.x,
  s.x = NULL,
  n.x = NULL,
  mean.y = NULL,
  s.y = NULL,
  n.y = NULL,
  alternative = c("two.sided", "less", "greater"),
  mu = 0,
  var.equal = FALSE,
  conf.level = 0.95,
  ...
)
```

### **Arguments**

```
mean.x a single number representing the sample mean of x
s.x a single number representing the sample standard deviation of x
n.x a single number representing the sample size of x
mean.y a single number representing the sample mean of y
s.y a single number representing the sample standard deviation of y
```

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n.y a single number representing the sample size of y

alternative is a character string, one of "greater", "less", or "two.sided", or just the

initial letter of each, indicating the specification of the alternative hypothesis. For one-sample tests, alternative refers to the true mean of the parent population in relation to the hypothesized value mu. For the standard two-sample tests, alternative refers to the difference between the true population mean for x and that for y, in relation to mu. For the one-sample and paired t-tests, alternative refers to the true mean of the parent population in relation to the hypothesized value mu. For the standard and Welch modified two-sample t-tests, alternative refers to the difference between the true population mean for x and that for y, in relation to mu. For the one-sample t-tests, alternative refers to the true mean of the parent population in relation to the hypothesized value mu. For the standard and Welch modified two-sample t-tests, alternative refers to the difference between the true population mean for x and that for y, in relation to

mu.

mu is a single number representing the value of the mean or difference in means

specified by the null hypothesis.

var.equal logical flag: if TRUE, the variances of the parent populations of x and y are as-

sumed equal. Argument var.equal should be supplied only for the two-sample

tests.

conf.level is the confidence level for the returned confidence interval; it must lie between

zero and one.

... Other arguments passed onto tsum.test()

### **Details**

If y is NULL, a one-sample t-test is carried out with x. If y is not NULL, either a standard or Welch modified two-sample t-test is performed, depending on whether var.equal is TRUE or FALSE.

## Value

A list of class htest, containing the following components:

statistic the t-statistic, with names attribute "t"

parameters is the degrees of freedom of the t-distribution associated with statistic. Compo-

nent parameters has names attribute "df".

p. value the p-value for the test

conf.int is a confidence interval (vector of length 2) for the true mean or difference in

means. The confidence level is recorded in the attribute conf.level. When alternative is not "two.sided", the confidence interval will be half-infinite, to reflect the interpretation of a confidence interval as the set of all values k for which one would not reject the null hypothesis that the true mean or difference

in means is k. Here infinity will be represented by Inf.

estimate is a vector of length 1 or 2, giving the sample mean(s) or mean of differences;

these estimate the corresponding population parameters. Component estimate

has a names attribute describing its elements.

tsum.test 67

null.value is the value of the mean or difference in means specified by the null hypothesis. This equals the input argument mu. Component null.value has a names attribute describing its elements.

alternative records the value of the input argument alternative: "greater", "less" or "two.sided".

data.name is a character string (vector of length 1) containing the names x and y for the two summarized samples.

### **Null Hypothesis**

For the one-sample t-test, the null hypothesis is that the mean of the population from which x is drawn is mu. For the standard and Welch modified two-sample t-tests, the null hypothesis is that the population mean for x less that for y is mu.

The alternative hypothesis in each case indicates the direction of divergence of the population mean for x (or difference of means for x and y) from mu (i.e., "greater", "less", or "two.sided").

#### **Test Assumptions**

The assumption of equal population variances is central to the standard two-sample t-test. This test can be misleading when population variances are not equal, as the null distribution of the test statistic is no longer a t-distribution. If the assumption of equal variances is doubtful with respect to a particular dataset, the Welch modification of the t-test should be used.

The t-test and the associated confidence interval are quite robust with respect to level toward heavy-tailed non-Gaussian distributions (e.g., data with outliers). However, the t-test is non-robust with respect to power, and the confidence interval is non-robust with respect to average length, toward these same types of distributions.

### **Confidence Intervals**

For each of the above tests, an expression for the related confidence interval (returned component conf.int) can be obtained in the usual way by inverting the expression for the test statistic. Note that, as explained under the description of conf.int, the confidence interval will be half-infinite when alternative is not "two.sided"; infinity will be represented by Inf.

### Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

#### References

- Kitchens, L.J. 2003. Basic Statistics and Data Analysis. Duxbury.
- Hogg, R. V. and Craig, A. T. 1970. *Introduction to Mathematical Statistics, 3rd ed.* Toronto, Canada: Macmillan.
- Mood, A. M., Graybill, F. A. and Boes, D. C. 1974. *Introduction to the Theory of Statistics*, *3rd ed*. New York: McGraw-Hill.
- Snedecor, G. W. and Cochran, W. G. 1980. Statistical Methods, 7th ed. Ames, Iowa: Iowa State University Press.

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#### See Also

