

# Package ‘fastR2’

May 8, 2026

**Type** Package

**Title** Foundations and Applications of Statistics Using R (2nd Edition)

**Version** 1.2.5

**Description** Data sets and utilities to accompany the second edition of “Foundations and Applications of Statistics: an Introduction using R” (R Pruim, published by AMS, 2017), a text covering topics from probability and mathematical statistics at an advanced undergraduate level. R is integrated throughout, and access to all the R code in the book is provided via the `snippet()` function.

**License** GPL (>= 2)

**Depends** R (>= 3.0.0), mosaic (>= 1.3.0)

**Imports** maxLik, numDeriv, dplyr, ggplot2 (>= 3.0.0), lattice, miscTools

**Suggests** ggformula, mosaicCalc, tidyr, readr, MASS, faraway, Hmisc, DAAG, multcomp, vcd, car, alr4, corrgram, BradleyTerry2, cubature, knitr, mosaicData, rmarkdown

**VignetteBuilder** knitr

**LazyLoad** yes

**LazyData** yes

**RoxygenNote** 7.3.2

**Encoding** UTF-8

**URL** <https://github.com/rpruim/fastR2>, <http://rpruim.github.io/fastR2/>

**BugReports** <https://github.com/rpruim/fastR2/issues>

**NeedsCompilation** no

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**Repository** CRAN

**Date/Publication** 2025-07-30 16:30:15 UTC

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---

 ACTgpa

*ACT scores and GPA*


---

**Description**

ACT scores and college GPA for a small sample of college students.

**Format**

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

**ACT** ACT score

**GPA** GPA

**Examples**

```
gf_point(GPA ~ ACT, data = ACTgpa)
```

---

 AirlineArrival

*Airline On-Time Arrival Data*


---

### Description

Flights categorized by destination city, airline, and whether or not the flight was on time.

### Format

A data frame with 11000 observations on the following 3 variables.

**airport** a factor with levels LosAngeles, Phoenix, SanDiego, SanFrancisco, Seattle

**result** a factor with levels Delayed, OnTime

**airline** a factor with levels Alaska, AmericaWest

### Source

Barnett, Arnold. 1994. "How numbers can trick you." *Technology Review*, vol. 97, no. 7, pp. 38–45.

### References

These and similar data appear in many text books under the topic of Simpson's paradox.

### Examples

```
tally(
  airline ~ result, data = AirlineArrival,
  format = "perc", margins = TRUE)
tally(
  result ~ airline + airport,
  data = AirlineArrival, format = "perc", margins = TRUE)
AirlineArrival2 <-
  AirlineArrival %>%
  group_by(airport, airline, result) %>%
  summarise(count = n()) %>%
  group_by(airport, airline) %>%
  mutate(total = sum(count), percent = count/total * 100) %>%
  filter(result == "Delayed")
AirlineArrival3 <-
  AirlineArrival %>%
  group_by(airline, result) %>%
  summarise(count = n()) %>%
  group_by(airline) %>%
  mutate(total = sum(count), percent = count/total * 100) %>%
  filter(result == "Delayed")
gf_line(percent ~ airport, color = ~ airline, group = ~ airline,
  data = AirlineArrival2) %>%
  gf_point(percent ~ airport, color = ~ airline, size = ~total,
```

```
data = AirlineArrival2) %>%  
gf_hline(yintercept = ~ percent, color = ~airline,  
data = AirlineArrival3, linetype = "dashed") %>%  
gf_labs(y = "percent delayed")
```

---

AirPollution

*Air pollution measurements*

---

### Description

Air pollution measurements at three locations.

### Format

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

**pollution** a numeric vector

**location** a factor with levels Hill Suburb, Plains Suburb, Urban City

### Source

David J. Saville and Graham R. Wood, *Statistical methods: A geometric primer*, Springer, 1996.

### Examples

```
data(AirPollution)  
summary(lm(pollution ~ location, data = AirPollution))
```

---

BallDrop

*Ball dropping data*

---

### Description

Undergraduate students in a physics lab recorded the height from which a ball was dropped and the time it took to reach the floor.

### Format

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

**height** height in meters

**time** time in seconds

### Source

Steve Plath, Calvin College Physics Department

**Examples**

```
gf_point(time ~ height, data = BallDrop)
```

---

Batting

*Major League Batting 2000-2005*

---

**Description**

Major League batting data for the seasons from 2000-2005.

**Format**

A data frame with 8062 observations on the following 22 variables.

**player** unique identifier for each player

**year** year

**stint** for players who were traded mid-season, indicates which portion of the season the data cover

**team** three-letter code for team

**league** a factor with levels AA AL NL

**G** games

**AB** at bats

**R** runs

**H** hits

**H2B** doubles

**H3B** triples

**HR** home runs

**RBI** runs batted in

**SB** stolen bases

**CS** caught stealing

**BB** bases on balls (walks)

**SO** strike outs

**IBB** intentional base on balls

**HBP** hit by pitch

**SH** a numeric vector

**SF** sacrifice fly

**GIDP** grounded into double play

**Examples**

```
data(Batting)
gf_histogram(~ HR, data = Batting)
```

---

Buckthorn

*Buckthorn*

---

### Description

Data from an experiment to determine the efficacy of various methods of eradicating buckthorn, an invasive woody shrub. Buckthorn plants were chopped down and the stumps treated with various concentrations of glyphosate. The next season, researchers returned to see whether the plant had regrown.

### Format

A data frame with 165 observations on the following 3 variables.

**shoots** number of new shoots coming from stump

**conc** concentration of glyphosate applied

**dead** whether the stump was considered dead

### Source

David Dornbos, Calvin College

### Examples

```
data(Buckthorn)
```

---

Bugs

*Bugs*

---

### Description

This data frame contains data from an experiment to see if insects are more attracted to some colors than to others. The researchers prepared colored cards with a sticky substance so that insects that landed on them could not escape. The cards were placed in a field of oats in July. Later the researchers returned, collected the cards, and counted the number of cereal leaf beetles trapped on each card.

### Format

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

**color** color of card; one of B(lue) G(reen) W(hite) Y(ellow)

**trapped** number of insects trapped on the card

**Source**

M. C. Wilson and R. E. Shade, Relative attractiveness of various luminescent colors to the cereal leaf beetle and the meadow spittlebug, *Journal of Economic Entomology* 60 (1967), 578–580.

**Examples**

```
data(Bugs)
favstats(trapped ~ color, data = Bugs)
```

---

`col.fastR`*Lattice Theme*

---

**Description**

A theme for use with lattice graphics.

**Usage**

```
col.fastR(bw = FALSE, lty = 1:7)
```

**Arguments**

<code>bw</code>	whether color scheme should be "black and white"
<code>lty</code>	vector of line type codes

**Value**

Returns a list that can be supplied as the theme to `[lattice::trellis.par.set()]`.

**Note**

This theme was used in the production of the book *Foundations and Applications of Statistics*

**Author(s)**

Randall Pruim

**See Also**

`[lattice::trellis.par.set()]`, `[lattice::show.settings()]`

**Examples**

```
trellis.par.set(theme=col.fastR(bw=TRUE))
show.settings()
trellis.par.set(theme=col.fastR())
show.settings()
```

---

`col.perc`*Row and Column Percentages*

---

**Description**

Convenience wrappers around `apply()` to compute row and column percentages of matrix-like structures, including output of `xtabs`.

**Usage**`col.perc(x)``row.perc(x)`**Arguments**

`x` matrix-like structure

**Author(s)**

Randall Pruim

**Examples**

```
row.perc(tally(~ airline + result, data = AirlineArrival))
col.perc(tally(~ airline + result, data = AirlineArrival))
```

---

`Concrete`*Concrete Compressive Strength Data*

---

**Description**

These data were collected by I-Cheng Yeh to determine how the compressive strength of concrete is related to its ingredients (cement, blast furnace slag, fly ash, water, superplasticizer, coarse aggregate, and fine aggregate) and age.

**Format**

Concrete is a data frame with the following variables.

**limestone** percentage of limestone

**water** water-cement ratio

**strength** compressive strength (MPa) after 28 days

## References

Appeared in Devore's "Probability and Statistics for Engineers and the Sciences (6th ed). The variables have been renamed.

---

ConcreteAll

#' Concrete Compressive Strength Data

---

## Description

These data were collected by I-Cheng Yeh to determine how the compressive strength of concrete is related to its ingredients (cement, blast furnace slag, fly ash, water, superplasticizer, coarse aggregate, and fine aggregate) and age.

## Format

concreteAll is a data frame with the following 9 variables.

**cement** amount of cement (kg/m<sup>3</sup>)

**slag** amount of blast furnace slag (kg/m<sup>3</sup>)

**ash** amount of fly ash(kg/m<sup>3</sup>)

**water** amount of water (kg/m<sup>3</sup>)

**superP** amount of superplasticizer (kg/m<sup>3</sup>)

**coarseAg** amount of coarse aggregate (kg/m<sup>3</sup>)

**fineAg** amount of fine aggregate (kg/m<sup>3</sup>)

**age** age of concrete in days

**strength** compressive strength measured in MPa

Concrete is a subset of ConcreteAll.

## Source

Data were obtained from the Machine Learning Repository (<https://archive.ics.uci.edu/ml/>) where they were deposited by I-Cheng Yeh (<icyeh@chu.edu.tw>) who retains the copyright for these data.

## References

I-Cheng Yeh (1998), "Modeling of strength of high performance concrete using artificial neural networks," *Cement and Concrete Research*, Vol. 28, No. 12, pp. 1797-1808.

