



Basics

```
print("Hello world!")
# assign value to object
myNumber <- 5
# combine values into vector
myVector <- c(1, 2, 3)
# first element in vector
myVector[1]
# orders values or cases
sort (myVector)
# arithmetic operations
sum(1, 2, 100), +, -, *, /
sqrt (157)
abs(data set$Y)
# logical operations
>, <, >=, <=, ==, !=, |, &
# results in a variable with values
# of TRUE or FALSE
data set$C <- data set$A > data set$B
```

Probability Distribution

```
# calculate the probability area
xpnorm(65.1, data set$mean, data set$sd)
zscore(data set$Y)
# returns t at this probability
at(.975, df = 999)
# returns F at this probability
qf(.95, df1 = 1, df2 = 100)
# CI using t distribution
confint(empty model)
# calculate p-value using F-distribution
xpf(sample F, df1 = 2, df2 = 10)
```

Summary Tables

```
# compute five-number summary
favstats(~ Y, data = data set)
# create frequency table
tally(data set$Y)
tally(~ Y, data = data set)
# tally by condition
tally (\sim Y < 1900, data = data set)
# two-way frequency table
tally(Y ~ X, data = data set, margin = TRUE,
format = "proportion")
```

Simple Statistics

```
mean(data set$Y)
var(data set$Y)
sd(data set$Y)
cohensD(Y ~ X, data = data set)
cor(Y ~ X, data = data set)
b1(Y ~ X, data = data set)
b1 (one model)
pre(Y ~ X, data = data set)
f(Y ~ X, data = data set)
```

Data Frame

```
# structure of data frame
str(data set)
# view first/last six rows
head(data set)
tail(data set)
# select multiple variables
select(data set, Y1, Y2)
# first six rows of selected variables
head(select(data set, Y1, Y2))
# select variable (a column)
data set$Y
# find rows that meet condition
data set[data set$Y > 40]
filter(data set, Y > 300)
filter(data set, Y != "NA")
```

```
# arrange rows by variable
arrange(data set, Y)
# creates data frame from csv file
data set <- read.csv("file name", header = TRUE)</pre>
# convert quantitative variable
# to categorical
factor(data set$Y)
factor(\frac{data}{set}, levels = c(1,2), labels =
c("A", "B"))
# transform values
recode(data set\$Y, "0" = 0, "1" = 50, "2" = 100)
# creates two equal sized groups
ntile(data set$Y, 2)
# convert categorical variable
# to quantitative
as.numeric(data set$Y)
```





Simulation

```
# sample without replacement
sample(data set, 6)
# sample with replacement
resample (data set, 10)
do(3) * resample (data set, 10)
# mixes up values in a variable
shuffle(data set$Y)
# simulate sampling 10000 Ys
# from normal distribution
sim Y <- rnorm(10000, Y stats$mean,</pre>
Y stats$sd)
# put simulated Ys into dataframe
data set<- data.frame(sim Y)</pre>
# simulate sampling distribution of
sim SDoM <- do(10000) * mean(rnorm(157,</pre>
Y stats$mean, Y stats$sd))
# bootstrap sampling distribution of
bootSDoM <- do(10000) *
mean(resample(data set$Y, 157))
# randomize sampling distribution
# of bls, centered on 0
sdob1 <- do(1000) *
 b1(shuffle(Y) ~ X, data = data set)
```

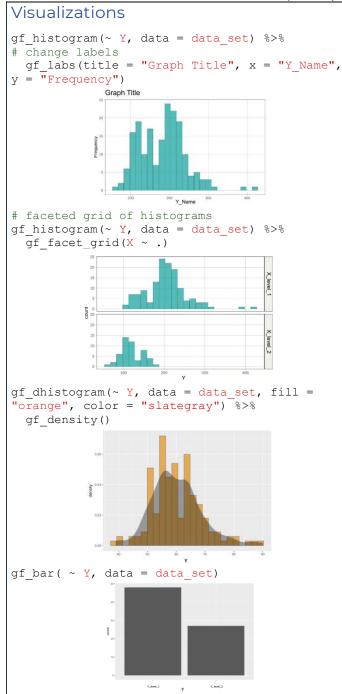
```
# bootstrap sampling distribution of b1s,
# centered on sample b1
sdob1 boot <- do(1000) *
  b1(Y ~ X, data = resample(data_set))
# count the number of b1s at the upper
# and lower extreme
tally(sdob1$b1 > sample b1 |
sdob1$b1 < -sample b1)</pre>
# return TRUE for middle 95% of distribution
middle(sdob1$b1, .95)
# randomize sampling distribution of PREs
sdoPRE \leftarrow do(1000) * PRE(shuffle(Y) \sim X,
data = data set)
# randomize sampling distribution of Fs
sdoF < - do(1000) *
  fVal(shuffle(Y) ~ X, data = data set)
# counts extreme Fs
tally(~fVal > sample F, data = sdoF)
```

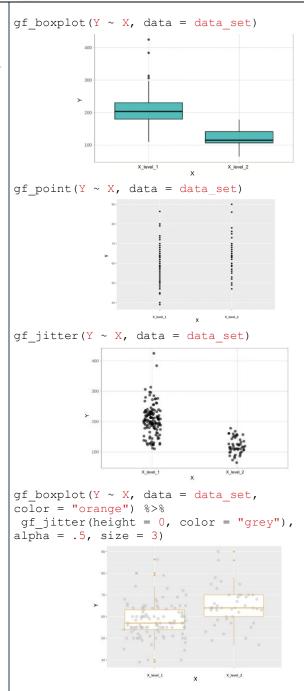
Fitting and Evaluating Models

```
# empty model
empty model <- lm(Y ~ NULL,</pre>
data = data set)
# use one expanatory variable
one model <- lm(Y ~ X, data = data set)</pre>
# create a function from a formula
one model fun <- makeFun(one model)</pre>
one model fun(x level 1)
# model predictions and residuals
data set$empty predict <- predict(empty model)</pre>
data set$empty resid <- resid(empty model)</pre>
# produce ANOVA table
anova (empty model)
supernova(one model)
# t-test, using pooled variance
t.test(Tip ~ Condition, data = data set,
var.equal=TRUE)
# pairwise comparison
# corrections: "Bonferroni" or "none"
pairwise(one model, correction = "none")
```









```
# sampling distribution of b1
gf histogram(\sim b1, data = sdob1,
fill = \simmiddle(b1, .95)) %>%
 \# modify the limits on x- and y-axes
 gf lims (x = c(-12, 12), y = c(0, 70))
gf point(Y ~ X, data = data set) %>%
# add model predictions as red points
gf point(Y \sim X, shape = 1, size = 3,
color = "firebrick") %>%
 # add best fitting model as a red line
 gf model(one model, color = "red")
pairwise(one model, plot = TRUE)
```