```
z.test, zsum.test
```

### **Examples**

```
# 95% Confidence Interval for mu1 - mu2, assuming equal variances round(tsum.test(mean.x = 53/15, mean.y = 77/11, s.x=sqrt((222 - 15*(53/15)^2)/14), s.y = sqrt((560 - 11*(77/11)^2)/10), n.x = 15, n.y = 11, var.equal = TRUE)$conf, 2) # One Sample t-test tsum.test(mean.x = 4, s.x = 2.89, n.x = 25, mu = 2.5)
```

twoway.plots

Exploratory Graphs for Two Factor Designs

## **Description**

Function creates side-by-side boxplots for each factor, a design plot (means), and an interaction plot.

### Usage

```
twoway.plots(Y, fac1, fac2, COL = c("#A9E2FF", "#0080FF"))
```

### **Arguments**

Y response variable

fac1 factor one fac2 factor two

COL a vector with two colors

### Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

#### See Also

```
oneway.plots, checking.plots
```

```
with(data = TireWear, twoway.plots(Wear, Treat, Block))
```

URLaddress 69

**URLaddress** 

Megabytes Downloaded

### **Description**

The manager of a URL commercial address is interested in predicting the number of megabytes downloaded, megasd, by clients according to the number minutes they are connected, mconnected. The manager randomly selects (megabyte, minute) pairs, and records the data. The pairs (megasd, mconnected) are stored in the data frame URLaddress.

#### **Format**

A data frame with 30 observations on the following 2 variables:

megasd megabytes dowloaded

mconnected number of minutes connected

#### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

### **Examples**

```
xyplot(mconnected ~ megasd, data = URLaddress, type=c("p", "r"))
```

vit2005

Apartments in Vitoria

## Description

Descriptive information and the appraised total price (in Euros) for apartments in Vitoria, Spain.

#### **Format**

A data frame with 218 observations on the following 16 variables:

row.labels the number of the observation

**totalprice** the market total price (in Euros) of the apartment including garage(s) and storage room(s) **area** the total living area of the apartment in square meters

**zone** a factor indicating the neighborhood where the apartment is located with levels Z11, Z21, Z31, Z32, Z34, Z35, Z36, Z37, Z38, Z41, Z42, Z43, Z44, Z45, Z46, Z47, Z48, Z49, Z52, Z53, Z56, Z61, and Z62.

70 Wait

**category** a factor indicating the condition of the apartment with levels 2A, 2B, 3A, 3B, 4A, 4B, and 5A. The factors are ordered so that 2A is the best and 5A is the worst.

age age of the aprtment

**floor** floor on which the apartment is located

rooms total number of rooms including bedrooms, dining room, and kitchen

**out** a factor indicating the percent of the apartment exposed to the elements. The levels E100, E75, E50, and E25, correspond to complete exposure, 75% exposure, 50% exposure, and 25% exposure respectively.

**conservation** is an ordered factor indicating the state of conservation of the apartment. The levels 1A, 2A, 2B, and 3A are ordered from best to worst conservation.

toilets the number of bathrooms

garage the number of garages

**elevator** indicates the absence (0) or presence (1) of elevators.

**streetcategory** an ordered factor from best to worst indicating the category of the street with levels S2, S3, S4, and S5

**heating** a factor indicating the type of heating with levels 1A, 3A, 3B, and 4A which correspond to: no heating, low-standard private heating, high-standard private heating, and central heating respectively.

tras the number of storage rooms outside of the apartment

#### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

#### **Examples**