## Examples

```
data(Concrete)
```

---

CoolingWater	<i>Cooling Water</i>
--------------	----------------------

---

**Description**

Temperature of a mug of water as it cools.

**Usage**

```
data(CoolingWater1)
```

```
data(CoolingWater2)
```

```
data(CoolingWater3)
```

```
data(CoolingWater4)
```

**Format**

A data frame with the following variables.

time time in seconds

temp temperature in Celsius (CoolingWater1, CoolingWater2) or Fahrenheit (CoolingWater3, CoolingWater4)

**Source**

These data were collected by Stan Wagon and his students at Macelester College to explore Newton's Law of Cooling and the ways that the law fails to capture all of the physics involved in cooling water. CoolingWater1 and CoolingWater2 appeared in a plot in Wagon (2013) and were (approximately) extracted from the plot. CoolingWater3 and CoolingWater4 appeared in a plot in Wagon (2005). The data in CoolingWater2 and CoolingWater4 were collected with a film of oil on the surface of the water to minimize evaporation.

**References**

- R. Portmann and S. Wagon. "How quickly does hot water cool?" *Mathematica in Education and Research*, 10(3):1-9, July 2005.
- R. Israel, P. Saltzman, and S. Wagon. "Cooling coffee without solving differential equations". *Mathematics Magazine*, 86(3):204-210, 2013.

**Examples**

```
data(CoolingWater1)
data(CoolingWater2)
data(CoolingWater3)
data(CoolingWater4)
if (require(ggformula)) {
```

```

gf_line(
  temp ~ time, color = ~ condition,
  data = rbind(CoolingWater1, CoolingWater2))
}
if (require(ggformula)) {
  gf_line(
    temp ~ time, color = ~ condition,
    data = rbind(CoolingWater3, CoolingWater4))
}

```

---

Corn

*Corn Yield*

---

### Description

William Gosset analyzed data from an experiment comparing the yield of regular and kiln-dried corn.

### Format

A data frame with 11 observations on the following 2 variables.

**reg** yield of regular corn (lbs/acre)

**kiln** yield of kiln-dried corn (lbs/acre)

### Details

Gosset (Student) reported on the results of seeding plots with two different kinds of seed. Each type of seed (regular and kiln-dried) was planted in adjacent plots, accounting for 11 pairs of "split" plots.

### Source

These data are also available at DASL, the data and story library (<https://dasl.datadescription.com/>).

### References

W.S. Gosset, "The Probable Error of a Mean," *Biometrika*, 6 (1908), pp 1-25.

### Examples

```

Corn2 <- stack(Corn)
names(Corn2) <- c('yield', 'treatment')
lm(yield ~ treatment, data = Corn2)
t.test(yield ~ treatment, data = Corn2)
t.test(Corn$reg, Corn$kiln)

```

---

Cuckoo *Cuckoo eggs in other birds' nests*

---

### Description

Cuckoos are known to lay their eggs in the nests of other (host) birds. The eggs are then adopted and hatched by the host birds. These data were originally collected by O. M. Latter in 1902 to see how the size of a cuckoo egg is related to the species of the host bird.

### Format

A data frame with 120 observations on the following 2 variables.

**length** length of egg (mm)

**species** a factor with levels hedge sparrow meadow pipet pied wagtail robin tree pipet wren

### Source

L.H.C. Tippett, *The Methods of Statistics*, 4th Edition, John Wiley and Sons, Inc., 1952, p. 176.

### References

These data are also available from DASL, the data and story library (<https://dasl.datadescription.com/>).

### Examples

```
data(Cuckoo)
gf_boxplot(length ~ species, data = Cuckoo)
```

---

DeathPenalty *Death Penalty and Race*

---

### Description

A famous example of Simpson's paradox.

### Format

A data frame with 326 observations.

**id** a subject id

**victim** a factor with levels B1 Wh

**defendant** a factor with levels B1, Wh

**death** a factor with levels Yes, No

**penalty** a factor with levels death other

**Source**

Radelet, M. (1981). Racial characteristics and imposition of the death penalty. *American Sociological Review*, 46:918–927.

**Examples**

```
tally(penalty ~ defendant, data = DeathPenalty)
tally(penalty ~ defendant + victim, data = DeathPenalty)
```

---

 Drag

---

*Drag force experiment*


---

**Description**

The data come from an experiment to determine how terminal velocity depends on the mass of the falling object. A helium balloon was rigged with a small basket and just the ballast to make it neutrally buoyant. Mass was then added and the terminal velocity is calculated by measuring the time it took to fall between two sensors once terminal velocity has been reached. Larger masses were drop from higher heights and used sensors more widely spaced.

**Format**

A data frame with 42 observations on the following 5 variables.

**time** time (in seconds) to travel between two sensors

**mass** net mass (in kg) of falling object

**height** distance (in meters) between two sensors

**velocity** average velocity (in m/s) computed from time and height

**force.drag** calculated drag force (in N,  $\text{force.drag} = \text{mass} * 9.8$ ) using the fact that at terminal velocity, the drag force is equal to the force of gravity

**Source**

Calvin College physics students under the supervision of Professor Steve Plath.

**Examples**

```
data(Drag)
with(Drag, force.drag / mass)
gf_point(velocity ~ mass, data = Drag)
```

---

Endurance

*Endurance and vitamin C*

---

### Description

The effect of a single 600 mg dose of ascorbic acid versus a sugar placebo on the muscular endurance (as measured by repetitive grip strength trials) of fifteen male volunteers (19-23 years old).

### Format

A data frame with 15 observations on the following 5 variables.

**vitamin** number of repetitions until reaching 50 maximal grip after taking vitamin

**first** which treatment was done first, a factor with levels Placebo Vitamin

**placebo** number of repetitions until reaching 50 strength after taking placebo

### Details

Three initial maximal contractions were performed for each subject, with the greatest value indicating maximal grip strength. Muscular endurance was measured by having the subjects squeeze the dynamometer, hold the contraction for three seconds, and repeat continuously until a value of 50 maximum grip strength was achieved for three consecutive contractions. Endurance was defined as the number of repetitions required to go from maximum grip strength to the initial 50 positive verbal encouragement in an effort to have them complete as many repetitions as possible.

The study was conducted in a double-blind manner with crossover.

### Source

These data are available from OzDASL, the Australasian data and story library (<https://dasl.datadescription.com/>).

### References

Keith, R. E., and Merrill, E. (1983). The effects of vitamin C on maximum grip strength and muscular endurance. *Journal of Sports Medicine and Physical Fitness*, 23, 253-256.

### Examples

```
data(Endurance)
t.test(Endurance$vitamin, Endurance$placebo, paired = TRUE)
t.test(log(Endurance$vitamin), log(Endurance$placebo), paired = TRUE)
t.test(1/Endurance$vitamin, 1/Endurance$placebo, paired = TRUE)
gf_qq( ~ vitamin - placebo, data = Endurance)
gf_qq( ~ log(vitamin) - log(placebo), data = Endurance)
gf_qq( ~ 1/vitamin - 1/placebo, data = Endurance)
```

---

FamilySmoking	<i>Family smoking</i>
---------------	-----------------------

---

### Description

A cross-tabulation of whether a student smokes and how many of his or her parents smoke from a study conducted in the 1960's.

### Format

A data frame with 5375 observations on the following 2 variables.

**student** a factor with levels DoesNotSmoke Smokes

**parents** a factor with levels NeitherSmokes OneSmokesBothSmoke

### Source

S. V. Zagona (ed.), *Studies and issues in smoking behavior*, University of Arizona Press, 1967.

### References

The data also appear in

Brigitte Baldi and David S. Moore, *The Practice of Statistics in the Life Sciences*, Freeman, 2009.

### Examples

```
data(FamilySmoking)
xchisq.test( tally(parents ~ student, data = FamilySmoking) )
```

---

Fumbles	<i>NCAA football fumbles</i>
---------	------------------------------

---

### Description

This data frame gives the number of fumbles by each NCAA FBS team for the first three weeks in November, 2010.

**Format**

A data frame with 120 observations on the following 7 variables.

**team** NCAA football team

**rank** rank based on fumbles per game through games on November 26, 2010

**W** number of wins through games on November 26, 2010

**L** number of losses through games on November 26, 2010

**week1** number of fumbles on November 6, 2010

**week2** number of fumbles on November 13, 2010

**week3** number of fumbles on November 20, 2010

**Details**

The fumble counts listed here are total fumbles, not fumbles lost. Some of these fumbles were recovered by the team that fumbled.

**Source**

<https://www.teamrankings.com/college-football/stat/fumbles-per-game>

**Examples**

```
data(Fumbles)
m <- max(Fumbles$week1)
table(factor(Fumbles$week1, levels = 0:m))
favstats( ~ week1, data = Fumbles)
# compare with Poisson distribution
cbind(
  fumbles = 0:m,
  observedCount = table(factor(Fumbles$week1, levels = 0:m)),
  modelCount = 120 * dpois(0:m, mean(Fumbles$week1)),
  observedPct = table(factor(Fumbles$week1, levels = 0:m))/120,
  modelPct = dpois(0:m, mean(Fumbles$week1))
) %>% signif(3)
showFumbles <- function(x, lambda = mean(x), ...) {
  result <-
    gf_dhistogram( ~ week1, data = Fumbles, binwidth = 1, alpha = 0.3) %>%
    gf_dist("pois", lambda = mean( ~ week1, data = Fumbles) )
  print(result)
  return(result)
}
showFumbles(Fumbles$week1)
showFumbles(Fumbles$week2)
showFumbles(Fumbles$week3)
```

---

`geom`*Geometric representation of linear model*

---

**Description**

`geom` create a graphical representation of the fit of a linear model.