```
modTotal <- lm(totalprice ~ area + as.factor(elevator) +
area:as.factor(elevator), data = vit2005)
modSimpl <- lm(totalprice ~ area, data = vit2005)
anova(modSimpl,modTotal)
rm(modSimpl, modTotal)</pre>
```

Wait

Waiting Time

### Description

A statistician records how long he must wait for his bus each morning. Data are used in Example 10.4.

Washer 71

### **Format**

A data frame with 15 observations on the following variable:

wt waiting time in minutes

### Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R*. Chapman & Hall/CRC.

## **Examples**

```
with(data = Wait,
wilcox.test(wt, mu = 6, alternative = "less"))
```

Washer

Washer Diameter

## Description

Diameter of washers.

#### **Format**

A data frame with 20 observations on the following variable:

diameters diameter of washer in cm

## Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
with(data = Washer,
EDA(diameters))
```

72 WCST

Water

Sodium Content of Water

## Description

An independent agency measures the sodium content in 20 samples from source X and in 10 samples from source Y and stores them in data frame Water. Data are used in Example 9.12.

#### **Format**

A data frame with 30 observations on the following 4 variables:

X sodium content measured in mg/L

Y sodium content measured in mg/L

Sodium combined sodium content measured in mg/L

Source a factor with levels X and Y

#### **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

### **Examples**

```
with(data = Water,
t.test(X, Y, alternative = "less"))
```

WCST

Wisconsin Card Sorting Test

## Description

The following data are the test scores from a group of 50 patients from the *Virgen del Camino* Hospital (Pamplona, Spain) on the Wisconsin Card Sorting Test.

#### **Format**

A data frame with 50 observations on the following variable:

score score on the Wisconsin Card Sorting Test

WeightGain 73

## **Details**

The "Wisconsin Card Sorting Test" is widely used by psychiatrists, neurologists, and neuropsychologists with patients who have a brain injury, neurodegenerative disease, or a mental illness such as schizophrenia. Patients with any sort of frontal lobe lesion generally do poorly on the test.

## Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

## **Examples**

```
densityplot(~score, data = WCST, ref = TRUE)
```

WeightGain

Weight Gain in Rats

# Description

The data come from an experiment to study the gain in weight of rats fed on four different diets, distinguished by amount of protein (low and high) and by source of pretein (beef and cereal).

## **Format**

A data frame with 40 observations on the following 4 variables.

ProteinSource a factor with levels Beef Cereal ProteinAmount a factor with levels High Low weightgain weight gain in grams

## **Details**

The design of the experiment is acompletely randomized with ten rats on each of the four treatments.

## Source

D. J. Hand, F. Daly, A. D. Lunn, K. J. McConway and E. Ostrowski (1994) *A Handbook of Small Datasets*. Chapman and Hall/CRC, London.

```
aov(weightgain ~ ProteinSource*ProteinAmount, data = WeightGain)
```

74 wheatUSA2004

WheatSpain

Wheat Surface Area in Spain

# Description

Seventeen Spanish communities and their corresponding surface area (in hecatares) dedicated to growing wheat

## **Format**

A data frame with 17 observations on the following 3 variables:

community a factor with levels Andalucia, Aragon, Asturias, Baleares, C. Valenciana, Canarias, Cantabria, Castilla-La Mancha, Castilla-Leon, Cataluna, Extremadura, Galicia, La Rioja, Madrid, Murcia, Navarra, and P. Vasco

hectares surface area measured in hectares

acres surface area measured in acres

## **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

# **Examples**

```
with(data = WheatSpain,
boxplot(hectares))
```

wheatUSA2004

USA Wheat Surface 2004

# **Description**

USA's 2004 harvested wheat surface by state

## **Format**

A data frame with 30 observations on the following 2 variables.