**Usage**

```
geom(formula, data = parent.env(), type = "xz", version = 1, plot = TRUE, ...)
to2d(x, y, z, type = NULL, xas = c(0.4, -0.3), yas = c(1, 0), zas = c(0, 1))
```

**Arguments**

<code>formula</code>	a formula as used in <a href="#">lm</a> .
<code>data</code>	a data frame as in <a href="#">lm</a> .
<code>type</code>	character: indicating the type of projection to use to collapse multi-dimensional data space into two dimensions of the display.
<code>version</code>	an integer (currently 1 or 2). Which version of the plot to display.
<code>plot</code>	a logical: should the plot be displayed?
<code>...</code>	other arguments passed to <a href="#">lm</a>
<code>x, y, z</code>	numeric.
<code>xas, yas, zas</code>	numeric vector of length 2 indicating the projection of $c(1, 0, 0)$ , $c(0, 1, 0)$ , and $c(0, 0, 1)$ .

**Author(s)**

Randall Pruim

**See Also**

[lm](#).

**Examples**

```
geom(pollution ~ location, data = AirPollution)
geom(distance ~ projectileWt, data = Trebuchet2)
```

---

`givenOrder`*Create ordered factor with order inferred from order given*

---

**Description**

The order of the resulting factor is determined by the order in which unique labels first appear in the vector or factor `x`.

**Usage**

```
givenOrder(x)
```

**Arguments**

`x` a vector or factor to be converted into an ordered factor.

**Examples**

```
givenOrder(c("First", "Second", "Third", "Fourth", "Fifth", "Sixth"))
```

---

`golfballs`*Golf ball numbers*

---

**Description**

Allan Rossman used to live on a golf course in a spot where dozens of balls would come into his yard every week. He collected the balls and eventually tallied up the numbers on the first 5000 golf balls he collected. Of these 486 bore the number 1, 2, 3, or 4. The remaining 14 golf balls were omitted from the data.

**Format**

The format is: num [1:4] 137 138 107 104

**Source**

Data collected by Allan Rossman in Carlisle, PA.

**Examples**

```
data(golfballs)
golfballs/sum(golfballs)
chisq.test(golfballs, p = rep(.25,4))
```

---

 GoosePermits

*Goose permits*


---

## Description

In a 1979 study by Bishop and Heberlein, 237 hunters were each offered one of 11 cash amounts (bids) ranging from \$1 to \$200 in return for their hunting permits. The data records how many hunters offered each bid kept or sold their permit.

## Format

A data frame with 11 rows and 5 columns. Each row corresponds to a bid (in US dollars) offered for a goose permit. The columns `keep` and `sell` indicate how many hunters offered that bid kept or sold their permit, respectively. `n` is the sum of `keep` and `sell` and `prop_sell` is the proportion that sold.

## References

Bishop and Heberlein (Amer. J. Agr. Econ. 61, 1979).

## Examples

```
goose.mod <- glm( cbind(sell, keep) ~ log(bid), data = GoosePermits, family = binomial())
gf_point(0 ~ bid, size = ~keep, color = "gray50", data = GoosePermits) %>%
  gf_point(1 ~ bid, size = ~sell, color = "navy") %>%
  gf_function(fun = makeFun(goose.mod)) %>%
  gf_refine(guides(size = "none"))

ggplot(data = GoosePermits) +
  geom_point( aes(x = bid, y = 0, size = keep), colour = "gray50") +
  geom_point( aes(x = bid, y = 1, size = sell), colour = "navy") +
  stat_function(fun = makeFun(goose.mod)) +
  guides( size = "none")

gf_point( (sell / (sell + keep)) ~ bid, data = GoosePermits,
  size = ~sell + keep, color = "navy") %>%
  gf_function(fun = makeFun(goose.mod)) %>%
  gf_text(label = ~as.character(sell + keep), colour = "white", size = 3) %>%
  gf_refine(scale_size_area()) %>%
  gf_labs(y = "probability of selling")

ggplot(data = GoosePermits) +
  stat_function(fun = makeFun(goose.mod)) +
  geom_point( aes(x = bid, y = sell / (sell + keep), size = sell + keep), colour = "navy") +
  geom_text( aes(x = bid, y = sell / (sell + keep), label = as.character(sell + keep)),
  colour = "white", size = 3) +
  scale_size_area() +
  labs(y = "probability of selling")
```

---

GPA

*GPA, ACT, and SAT scores*

---

### Description

GPA, ACT, and SAT scores for a sample of students.

### Format

A data frame with 271 observations on the following 4 variables.

**act** ACT score

**gpa** college grade point average

**satm** SAT mathematics score

**satv** SAT verbal score

### Examples

```
data(GPA)
splom(GPA)
```

---

HeliumFootballs

*Punting helium- and air-filled footballs*

---

### Description

Two identical footballs, one air-filled and one helium-filled, were used outdoors on a windless day at The Ohio State University's athletic complex. Each football was kicked 39 times and the two footballs were alternated with each kick. The experimenter recorded the distance traveled by each ball.

### Format

A data frame with 39 observations on the following 3 variables.

**trial** trial number

**air** distance traveled by air-filled football (yards)

**helium** distance traveled by helium-filled football (yards)

### Source

These data are available from DASL, the data and story library (<https://dasl.datadescription.com/>).

## References

Lafferty, M. B. (1993), "OSU scientists get a kick out of sports controversy", *The Columbus Dispatch* (November, 21, 1993), B7.

## Examples

```
data(HeliumFootballs)
gf_point(helium ~ air, data = HeliumFootballs)
gf_dhistogram(
  ~ (helium - air), data = HeliumFootballs,
  fill = ~ (helium > air), bins = 15, boundary = 0
)
```

---

Ice

*Cooling muscles with ice*

---

## Description

This data set contains the results of an experiment comparing the efficacy of different forms of dry ice application in reducing the temperature of the calf muscle.

## Details

The 12 subjects in this study came three times, at least four days apart, and received one of three ice treatments (cubed ice, crushed ice, or ice mixed with water). In each case, the ice was prepared in a plastic bag and applied dry to the subjects calf muscle. The temperature measurements were taken on the skin surface and inside the calf muscle (via a 4 cm long probe) every 30 seconds for 20 minutes prior to icing, for 20 minutes during icing, and for 2 hours after the ice had been removed. The temperature measurements are stored in variables that begin with b (baseline), t (treatment), or r (recovery) followed by a numerical code for the elapsed time formed by concatenating the number of minutes and seconds. For example, R1230 contains the temperatures 12 minutes and 30 seconds after the ice had been removed.

Variables include

**Subject** identification number

**sex** a factor with levels female male

**weight** weight of subject (kg)

**Height** height of subject (cm)

**Skinfold** skinfold thickness

**calf** calf diameter (cm)

**Age** age of subject

**location** a factor with levels intramuscular surface

**Treatment** a factor with levels crushed cubed wet

**B0** baseline temperature at time 0  
**b30** baseline temperature 30 seconds after start  
**b100** baseline temperature 1 minute after start  
**b1930** baseline temperature 19 minutes 30 seconds start  
**t0** treatment temperature at beginning of treatment  
**t30** treatment temperature 30 seconds after start of treatment  
**t100** treatment temperature 1 minute after start of treatment  
**t1930** treatment temperature 19 minutes 30 seconds after start of treatment  
**R0** recovery temperature at start of recovery  
**r30** recovery temperature 30 seconds after start of recovery  
**r100** recovery temperature 1 minute after start of recovery  
**r12000** recovery temperature 120 minutes after start of recovery

### Source

Dykstra, J. H., Hill, H. M., Miller, M. G., Michael T. J., Cheatham, C. C., and Baker, R.J., Comparisons of cubed ice, crushed ice, and wetted ice on intramuscular and surface temperature changes, *Journal of Athletic Training* 44 (2009), no. 2, 136–141.

### Examples

```

data(Ice)
gf_point(weight ~ skinfold, color = ~ sex, data = Ice)
if (require(readr) && require(tidyr)) {
  Ice2 <- Ice %>%
    gather("key", "temp", b0:r12000) %>%
    separate(key, c("phase", "time"), sep = 1) %>%
    mutate(time = parse_number(time), subject = as.character(subject))

  gf_line( temp ~ time, data = Ice2 %>% filter(phase == "t"),
          color = ~ sex, group = ~subject, alpha = 0.6) %>%
    gf_facet_grid( treatment ~ location)
}

```

---

Inflation

*Inflation data*

---

### Description

The article developed four measures of central bank independence and explored their relation to inflation outcomes in developed and developing countries. This datafile deals with two of these measures in 23 nations.

**Format**

A data frame with 23 observations on the following 5 variables.

**country** country where data were collected

**ques** questionnaire index of independence

**inf** annual inflation rate, 1980-1989 (percent)

**legal** legal index of independence

**dev** developed (1) or developing (2) nation

**Source**

These data are available from OzDASL, the Australasian Data and Story Library (<https://dasl.datadescription.com/>).

**References**

A. Cukierman, S.B. Webb, and B. Negapi, "Measuring the Independence of Central Banks and Its Effect on Policy Outcomes," World Bank Economic Review, Vol. 6 No. 3 (Sept 1992), 353-398.

**Examples**

```
data(Inflation)
```

---

information

*Information*

---

**Description**

Extract information from a maxLik object

**Usage**

```
information(object, ...)
```

**Arguments**

**object** an object of class "maxLik".

**...** additional arguments

---

 Jordan8687

*Michael Jordan personal scoring*


---

**Description**

The number of points scored by Michael Jordan in each game of the 1986-87 regular season.