STATES a factor with levels AR, CA, CO, DE, GA, ID, IL, IN, KS, KY, MD, MI, MO, MS, MT, NC, NE, NY, OH, OK, OR, Other, PA, SC, SD, TN, TX, VA, WA, and WI

ACRES wheat surface area measured in 1000s of acres

wilcoxE.test 75

# Source

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

# **Examples**

```
with(data = wheatUSA2004,
hist(ACRES))
```

wilcoxE.test

Wilcoxon Exact Test

# Description

Performs exact one sample and two sample Wilcoxon tests on vectors of data

# Usage

```
wilcoxE.test(
    x,
    y = NULL,
    mu = 0,
    paired = FALSE,
    alternative = c("two.sided", "less", "greater"),
    conf.level = 0.95
)
```

# Arguments

х	is a numeric vector of data values. Non-finite (i.e. infinite or missing) values will be omitted.
У	an optional numeric vector of data values
mu	a number specifying an optional parameter used to form the null hypothesis
paired	a logical indicating whether you want a paired test
alternative	a character string specifying the alternative hypothesis, must be one of "two.sided" (default), "less", or "greater". You can specify just the initial letter.
conf.level	confidence level of the interval

76 wilcoxE.test

## **Details**

If only x is given, or if both x and y are given and paired = TRUE, a Wilcoxon signed rank test of the null hypothesis that the distribution of x (in the one sample case) or of x - y (in the paired two sample case) is symmetric about mu is performed.

Otherwise, if both x and y are given and paired = FALSE, a Wilcoxon rank sum test is done. In this case, the null hypothesis is that the distribution of x and y differ by a location shift mu, and the alternative is that they differ by some other location shift (and the one-sided alternative "greater" is that x is shifted to the right of y).

## Value

A list of class htest, containing the following components:

statistic	the value of the test statistic with a name describing it
p.value	the p-value for the test
null.value	the location parameter mu
alternative	a character string describing the alternative hypothesis
method	the type of test applied
data.name	a character string giving the names of the data
conf.int	a confidence interval for the location parameter
estimate	an estimate of the location parameter

# Note

The function is rather primitive and should only be used for problems with fewer than 19 observations as the memory requirements are rather large.

## Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

## References

- Gibbons, J.D. and Chakraborti, S. 1992. *Nonparametric Statistical Inference*. Marcel Dekker Inc., New York.
- Hollander, M. and Wolfe, D.A. 1999. *Nonparametric Statistical Methods*. New York: John Wiley & Sons.

## See Also

wilcox.test

Wool 77

## **Examples**

```
# Wilcoxon Signed Rank Test PH <- c(7.2, 7.3, 7.3, 7.4) wilcoxE.test(PH, mu = 7.25, alternative = "greater") # Wilcoxon Signed Rank Test (Dependent Samples) with(data = Aggression, wilcoxE.test(violence, noviolence, paired = TRUE, alternative = "greater")) # Wilcoxon Rank Sum Test x <- c(7.2, 7.2, 7.3, 7.3) y <- c(7.3, 7.3, 7.4, 7.4) wilcoxE.test(x, y) rm(PH, x, y)
```

Wool

Wool Production

# **Description**

Random sample of wool production in kilograms on 5 different days at two different locations

# **Format**

A data frame with 15 observations on the following 2 variables:

**textileA** wool production in thousands of kilograms

textileB wool production in thousands of kilograms

## **Source**

Ugarte, M. D., Militino, A. F., and Arnholt, A. T. (2008) *Probability and Statistics with R.* Chapman & Hall/CRC.

```
with(data = Wool,
t.test(textileA, textileB))
```

78 z.test

z.test *z-Test* 

# Description

This function is based on the standard normal distribution and creates confidence intervals and tests hypotheses for both one and two sample problems.

# Usage

```
z.test(
    x,
    sigma.x = NULL,
    y = NULL,
    sigma.y = NULL,
    sigma.d = NULL,
    alternative = c("two.sided", "less", "greater"),
    mu = 0,
    paired = FALSE,
    conf.level = 0.95,
    ...
)
```

# Arguments

X	a (non-empty) numeric vector of data values
sigma.x	a single number representing the population standard deviation for x
У	an optional (non-empty) numeric vector of data values
sigma.y	a single number representing the population standard deviation for y
sigma.d	a single number representing the population standard deviation for the paired differences
alternative	character string, one of "greater", "less", or "two.sided", or the initial letter of each, indicating the specification of the alternative hypothesis. For one-sample tests, alternative refers to the true mean of the parent population in relation to the hypothesized value mu. For the standard two-sample tests, alternative refers to the difference between the true population mean for $x$ and that for $y$ , in relation to mu.
mu	a single number representing the value of the mean or difference in means specified by the null hypothesis
paired	a logical indicating whether you want a paired z-test
conf.level	confidence level for the returned confidence interval, restricted to lie between zero and one
	Other arguments passed onto z.test()

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## **Details**

If y is NULL, a one-sample z-test is carried out with x provided sigma.x is not NULL. If y is not NULL, a standard two-sample z-test is performed provided both sigma.x and sigma.y are finite. If paired = TRUE, a paired z-test where the differences are defined as x - y is performed when the user enters a finite value for sigma.d (the population standard deviation for the differences).

## Value

A list of class htest, containing the following components:

statistic	the z-statistic, with names attribute z
p.value	the p-value for the test
conf.int	is a confidence interval (vector of length 2) for the true mean or difference in means. The confidence level is recorded in the attribute conf.level. When alternative is not "two.sided," the confidence interval will be half-infinite, to reflect the interpretation of a confidence interval as the set of all values k for which one would not reject the null hypothesis that the true mean or difference in means is k. Here, infinity will be represented by Inf.
estimate	vector of length 1 or 2, giving the sample mean(s) or mean of differences; these estimate the corresponding population parameters. Component estimate has a names attribute describing its elements.
null.value	the value of the mean or difference of means specified by the null hypothesis. This equals the input argument mu. Component null.value has a names attribute describing its elements.
alternative	records the value of the input argument alternative: "greater", "less", or "two.sided".
data.name	a character string (vector of length 1) containing the actual names of the input vectors $\boldsymbol{x}$ and $\boldsymbol{y}$

# **Null Hypothesis**

For the one-sample z-test, the null hypothesis is that the mean of the population from which x is drawn is mu. For the standard two-sample z-test, the null hypothesis is that the population mean for x less that for y is mu. For the paired z-test, the null hypothesis is that the mean difference between x and y is mu.

The alternative hypothesis in each case indicates the direction of divergence of the population mean for x (or difference of means for x and y) from mu (i.e., "greater", "less", or "two.sided").

## **Test Assumptions**

The assumption of normality for the underlying distribution or a sufficiently large sample size is required along with the population standard deviation to use Z procedures.

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## **Confidence Intervals**

For each of the above tests, an expression for the related confidence interval (returned component conf.int) can be obtained in the usual way by inverting the expression for the test statistic. Note that, as explained under the description of conf.int, the confidence interval will be half-infinite when alternative is not "two.sided"; infinity will be represented by Inf.

## Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

#### References

- Kitchens, L.J. 2003. Basic Statistics and Data Analysis. Duxbury.
- Hogg, R. V. and Craig, A. T. 1970. Introduction to Mathematical Statistics, 3rd ed. Toronto, Canada: Macmillan.
- Mood, A. M., Graybill, F. A. and Boes, D. C. 1974. *Introduction to the Theory of Statistics,* 3rd ed. New York: McGraw-Hill.
- Snedecor, G. W. and Cochran, W. G. 1980. *Statistical Methods*, 7th ed. Ames, Iowa: Iowa State University Press.