**Format**

A data frame with 82 observations on the following 2 variables.

**game** a numeric vector

**points** a numeric vector

**Examples**

```
data(Jordan8687)
gf_qq(~ points, data = Jordan8687)
```

---

Kids

*Goals and popularity factors for school kids*


---

**Description**

Subjects were students in grades 4-6 from three school districts in Michigan. Students were selected from urban, suburban, and rural school districts with approximately 1/3 of their sample coming from each district. Students indicated whether good grades, athletic ability, or popularity was most important to them. They also ranked four factors: grades, sports, looks, and money, in order of their importance for popularity. The questionnaire also asked for gender, grade level, and other demographic information.

**Format**

A data frame with 478 observations on the following 11 variables.

**gender** a factor with levels boy girl

**grade** grade in school

**age** student age

**race** a factor with levels other White

**urban.rural** a factor with levels Rural Suburban Urban

**school** a factor with levels Brentwood Elementary Brentwood Middle Brown Middle Elm Main Portage Ridge Sand Westdale Middle

**goals** a factor with levels Grades Popular Sports

**grades** rank of ‘make good grades’ (1 = most important for popularity; 4 = least important)  
**sports** rank of ‘being good at sports’ (1 = most important for popularity; 4 = least important)  
**looks** rank of ‘being handsome or pretty’ (1 = most important for popularity; 4 = least important)  
**money** rank of ‘having lots of money’ (1 = most important for popularity; 4 = least important)

### Source

These data are available at DASL, the data and story library (<https://dasl.datadescription.com/>).

### References

Chase, M. A., and Dummer, G. M. (1992), "The Role of Sports as a Social Determinant for Children," *Research Quarterly for Exercise and Sport*, 63, 418-424.

### Examples

```
data(Kids)
tally(goals ~ urban.rural, data = Kids)
chisq.test(tally(~ goals + urban.rural, data = Kids))
```

---

LittleSurvey

*Results from a little survey*

---

### Description

These data are from a little survey given to a number of students in introductory statistics courses. Several of the items were prepared in multiple versions and distributed randomly to the students.

### Format

A data frame with 279 observations on the following 20 variables.

**number** a number between 1 and 30

**colorver** which version of the ‘favorite color’ question was on the survey. A factor with levels v1 v2

**color** favorite color if among predefined choices. A factor with levels black green other purple red

**othercolor** favorite color if not among choices above.

**animalver** which version of the ‘favorite color’ question was on the survey. A factor with levels v1 v2

**animal** favorite animal if among predefined choices. A factor with levels elephant giraffe lion other.

**otheranimal** favorite animal if not among the predefined choices.

**pulsever** which version of the 'pulse' question was on the survey  
**pulse** self-reported pulse  
**tvver** which of three versions of the TV question was on the survey  
**tvbox** a factor with levels <1 >4 >8 1-2 2-4 4-8 none other  
**tvhours** a numeric vector  
**surprisever** which of two versions of the 'surprise' question was on the survey  
**surprise** a factor with levels no yes  
**playver** which of two versions of the 'play' question was on the survey  
**play** a factor with levels no yes  
**diseasever** which of two versions of the 'play' question was on the survey  
**disease** a factor with levels A B  
**homeworkver** which of two versions of the 'homework' question was on the survey  
**homework** a factor with levels A B

### Question Wording

- 1.1. Write down any number between 1 and 30 (inclusive).
- 2.1. What is your favorite color? Choices: black red; green; purple; other
- 2.2. What is your favorite color?
- 3.1. What is your favorite zoo animal? Choices: giraffe; lion; elephant; other
- 3.2. What is your favorite zoo animal?
- 4.1. Measure and record your pulse.
- 5.1. How much time have you spent watching TV in the last week?
- 5.2. How much time have you spent watching TV in the last week? Choices: none; under 1; hour 1-2 hours; 2-4 hours; more than 4 hours
- 5.3. How much time have you spent watching TV in the last week? Choices: under 1 hour; 1-2 hours; 2-4 hours; 4-8 hours; more than 8 hours
- 6.1. Social science researchers have conducted extensive empirical studies and concluded that the expression "absence makes the heart grow fonder" is generally true. Do you find this result surprising or not surprising?
- 6.2. Social science researchers have conducted extensive empirical studies and concluded that the expression "out of sight out of mind" is generally true. Do you find this result surprising or not surprising?
- 7.1. Suppose that you have decided to see a play for which the admission charge is \$20 per ticket. As you prepare to purchase the ticket, you discover that you have lost a \$20 bill. Would you still pay \$20 for a ticket to see the play?
- 7.2. Suppose that you have decided to see a play for which the admission charge is \$20 per ticket. As you prepare to enter the theater, you discover that you have lost your ticket. Would you pay \$20 to buy a new ticket to see the play?
- 8.1. suppose that the United States is preparing for the outbreak of an unusual Asian disease that is expected to kill 600 people. Two alternative programs to combat the disease have been proposed.

Assume that the exact scientific estimates of the consequences of the programs are as follows: If program A is adopted, 200 people will be saved. If program B is adopted, there is a  $1/3$  probability that 600 people will be saved and a  $2/3$  probability that nobody will be saved. Which of the two programs would you favor?

8.2. Suppose that the United States is preparing for the outbreak of an unusual Asian disease that is expected to kill 600 people. two alternative programs to combat the disease have been proposed. Assume that the exact scientific estimates of the consequences of the programs are as follows:

If program A is adopted, 400 people will die. If program B is adopted, there is a  $1/3$  probability that no one will die and a  $2/3$  probability that all 600 people will die. Which of the two programs would you favor? A or B

9.1. A national survey of college students revealed that professors at this college assign "significantly more homework than the nationwide average for an institution of its type." How does this finding compare with your experience? Choices: a. That sounds about right to me; b that doesn't sound right to me.

9.2. A national survey of college students revealed that professors at this college assign an amount of homework that "is fairly typical for institutions of its type." How does this finding compare with your experience? Choices: A that sounds about right to me; b that doesn't sound right to me.

### Examples

```
data(LittleSurvey)
tally(surprise ~ surprisever, data = LittleSurvey)
tally(disease ~ diseasever, data = LittleSurvey)
```

---

MathNoise

*Test performance and noise*

---

### Description

In this experiment, hyperactive and control students were given a mathematics test in either a quiet or loud testing environment.

### Format

A data frame with 40 observations on the following 3 variables.

**score** score on a mathematics test

**noise** a factor with levels hi lo

**group** a factor with levels control hyper

### Source

Sydney S. Zentall and Jandira H. Shaw, Effects of classroom noise on performance and activity of second-grade hyperactive and control children, *Journal of Educational Psychology* 72 (1980), no. 6, 830.

**Examples**

```

data(MathNoise)
xyplot (score ~ noise, data = MathNoise, group = group, type = 'a',
auto.key = list(columns = 2, lines = TRUE, points = FALSE))

gf_jitter(score ~ noise, data = MathNoise, color = ~ group, alpha = 0.4,
width = 0.1, height = 0) %>%
gf_line(score ~ noise, data = MathNoise, color = ~ group, group = ~ group,
stat = "summary")

```

---

maxLik2

*Augmented version of maxLik*


---

**Description**

This version of [max::maxLik()] stores additional information in the returned object enabling a plot method.

**Usage**

```
maxLik2(loglik, ..., env = parent.frame())
```

**Arguments**

loglik	a log-likelihood function as for [maxLik::maxLik()].
...	additional arguments passed to [maxLik::maxLik()].
env	an environment in which to evaluate loglik.

---

MIAA05

*MIAA basketball 2004-2005 season*


---

**Description**

Individual player statistics for the 2004-2005 Michigan Intercollegiate Athletic Association basketball season.

**Format**

A data frame with 134 observations on the following 27 variables.

**number** jersey number

**player** player's name

**GP** games played

**GS** games started

**Min** minutes played  
**AvgMin** average minutes played per game  
**FG** field goals made  
**FGA** field goals attempted  
**FGPct** field goal percentage  
**FG3** 3-point field goals made  
**FG3A** 3-point field goals attempted  
**FG3Pct** 3-point field goal percentage  
**FT** free throws made  
**FTA** free throws attempted  
**FTPct** free throw percentage  
**Off** offensive rebounds  
**Def** defensive rebounds  
**Tot** total rebounds  
**RBG** rebounds per game  
**PF** personal fouls  
**FO** games fouled out  
**A** assists  
**TO** turn overs  
**Blk** blocked shots  
**Stl** steals  
**Pts** points scored  
**PTSG** points per game

### Source

MIAA sports archives (<https://miaa.org/>)

### Examples

```
data(MIAA05)  
gf_histogram(~ FTPct, data = MIAA05)
```

---

MLB2004

*Major League Baseball 2004 team data*

---

### Description

Team batting statistics, runs allowed, and runs scored for the 2004 Major League Baseball season.

### Format

A data frame with 30 observations on the following 20 variables.

**team** team city, a factor

**league** League, a factor with levels AL NL

**W** number of wins

**L** number of losses

**G** number of games

**R** number of runs scored

**OR** opponents' runs – number of runs allowed

**Rdiff** run difference – R – OR

**AB** number of at bats

**H** number of hits

**DBL** number of doubles

**TPL** number of triples

**HR** number of home runs

**BB** number of walks (bases on balls)

**SO** number of strike outs

**SB** number of stolen bases

**CS** number of times caught stealing

**BA** batting average

**SLG** slugging percentage

**OBA** on base average

### Examples

```
data(MLB2004)
gf_point(W ~ Rdiff, data = MLB2004)
```

---

 NCAAbb

*NCAA Division I Basketball Results*


---

**Description**

Results of NCAA basketball games

**Format**

Nine variables describing NCAA Division I basketball games.

**date** date on which game was played

**away** visiting team

**ascore** visiting team's score

**home** home team

**hscore** home team's score

**notes** code indicting games played at neutral sites (n or N) or in tournaments (T)

**location** where game was played

**season** a character indicating which season the game belonged to

**postseason** a logical indicating whether the game is a postseason game

**Source**

<https://kenpom.com>

**Examples**

```
data(NCAAbb)
# select one year and add some additional variables to the data frame
NCAA2010 <-
  NCAAbb %>%
  filter(season == "2009-10") %>%
  mutate(
    dscore = hscore - ascore,
    homeTeamWon = dscore > 0,
    numHomeTeamWon <- -1 + 2 * as.numeric(homeTeamWon),
    winner = ifelse(homeTeamWon, home, away),
    loser = ifelse(homeTeamWon, away, home),
    wscore = ifelse(homeTeamWon, hscore, ascore),
    lscore = ifelse(homeTeamWon, ascore, hscore)
  )
NCAA2010 %>% select(date, winner, loser, wscore, lscore, dscore, homeTeamWon) %>% head()
```

---

`NFL2007`*NFL 2007 season*

---

**Description**

Results of National Football League games (2007 season, including playoffs)

**Format**

A data frame with 267 observations on the following 7 variables.

**date** date on which game was played

**visitor** visiting team

**visitorScore** score for visiting team

**home** home team

**homeScore** score for home team

**line** 'betting line'

**totalLine** 'over/under' line (for combined score of both teams)

**Examples**

```
data(NFL2007)
NFL <- NFL2007
NFL$dscore <- NFL$homeScore - NFL$visitorScore
w <- which(NFL$dscore > 0)
NFL$winner <- NFL$visitor; NFL$winner[w] <- NFL$home[w]
NFL$loser <- NFL$home; NFL$loser[w] <- NFL$visitor[w]
# did the home team win?
NFL$homeTeamWon <- NFL$dscore > 0
table(NFL$homeTeamWon)
table(NFL$dscore > NFL$line)
```

---

`n1max`*Nonlinear maximization and minimization*

---

**Description**

`n1min` and `n1max` are thin wrappers around `nlm`, a non-linear minimizer. `n1max` avoids the necessity of modifying the function to construct a minimization problem from a problem that is naturally a maximization problem. The summary method for the resulting objects provides output that is easier for humans to read.

**Usage**

```
nlmax(f, ...)

nlmin(f, ...)

## S3 method for class 'nlmax'
summary(object, nsmall = 4, ...)

## S3 method for class 'nlmin'
summary(object, nsmall = 4, ...)
```

**Arguments**

<code>f</code>	a function to optimize
<code>...</code>	additional arguments passed to <code>nlm</code> . Note that <code>p</code> is a required argument for <code>nlm</code> . See the help for <code>nlm</code> for details.
<code>object</code>	an object returned from <code>nlmin</code> or <code>nlmax</code>
<code>nsmall</code>	a numeric passed through to <code>format</code>

**Examples**

```
summary( nlmax( function(x) 5 - 3*x - 5*x^2, p=0 ) )
```

---

Noise

*Noise*


---

**Description**

In order to test the effect of room noise, subjects were given a test under 5 different sets of conditions: 1) no noise, 2) intermittent low volume, 3) intermittent high volume, 4) continuous low volume, and 5) continuous high volume.

**Format**

A data frame with 50 observations on the following 5 variables.

**id** subject identifier

**score** score on the test

**condition** numeric code for condition

**volume** a factor with levels high low none

**frequency** a factor with levels continuous intermittent none

**Examples**

```

data(Noise)
Noise2 <- Noise %>% filter(volume != 'none')
model <- lm(score ~ volume * frequency, data = Noise2)
anova(model)
gf_jitter(score ~ volume, data = Noise2, color = ~ frequency,
          alpha = 0.4, width = 0.1, height = 0) %>%
  gf_line(score ~ volume, data = Noise2, group = ~frequency, color = ~ frequency,
          stat = "summary")

gf_jitter(score ~ frequency, data = Noise2, color = ~ volume,
          alpha = 0.4, width = 0.1, height = 0) %>%
  gf_line(score ~ frequency, data = Noise2, group = ~ volume, color = ~ volume,
          stat = "summary")

```

---

Pallets

*Pallet repair data*


---

**Description**

The pallets data set contains data from a firm that recycles pallets. Pallets from warehouses are bought, repaired, and resold. (Repairing a pallet typically involves replacing one or two boards.) The company has four employees who do the repairs. The employer sampled five days for each employee and recorded the number of pallets repaired.

**Format**

A data frame with 20 observations on the following 3 variables.

**pallets** number of pallets repaired

**employee** a factor with levels A B C D

**day** a factor with levels day1 day2 day3 day4 day5

**Source**

Michael Stob, Calvin College

**Examples**

```

data(Pallets)
# Do the employees differ in the rate at which they repair pallets?
pal.lm1 <- lm(pallets ~ employee, data = Pallets)
anova(pal.lm1)
# Now using day as a blocking variable
pal.lm2 <- lm(pallets ~ employee + day, data = Pallets)
anova(pal.lm2)
gf_line(pallets ~ day, data = Pallets,

```

```
group = ~employee,
color = ~employee) %>%
  gf_point() %>%
  gf_labs(title = "Productivity by day and employee")
```

---

 PaperPlanes

*Paper airplanes*


---

## Description

Student-collected data from an experiment investigating the design of paper airplanes.

## Format

A data frame with 16 observations on the following 5 variables.

**distance** distance plane traveled (cm)

**paper** type of paper used

**angle** a numeric vector

**design** design of plane (hi performance or simple)

**order** order in which planes were thrown

## Details

These data were collected by Stewart Fischer and David Tippetts, statistics students at the Queensland University of Technology in a subject taught by Dr. Margaret Mackisack. Here is their description of the data and its collection:

The experiment decided upon was to see if by using two different designs of paper aeroplane, how far the plane would travel. In considering this, the question arose, whether different types of paper and different angles of release would have any effect on the distance travelled. Knowing that paper aeroplanes are greatly influenced by wind, we had to find a way to eliminate this factor. We decided to perform the experiment in a hallway of the University, where the effects of wind can be controlled to some extent by closing doors.

In order to make the experimental units as homogeneous as possible we allocated one person to a task, so person 1 folded and threw all planes, person 2 calculated the random order assignment, measured all the distances, checked that the angles of flight were right, and checked that the plane release was the same each time.

The factors that we considered each had two levels as follows:

Paper: A4 size, 80g and 50g

Design: High Performance Dual Glider, and Incredibly Simple Glider (patterns attached to original report)

Angle of release: Horizontal, or 45 degrees upward.

The random order assignment was calculated using the random number function of a calculator. Each combination of factors was assigned a number from one to eight, the random numbers were generated and accordingly the order of the experiment was found.

**Source**

These data are also available at OzDASL, the Australasian Data and Story Library (<https://dasl.datadescription.com/>).

**References**

Mackisack, M. S. (1994). What is the use of experiments conducted by statistics students? *Journal of Statistics Education*, 2, no 1.

**Examples**

```
data(PaperPlanes)
```

---

Pendulum

*Pendulum data*

---

**Description**

Period and pendulum length for a number of string and mass pendulums constructed by physics students. The same mass was used throughout, but the length of the string was varied from 10cm to 16 m.

**Format**

A data frame with 27 observations on the following 3 variables.

**length** length of the pendulum (in meters)

**period** average time of period (in seconds) over several swings of the pendulum

**delta.length** an estimate of the accuracy of the length measurement

**Source**

Calvin College physics students under the direction of Professor Steve Plath.

**Examples**

```
data(Pendulum)
gf_point(period ~ length, data = Pendulum)
```

---

PetStress

*Pets and stress*

---

### Description

Does having a pet or a friend cause more stress?

### Format

A data frame with 45 observations on the following 2 variables.

**group** a factor with levels Control, Friend, or Pet

**rate** average heart rate while performing a stressful task

### Details

Forty-five women, all self-proclaimed dog-lovers, were randomly divided into three groups of subjects. Each performed a stressful task either alone, with a friend present, or with their dog present. The average heart rate during the task was used as a measure of stress.

### Source

K. M. Allen, J. Blascovich, J. Tomaka, and R. M. Kelsey, Presence of human friends and pet dogs as moderators of autonomic responses to stress in women, *Journal of Personality and Social Psychology* 61 (1991), no. 4, 582–589.

### References

These data also appear in

Brigitte Baldi and David S. Moore, *The Practice of Statistics in the Life Sciences*, Freeman, 2009.

### Examples

```
data(PetStress)
xyplot(rate ~ group, data = PetStress, jitter.x = TRUE, type = c('p', 'a'))
gf_jitter(rate ~ group, data = PetStress, width = 0.1, height = 0) %>%
  gf_line(group = 1, stat = "summary", color = "red")
```

---

Pheno

*FUSION type 2 diabetes study*

---

### Description

Phenotype and genotype data from the Finland United States Investigation of NIDDM (type 2) Diabetes (FUSION) study.

### Format

Data frames with the following variables.