## See Also

```
zsum.test, tsum.test
```

```
with(data = Grocery, z.test(x = groceries, sigma.x = 30, conf.level = 0.97)$conf)
# Example 8.3 from PASWR.
x <- rnorm(12)
z.test(x, sigma.x = 1)
# Two-sided one-sample z-test where the assumed value for
# sigma.x is one. The null hypothesis is that the population
# mean for 'x' is zero. The alternative hypothesis states
# that it is either greater or less than zero. A confidence
# interval for the population mean will be computed.
x \leftarrow c(7.8, 6.6, 6.5, 7.4, 7.3, 7., 6.4, 7.1, 6.7, 7.6, 6.8)
y \leftarrow c(4.5, 5.4, 6.1, 6.1, 5.4, 5., 4.1, 5.5)
z.test(x, sigma.x=0.5, y, sigma.y=0.5, mu=2)
# Two-sided standard two-sample z-test where both sigma.x
# and sigma.y are both assumed to equal 0.5. The null hypothesis
# is that the population mean for 'x' less that for 'y' is 2.
# The alternative hypothesis is that this difference is not 2.
# A confidence interval for the true difference will be computed.
z.test(x, sigma.x = 0.5, y, sigma.y = 0.5, conf.level = 0.90)
# Two-sided standard two-sample z-test where both sigma.x and
# sigma.y are both assumed to equal 0.5. The null hypothesis
\# is that the population mean for 'x' less that for 'y' is zero.
# The alternative hypothesis is that this difference is not
# zero. A 90\% confidence interval for the true difference will
# be computed.
```

```
rm(x, y)
```

zsum.test

Summarized z-test

# **Description**

This function is based on the standard normal distribution and creates confidence intervals and tests hypotheses for both one and two sample problems based on summarized information the user passes to the function. Output is identical to that produced with z.test.

# Usage

```
zsum.test(
  mean.x,
  sigma.x = NULL,
  n.x = NULL,
  mean.y = NULL,
  sigma.y = NULL,
  n.y = NULL,
  alternative = c("two.sided", "less", "greater"),
  mu = 0,
  conf.level = 0.95,
  ...
)
```

# Arguments

mean.x	a single number representing the sample mean of x
sigma.x	a single number representing the population standard deviation for x
n.x	a single number representing the sample size for y
mean.y	a single number representing the sample mean of y
sigma.y	a single number representing the population standard deviation for y
n.y	a single number representing the sample size for y
alternative	is a character string, one of "greater", "less", or "two.sided", or the initial letter of each, indicating the specification of the alternative hypothesis. For one-sample tests, alternative refers to the true mean of the parent population in relation to the hypothesized value mu. For the standard two-sample tests, alternative refers to the difference between the true population mean for $x$ and that for $y$ , in relation to mu.
mu	a single number representing the value of the mean or difference in means specified by the null hypothesis
conf.level	confidence level for the returned confidence interval, restricted to lie between zero and one
	Other arguments passed onto z.test()

## **Details**

If y is NULL, a one-sample z-test is carried out with x provided sigma.x is finite. If y is not NULL, a standard two-sample z-test is performed provided both sigma.x and sigma.y are finite.

## Value

A list of class htest, containing the following components:

summarized samples.

the z-statistic, with names attribute z statistic p.value the p-value for the test conf.int is a confidence interval (vector of length 2) for the true mean or difference in means. The confidence level is recorded in the attribute conf.level. When alternative is not "two.sided," the confidence interval will be half-infinite, to reflect the interpretation of a confidence interval as the set of all values k for which one would not reject the null hypothesis that the true mean or difference in means is k. Here, infinity will be represented by Inf. estimate vector of length 1 or 2, giving the sample mean(s) or mean of differences; these estimate the corresponding population parameters. Component estimate has a names attribute describing its elements. null.value the value of the mean or difference in means specified by the null hypothesis. This equals the input argument mu. Component null.value has a names attribute describing its elements. records the value of the input argument alternative: "greater", "less", or alternative "two.sided".