**id** subject ID number for matching between data sets

**t2d** a factor with levels case control

**bmi** body mass index

**sex** a factor with levels F M

**age** age of subject at time phenotypes were collected

**smoker** a factor with levels former never occasional regular

**chol** total cholesterol

**waist** waist circumference (cm)

**weight** weight (kg)

**height** height (cm)

**whr** waist hip ratio

**sbp** systolic blood pressure

**dbp** diastolic blood pressure

**marker** RS name of SNP

**markerID** numeric ID for SNP

**allele1** first allele coded as 1 = A, 2 = C, 3 = G, 4 = T

**allele2** second allele coded as 1 = A, 2 = C, 3 = G, 4 = T

**genotype** both alleles coded as a factor

**Adose** number of A alleles

**Cdose** number of C alleles

**Gdose** number of G alleles

**Tdose** number of T alleles

**Source**

Similar to the data presented in

Laura J. Scott, Karen L. Mohlke, Lori L. Bonnycastle, Cristen J. Willer, Yun Li, William L. Duren, Michael R. Erdos, Heather M. Stringham, Peter S. Chines, Anne U. Jackson, Ludmila Prokunina-Olsson, Chia-Jen J. Ding, Amy J. Swift, Narisu Narisu, Tianle Hu, Randall Pruim, Rui Xiao, Xiao-Yi Y. Li, Karen N. Conneely, Nancy L. Riebow, Andrew G. Sprau, Maurine Tong, Peggy P. White, Kurt N. Hetrick, Michael W. Barnhart, Craig W. Bark, Janet L. Goldstein, Lee Watkins, Fang Xiang, Jouko Saramies, Thomas A. Buchanan, Richard M. Watanabe, Timo T. Valle, Leena Kinnunen, Goncalo R. Abecasis, Elizabeth W. Pugh, Kimberly F. Doheny, Richard N. Bergman, Jaakko Tuomilehto, Francis S. Collins, and Michael Boehnke, A genome-wide association study of type 2 diabetes in Finns detects multiple susceptibility variants, *Science* (2007).

**Examples**

```
data(Pheno); data(FUSION1); data(FUSION2)
FUSION1m <- merge(FUSION1, Pheno, by = "id", all.x = FALSE, all.y = FALSE)
xtabs( ~ t2d + genotype, data = FUSION1m)
xtabs( ~ t2d + Gdose, data = FUSION1m)
chisq.test( xtabs( ~ t2d + genotype, data = FUSION1m ) )
f1.glm <- glm( factor(t2d) ~ Gdose, data = FUSION1m, family = binomial)
summary(f1.glm)
```

---

Pigs

*Pass the Pigs*

---

**Description**

This data set contains information collected from rolling the pair of pigs (found in the game "Pass the Pigs") 6000 times.

**Format**

A data frame with 6000 observations on the following 6 variables.

**roll** roll number (1-6000)

**blackScore** numerical code for position of black pig

**black** position of black pig coded as a factor

**pinkScore** numerical code for position of pink pig

**pink** position of pink pig coded as a factor

**score** score of the roll

**height** height from which pigs were rolled (5 or 8 inches)

**start** starting position of the pigs (0 = both pigs backwards, 1 = one backwards one forwards, 2 = both forwards)

## Details

In "Pass the Pigs", players roll two pig-shaped rubber dice and earn or lose points depending on the configuration of the rolled pigs. Players compete individually to earn 100 points. On each turn, a player rolls he or she decides to stop or until "pigging out" or

The pig configurations and their associated scores are

1 = Dot Up (0)

2 = Dot Down (0)

3 = Trotter (5)

4 = Razorback (5)

5 = Snouter (10)

6 = Leaning Jowler (15)

7 = Pigs are touching one another (-1; lose all points)

One pig Dot Up and one Dot Down ends the turn (a "pig out") and results in 0 points for the turn. If the pigs touch, the turn is ended and all points for the game must be forfeited. Two pigs in the Dot Up or Dot Down configuration score 1 point. Otherwise, The scores of the two pigs in different configurations are added together. The score is doubled if both both pigs have the same configuration, so, for example, two Snouters are worth 40 rather than 20.

## Source

John C. Kern II, Duquesne University (<kern@mathcs.duq.edu>)

## Examples

```
data(Pigs)
tally( ~ black, data = Pigs )
if (require(tidyr)) {
  Pigs %>%
  select(roll, black, pink) %>%
  gather(pig, state, black, pink) %>%
  tally( state ~ pig, data = ., format = "prop", margins = TRUE)
}
```

## Description

Major League Baseball pitching statistics for the 2005 season.

**Format**

A data frame with 653 observations on the following 26 variables.

**playerID** unique identifier for each player

**yearID** year

**stint** for players who played with multiple teams in the same season, **stint** is increased by one each time the player joins a new team

**teamID** three-letter identifier for team

**lgID** league team plays in, coded as AL or NL

**W** wins

**L** losses

**G** games played in

**GS** games started

**CG** complete games

**SHO** shut outs

**SV** saves recorded

**IPouts** outs recorded (innings pitched, measured in outs rather than innings)

**H** hits allowed

**ER** earned runs allowed

**HR** home runs allowed

**BB** walks (bases on balls) allowed

**SO** strike outs

**ERA** earned run average

**IBB** intentional walks

**WP** wild pitches

**HBP** number of batters hit by pitch

**BK** balks

**BFP** batters faced pitching

**GF** ratio of ground balls to fly balls

**R** runs allowed

**Examples**

```
data(Pitching2005)
gf_point(IPouts/3 ~ W, data = Pitching2005, ylab = "innings pitched", xlab = "wins")
```

---

plot.maxLik2	<i>plot method for augment maxLik objects</i>
--------------	---

---

### Description

See [maxLik2()] and [maxLik::maxLik()] for how to create the objects this method prints.

### Usage

```
## S3 method for class 'maxLik2'
plot(x, y, ci = "Wald", hline = FALSE, ...)
```

### Arguments

x	an object of class "maxLik2"
y	ignored
ci	a character vector with values among "Wald" and "likelihood" specifying the type of interval to display
hline	a logical indicating whether a horizontal line should be added
...	additional arguments, currently ignored.

---

Poison	<i>Poison data</i>
--------	--------------------

---

### Description

The data give the survival times (in hours) in a 3 x 4 factorial experiment, the factors being (a) three poisons and (b) four treatments. Each combination of the two factors is used for four animals. The allocation to animals is completely randomized.

### Format

A data frame with 48 observations on the following 3 variables.

**poison** type of poison (1, 2, or 3)  
**treatment** manner of treatment (1, 2, 3, or 4)  
**time** time until death (hours)

### Source

These data are also available from OzDASL, the Australian Data and Story Library (<https://dasl.datadescription.com/>). (Note: The time measurements of the data at OzDASL are in units of tens of hours.)

## References

- Box, G. E. P., and Cox, D. R. (1964). An analysis of transformations (with Discussion). *J. R. Statist. Soc. B*, 26, 211-252.
- Aitkin, M. (1987). Modelling variance heterogeneity in normal regression using GLIM. *Appl. Statist.*, 36, 332-339.
- Smyth, G. K., and Verbyla, A. P. (1999). Adjusted likelihood methods for modelling dispersion in generalized linear models. *Environmetrics* 10, 696-709. <http://www.statsci.org/smyth/pubs/ties98tr.html>.

## Examples

```
data(poison)
poison.lm <- lm(time~factor(poison) * factor(treatment), data = Poison)
plot(poison.lm,w = c(4,2))
anova(poison.lm)
# improved fit using a transformation
poison.lm2 <- lm(1/time ~ factor(poison) * factor(treatment), data = Poison)
plot(poison.lm2,w = c(4,2))
anova(poison.lm)
```

---

Punting

*American football punting*

---

## Description

Investigators studied physical characteristics and ability in 13 football punters. Each volunteer punted a football ten times. The investigators recorded the average distance for the ten punts, in feet. They also recorded the average hang time (time the ball is in the air before the receiver catches it), and a number of measures of leg strength and flexibility.

## Format

A data frame with 13 observations on the following 7 variables.

**distance** mean distance for 10 punts (feet)

**hang** mean hang time (seconds)

**rStrength** right leg strength (pounds)

**lStrength** left leg strength (pounds)

**rFlexibility** right leg flexibility (degrees)

**lFlexibility** left leg flexibility (degrees)

**oStrength** overall leg strength (foot-pounds)

## Source

These data are also available at OzDASL (<https://dasl.datadescription.com/>).

## References

"The relationship between selected physical performance variables and football punting ability" by the Department of Health, Physical Education and Recreation at the Virginia Polytechnic Institute and State University, 1983.

## Examples

```
data(Punting)
gf_point(hang ~ distance, data = Punting)
```

---

RatPoison	<i>Rat poison – unfinished documentation</i>
-----------	--

---

## Description

Data from an experiment to see whether flavor and location of rat poison influence the consumption by rats.

## Format

A data frame with 20 observations on the following 3 variables.

**consumption** a numeric vector

**flavor** a factor with levels bread butter-vanilla plain roast beef

**location** a factor with levels A B C D E

## Examples

```
data(RatPoison)
gf_line(consumption ~ flavor, group = ~ location, color = ~ location, data = RatPoison) %>%
  gf_point()
```

---

rgolfballs	<i>Simulated golf ball data</i>
------------	---------------------------------

---

## Description

A matrix of random golf ball numbers simulated using `rmultinom(n = 10000, size = 486, prob = rep(0.25, 4))`.

## Examples

```
data(rgolfballs)
```

---

 RubberBand

*Rubber band launching – unfinished documentation*


---

**Description**

Results of an experiment comparing a rubber band travels to the amount it was stretched prior to launch.

**Format**

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

**stretch** amount rubber band was stretched before launch

**distance** distance rubber band traveled

**Examples**

```
data(RubberBand)
gf_point(distance ~ stretch, data = RubberBand) %>%
  gf_lm(interval = "confidence")
```

---

 Scent

*Maze tracing and scents*


---

**Description**

Subjects were asked to complete a pencil and paper maze when they were smelling a floral scent and when they were not.

**Format**

A data frame with 21 observations on the following 12 variables.

**id** ID number

**sex** a factor with levels F and M

**smoker** a factor with levels N, Y

**opinion** opinion of the odor (indiff, neg, or pos)

**age** age of subject (in years)

**first** which treatment was first, scented or unscented

**u1** time (in seconds) in first unscented trial

**u2** time (in seconds) in second unscented trial

**u3** time (in seconds) in third unscented trial

**s1** time (in seconds) in first scented trial

**s2** time (in seconds) in second scented trial

**s3** time (in seconds) in third scented trial

**Source**

These data are also available at DASL, the data and story library (<https://dasl.datadescription.com/>).

**References**

Hirsch, A. R., and Johnston, L. H. "Odors and Learning," Smell & Taste Treatment and Research Foundation, Chicago.

**Examples**

```
data(Scent)
summary(Scent)
```

---

snippet

*Display or execute a snippet of R code*

---

**Description**

This command will display and/or execute small snippets of R code from the book *Foundations and Applications of Statistics: An Introduction Using R*.

**Usage**

```
snippet(
  name,
  eval = TRUE,
  execute = eval,
  view = !execute,
  echo = TRUE,
  ask = getOption("demo.ask"),
  verbose = getOption("verbose"),
  lib.loc = NULL,
  character.only = FALSE,
  regex = NULL,
  max.files = 10L
)
```

**Arguments**

name	name of snippet
eval	a logical. An alias for 'execute'.
execute	a logical. If TRUE, snippet code is executed. (The code and the results of the execution will be visible if echo is TRUE.)
view	a logical. If TRUE, snippet code is displayed 'as is'.

echo	a logical. If TRUE, show the R input when executing.
ask	a logical (or "default") indicating if <code>devAskNewPage(ask=TRUE)</code> should be called before graphical output happens from the snippet code. The value "default" (the factory-fresh default) means to ask if <code>echo == TRUE</code> and the graphics device appears to be interactive. This parameter applies both to any currently opened device and to any devices opened by the demo code. If this is evaluated to TRUE and the session is interactive, the user is asked to press RETURN to start.
verbose	a logical. If TRUE, additional diagnostics are printed.
lib.loc	character vector of directory names of R libraries, or NULL. The default value of NULL corresponds to all libraries currently known.
character.only	logical. If TRUE, use <code>nameas</code> character string.
regex	ignored. Retained for backwards compatibility.
max.files	an integer limiting the number of files retrieved.

### Details

snippet works much like [demo](#), but the interface is simplified. Partial matching is used to select snippets, so any unique prefix is sufficient to specify a snippet. Sequenced snippets (identified by trailing 2-digit numbers) will be executed in sequence if a unique prefix to the non-numeric portion is given. To run just one of a sequence of snippets, provide the full snippet name. See the examples.

### Author(s)

Randall Pruim

### See Also

[demo](#), [source](#).

### Examples

```
snippet("normal01")
# prefix works
snippet("normal")
# this prefix is ambiguous
snippet("norm")
# sequence of "histogram" snippets
snippet("hist", eval = FALSE, echo = TRUE, view = FALSE)
# just one of the "histogram" snippets
snippet("histogram04", eval = FALSE, echo = TRUE, view = FALSE)
# Prefix too short, but a helpful message is displayed
snippet("h", eval = FALSE, echo = TRUE, view = FALSE)
```

---

Soap

*Dwindling soap*

---

## Description

A bar of soap was weighed after showering to see how much soap was used each shower.

## Format

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

### **date**

**day** days since start of soap usage and data collection

**weight** weight of bar of soap (in grams)

## Details

According to Rex Boggs:

I had a hypothesis that the daily weight of my bar of soap [in grams] in my shower wasn't a linear function, the reason being that the tiny little bar of soap at the end of its life seemed to hang around for just about ever. I wanted to throw it out, but I felt I shouldn't do so until it became unusable. And that seemed to take weeks.

Also I had recently bought some digital kitchen scales and felt I needed to use them to justify the cost. I hypothesized that the daily weight of a bar of soap might be dependent upon surface area, and hence would be a quadratic function . . . .

The data ends at day 22. On day 23 the soap broke into two pieces and one piece went down the plughole.

## Source

Data collected by Rex Boggs and available from OzDASL (<https://das1.datadescription.com/>).

## Examples

```
data(Soap)
gf_point(weight ~ day, data = Soap)
```

Spheres

*Measuring spheres*

---

**Description**

Measurements of the diameter (in meters) and mass (in kilograms) of a set of steel ball bearings.

**Format**

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

**diameter** diameter of bearing (m)

**mass** mass of the bearing (kg)

**Source**

These data were collected by Calvin College physics students under the direction of Steve Plath.

**Examples**

```
data(Spheres)
gf_point(mass ~ diameter, data = Spheres)
gf_point(mass ~ diameter, data = Spheres) %>%
  gf_refine(scale_x_log10(), scale_y_log10())
```

---

SSplot

*Sum of Squares Plots*

---

**Description**

This function creates plots showing the "consumption" of residual sum of squares resulting from adding predictors to a model.

**Usage**

```
SSplot(
  model1,
  model2,
  n = 1,
  col1 = "gray50",
  size1 = 0.6,
  col2 = "navy",
  size2 = 1,
  col3 = "red",
  size3 = 1,
  ...,
  env = parent.frame()
)
```

**Arguments**

model1	a linear model
model2	a linear model, often using [mosaic::rand()].
n	an integer specifying how many times to regenerate model2.
col1, col2, col3	Colors for the line segments in the plot
size1, size2, size3	Sizes of the line segments in the plot
...	additional arguments (currently ignored)
env	an environment in which to evaluate the models.

**Examples**

```
SSplot(
  lm(strength ~ limestone + water, data = Concrete),
  lm(strength ~ limestone + rand(7), data = Concrete),
  n = 50)
## Not run:
SSplot(
  lm(strength ~ water + limestone, data = Concrete),
  lm(strength ~ water + rand(7), data = Concrete),
  n = 1000)

## End(Not run)
```

---

Step

*Stepping experiment*


---

**Description**

An experiment was conducted by students at The Ohio State University in the fall of 1993 to explore the nature of the relationship between a person's heart rate and the frequency at which that person stepped up and down on steps of various heights.

**Format**

A data frame with 30 observations on the following 7 variables.

**order** performance order

**block** number of experimenter block

**restHR** resting heart rate (beats per minute)

**HR** final heart rate

**height** height of step (hi or lo)

**freq** whether subject stepped fast, medium, or slow

### Details

An experiment was conducted by students at The Ohio State University in the fall of 1993 to explore the nature of the relationship between a person's heart rate and the frequency at which that person stepped up and down on steps of various heights. The response variable, heart rate, was measured in beats per minute. There were two different step heights: 5.75 inches (coded as lo), and 11.5 inches (coded as hi). There were three rates of stepping: 14 steps/min. (coded as slow), 21 steps/min. (coded as medium), and 28 steps/min. (coded as fast). This resulted in six possible height/frequency combinations. Each subject performed the activity for three minutes. Subjects were kept on pace by the beat of an electric metronome. One experimenter counted the subject's pulse for 20 seconds before and after each trial. The subject always rested between trials until her or his heart rate returned to close to the beginning rate. Another experimenter kept track of the time spent stepping. Each subject was always measured and timed by the same pair of experimenters to reduce variability in the experiment. Each pair of experimenters was treated as a block.

### Source

These data are available at DASL, the data and story library (<https://dasl.datadescription.com/>).

### Examples

```
data(Step)
gf_jitter(HR-restHR ~ freq, color = ~height, data = Step, group = ~height,
          height = 0, width = 0.1) %>%
  gf_line(stat = "summary", group = ~height)
gf_jitter(HR-restHR ~ height, color = ~freq, data = Step, group = ~freq,
          height = 0, width = 0.1) %>%
  gf_line(stat = "summary", group = ~freq)
```

---

Stereogram

*Stereogram fusion*

---

### Description

Results of an experiment on the effect of prior information on the time to fuse random dot stereograms. One group (NV) was given either no information or just verbal information about the shape of the embedded object. A second group (group VV) received both verbal information and visual information (e.g., a drawing of the object).

### Format

A data frame with 78 observations on the following 2 variables.

**time** time until subject was able to fuse a random dot stereogram

**group** treatment group: NV(no visual instructions) VV (visual instructions)

**Source**

These data are available at DASL, the data and story library (<https://dasl.datadescription.com/>).

**References**

Frisby, J. P. and Clatworthy, J. L., "Learning to see complex random-dot stereograms," *Perception*, 4, (1975), pp. 173-178.

Cleveland, W. S. *Visualizing Data*. 1993.

**Examples**

```
data(Stereogram)
favstats(time ~ group, data = Stereogram)
gf_violin(time ~ group, data = Stereogram, alpha = 0.2, fill = "skyblue") %>%
gf_jitter(time ~ group, data = Stereogram, height = 0, width = 0.25)
```

---

Students

*Standardized test scores and GPAs*

---

**Description**

Standardized test scores and GPAs for 1000 students.

**Format**

A data frame with 1000 observations on the following 6 variables.