# **Null Hypothesis**

data.name

For the one-sample z-test, the null hypothesis is that the mean of the population from which x is drawn is mu. For the standard two-sample z-test, the null hypothesis is that the population mean for x less that for y is mu.

a character string (vector of length 1) containing the names x and y for the two

The alternative hypothesis in each case indicates the direction of divergence of the population mean for x (or difference of means for x and y) from mu (i.e., "greater", "less", or "two.sided").

## **Test Assumptions**

The assumption of normality for the underlying distribution or a sufficiently large sample size is required along with the population standard deviation to use Z procedures.

## **Confidence Intervals**

For each of the above tests, an expression for the related confidence interval (returned component conf.int) can be obtained in the usual way by inverting the expression for the test statistic. Note that, as explained under the description of conf.int, the confidence interval will be half-infinite when alternative is not "two.sided"; infinity will be represented by Inf.

## Author(s)

Alan T. Arnholt <arnholtat@appstate.edu>

#### References

- Kitchens, L.J. 2003. Basic Statistics and Data Analysis. Duxbury.
- Hogg, R. V. and Craig, A. T. 1970. *Introduction to Mathematical Statistics, 3rd ed.* Toronto, Canada: Macmillan.
- Mood, A. M., Graybill, F. A. and Boes, D. C. 1974. Introduction to the Theory of Statistics, 3rd ed. New York: McGraw-Hill.
- Snedecor, G. W. and Cochran, W. G. 1980. *Statistical Methods*, 7th ed. Ames, Iowa: Iowa State University Press.

#### See Also

```
z.test, tsum.test
```

```
zsum.test(mean.x = 56/30,sigma.x = 2, n.x = 30, alternative="greater", mu = 1.8)
# Example 9.7 part a. from PASWR.
x <- rnorm(12)
zsum.test(mean(x), sigma.x = 1, n.x = 12)
# Two-sided one-sample z-test where the assumed value for
# sigma.x is one. The null hypothesis is that the population
# mean for 'x' is zero. The alternative hypothesis states
# that it is either greater or less than zero. A confidence
# interval for the population mean will be computed.
# Note: returns same answer as:
z.test(x, sigma.x = 1)
x \leftarrow c(7.8, 6.6, 6.5, 7.4, 7.3, 7.0, 6.4, 7.1, 6.7, 7.6, 6.8)
y \leftarrow c(4.5, 5.4, 6.1, 6.1, 5.4, 5.0, 4.1, 5.5)
zsum.test(mean(x), sigma.x = 0.5, n.x = 11, mean(y), sigma.y = 0.5, n.y = 8, mu = 2)
# Two-sided standard two-sample z-test where both sigma.x
# and sigma.y are both assumed to equal 0.5. The null hypothesis
# is that the population mean for 'x' less that for 'y' is 2.
# The alternative hypothesis is that this difference is not 2.
# A confidence interval for the true difference will be computed.
# Note: returns same answer as:
z.test(x, sigma.x = 0.5, y, sigma.y = 0.5)
zsum.test(mean(x), sigma.x = 0.5, n.x = 11, mean(y), sigma.y = 0.5, n.y = 8,
conf.level=0.90)
# Two-sided standard two-sample z-test where both sigma.x and
# sigma.y are both assumed to equal 0.5. The null hypothesis
# is that the population mean for 'x' less that for 'y' is zero.
# The alternative hypothesis is that this difference is not
# zero. A 90% confidence interval for the true difference will
```

```
# be computed. Note: returns same answer as: z.test(x, sigma.x=0.5, y, sigma.y=0.5, conf.level=0.90) rm(x, y)
```

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