**ACT** ACT score

**SAT** SAT score

**grad** has the student graduated from college?

**gradGPA** college GPA at graduation

**hsGPA** high school GPA

**cohort** year of graduation or expected graduation

**Examples**

```
data(Students)
gf_point(ACT ~ SAT, data = Students)
gf_point(gradGPA ~ hsGPA, data = Students)
```

---

TasteTest

*Taste test data*

---

## Description

The results from a study comparing different preparation methods for taste test samples.

## Format

A data frame with 16 observations on 2 (taste1) or 4 (tastetest) variables.

**score** taste score from a group of 50 testers

**scr** a factor with levels coarse fine

**liq** a factor with levels hi lo

**type** a factor with levels A B C D

## Details

The samples were prepared for tasting using either a coarse screen or a fine screen, and with either a high or low liquid content. A total taste score is recorded for each of 16 groups of 50 testers each. Each group had 25 men and 25 women, each of whom scored the samples on a scale from -3 (terrible) to 3 (excellent). The sum of these individual scores is the overall taste score for the group.

## Source

E. Street and M. G. Carroll, *Preliminary evaluation of a food product*, *Statistics: A Guide to the Unknown* (Judith M. Tanur et al., eds.), Holden-Day, 1972, pp. 220-238.

## Examples

```
data(TasteTest)
data(Taste1)
gf_jitter(score ~ scr, data = TasteTest, color = ~liq, width = 0.2, height = 0) %>%
  gf_line(stat = "summary", group = ~liq)
df_stats(score ~ scr | liq, data = TasteTest)
```

---

tdf	<i>Compute degrees of freedom for a 2-sample t-test</i>
-----	---

---

### Description

This function computes degrees of freedom for a 2-sample t-test from the standard deviations and sample sizes of the two samples.

### Usage

```
tdf(sd1, sd2, n1, n2)
```

### Arguments

sd1	standard deviation of the sample 1
sd2	standard deviation of the sample 2
n1	size of sample 1
n2	size of sample 2

### Value

estimated degrees of freedom for 2-sample t-test

### Examples

```
data(KidsFeet, package="mosaicData")
fs <- favstats( length ~ sex, data=KidsFeet ); fs
t.test( length ~ sex, data=KidsFeet )
tdf( fs[1,'sd'], fs[2,'sd'], fs[1,'n'], fs[2,'n'] )
```

---

TireWear	<i>Estimating tire wear</i>
----------	-----------------------------

---

### Description

Tread wear is estimated by two methods: weight loss and groove wear.

### Format

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

**weight** estimated wear (1000's of miles) base on weight loss

**groove** estimated wear (1000's of miles) based on groove wear

**Source**

These data are available at DASL, the Data and Story Library (<https://dasl.datadescription.com/>).

**References**

R. D. Stichler, G. G. Richey, and J. Mandel, "Measurement of Treadware of Commercial Tires", *Rubber Age*, 73:2 (May 1953).

**Examples**

```
data(TireWear)
gf_point(weight ~ groove, data = TireWear)
```

---

Traffic

*New England traffic fatalities (1951-1959)*

---

**Description**

Used by Tufte as an example of the importance of context, these data show the traffic fatality rates in New England in the 1950s. Connecticut increased enforcement of speed limits in 1956. In their full context, it is difficult to say if the decline in Connecticut traffic fatalities from 1955 to 1956 can be attributed to the stricter enforcement.

**Format**

A data frame with 9 observations on the following 6 variables.

**year** a year from 1951 to 1959  
**cn.deaths** number of traffic deaths in Connecticut  
**ny** deaths per 100,000 in New York  
**cn** deaths per 100,000 in Connecticut  
**ma** deaths per 100,000 in Massachusetts  
**ri** deaths per 100,000 in in Rhode Island

**Source**

Tufte, E. R. *The Visual Display of Quantitative Information*, 2nd ed. Graphics Press, 2001.

**References**

Donald T. Campbell and H. Laurence Ross. "The Connecticut Crackdown on Speeding: Time-Series Data in Quasi-Experimental Analysis", *Law & Society Review* Vol. 3, No. 1 (Aug., 1968), pp. 33-54.

Gene V. Glass. "Analysis of Data on the Connecticut Speeding Crackdown as a Time-Series Quasi-Experiment" *Law & Society Review*, Vol. 3, No. 1 (Aug., 1968), pp. 55-76.

**Examples**

```

data(Traffic)
gf_line(cn.deaths ~ year, data = Traffic)
if (require(tidyr)) {
  TrafficLong <-
    Traffic %>%
    select(-2) %>%
    gather(state, fatality.rate, ny:ri)
  gf_line(fatality.rate ~ year, group = ~state, color = ~state, data = TrafficLong) %>%
  gf_point(fatality.rate ~ year, group = ~state, color = ~state, data = TrafficLong) %>%
  gf_lims(y = c(0, NA))
}

```

Trebuchet

*Trebuchet data***Description**

Measurements from an experiment that involved firing projectiles with a small trebuchet under different conditions.

**Format**

Data frames with the following variables.

**object** the object serving as projectilebean big washerb bigWash BWB foose golf MWB SWB tennis ball wood

**projectileWt** weight of projectile (in grams)

**counterWt** weight of counter weight (in kg)

**distance** distance projectile traveled (in cm)

**form** a factor with levels a b B c describing the configuration of the trebuchet.

**Details**

Trebuchet1 and Trebuchet2 are subsets of Trebuchet restricted to a single value of counterWt

**Source**

Data collected by Andrew Pruiam as part of a Science Olympiad competition.

**Examples**

```

data(Trebuchet); data(Trebuchet1); data(Trebuchet2)
gf_point(distance ~ projectileWt, data = Trebuchet1)
gf_point(distance ~ projectileWt, data = Trebuchet2)
gf_point(distance ~ projectileWt, color = ~ factor(counterWt), data = Trebuchet) %>%
  gf_smooth(alpha = 0.2, fill = ~factor(counterWt))

```

---

undocumented	<i>Undocumented functions</i>
--------------	-------------------------------

---

**Description**

These objects are undocumented.

**Details**

Some are left-overs from a previous version of the book and package. In other cases, the functions are of limited suitability for general use.

**Author(s)**

Randall Pruim

---

Unemployment	<i>Unemployment data</i>
--------------	--------------------------

---

**Description**

Unemployment data

**Usage**

```
data(Unemployment)
```

**Format**

A data.frame with 10 observations on the following 4 variables.

unemp Millions of unemployed people

production Federal Reserve Board index of industrial production

year

iyear indexed year

**Source**

Paul F. Velleman and Roy E. Welsch. "Efficient Computing of Regression Diagnostics", *The American Statistician*, Vol. 35, No. 4 (Nov., 1981), pp. 234-242. (<https://www.jstor.org/stable/2683296>)

**Examples**

```
data(Unemployment)
```

---

vaov	<i>ANOVA vectors</i>
------	----------------------

---

### Description

Compute vectors associated with 1-way ANOVA

### Usage

```
vaov(x, ...)  
  
## S3 method for class 'formula'  
vaov(x, data = parent.frame(), ...)
```

### Arguments

x	a formula.
...	additional arguments.
data	a data frame.

### Details

This is primarily designed for demonstration purposes to show how 1-way ANOVA models partition variance. It may not work properly for more complicated models.

### Value

A data frame with variables including grandMean, groupMean, ObsVsGrand, STotal, ObsVsGroup, SError, GroupVsGrand, and STreatment. The usual SS terms can be computed from these by summing.

### Examples

```
aov(pollution ~ location, data = AirPollution)  
vaov(pollution ~ location, data = AirPollution)
```

---

`wilson.ci`*Confidence Intervals for Proportions*

---

**Description**

Alternatives to `prop.test` and `binom.test`.

**Usage**

```
wilson.ci(x, n = 100, conf.level = 0.95)
```

**Arguments**

<code>x</code>	number of 'successes'
<code>n</code>	number of trials
<code>conf.level</code>	confidence level

**Details**

`wald.ci` produces Wald confidence intervals. `wilson.ci` produces Wilson confidence intervals (also called “plus-4” confidence intervals) which are Wald intervals computed from data formed by adding 2 successes and 2 failures. The Wilson confidence intervals have better coverage rates for small samples.

**Value**

Lower and upper bounds of a two-sided confidence interval.

**Author(s)**

Randall Pruim

**References**

A. Agresti and B. A. Coull, Approximate is better than ‘exact’ for interval estimation of binomial proportions, *American Statistician* 52 (1998), 119–126.

**Examples**

```
prop.test(12, 30)
prop.test(12, 30, correct=FALSE)
wald.ci(12, 30)
wilson.ci(12, 30)
wald.ci(12+2, 30+4)
```

---

WorkingWomen

*Women in the workforce*

---

### Description

The labor force participation rate of women in each of 19 U.S. cities in each of two years. #  
Reference: United States Department of Labor Statistics ## Authorization: free use ## Description:  
## Variable Names: ## 1. City: City in the United States # 2. labor72: Labor Force Participation  
rate of women in 1972 # 3. labor68: Labor Force Participation rate of women in 1968 ## The Data:  
#

### Format

A data frame with 19 observations on the following 3 variables.

**city** name of a U.S. city (coded as a factor with 19 levels)

**labor72** percent of women in labor force in 1972

**labor68** percent of women in labor force in 1968

### Source

These data are from the United States Department of Labor Statistics and are also available at  
DASL, the Data and Story Library (<https://dasl.datadescription.com/>).

### Examples

```
data(WorkingWomen)
gf_point(labor72 ~ labor68, data = WorkingWomen)
```

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