

Introduction to R: Core Language Tutorial

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```
source("data-prep.R")
```

Basic Arithmetic and Logical Operations

```
# You can use R like a calculator  
1 + 1 # addition
```

```
## [1] 2
```

```
10 - 9 # subtraction
```

```
## [1] 1
```

```
10 * 10 # multiplcation
```

```
## [1] 100
```

```
100 / 10 # division
```

```
## [1] 10
```

```
10 ^ 2 # exponentiation
```

```
## [1] 100
```

```
abs(-10) # absolute value
```

```
## [1] 10
```

```
ceiling(3.5) # round up to next integer
```

```
## [1] 4
```

```
floor(3.5) # round down to next integer
```

```
## [1] 3
```

```
sqrt(100) # square roots
```

```
## [1] 10
```

```
exp(2) # exponents
```

```
## [1] 7.389056
```

```
pi # mathematical constant pi
```

```
## [1] 3.141593
```

```
exp(1) # mathematical constant e
```

```
## [1] 2.718282
```

```
log(100) # natural logs (i.e., base e)
```

```
## [1] 4.60517
```

```

log(100, base= 10) # base 10 logs

## [1] 2
# Use parentheses to clarify order of operations
(1 + 1 ) * 2

## [1] 4
1 + (1 * 2)

## [1] 3
# You can test for equality
# TRUE and FALSE are keywords
# T and F are synonyms, but are generally discouraged
TRUE

## [1] TRUE
FALSE

## [1] FALSE
1 == 2 # Equality (Return TRUE if equal)

## [1] FALSE
1 != 2 # Inequality (Return FALSE if unequal)

## [1] TRUE
10 > 9 # Greater than

## [1] TRUE
9 < 10 # Less than

## [1] TRUE
10 <= 10 # Less than or equal

## [1] TRUE
2 %in% c(1, 2 ,3) # is the number in the vector

## [1] TRUE
# TRUE and FALSE coerces to 1 and 0 respectively
as.numeric(TRUE)

## [1] 1
as.numeric(FALSE)

## [1] 0
# Logical converting to 0, 1 is useful
x <- c(2, 5 ,7 ,10, 15)
x > 5

## [1] FALSE FALSE TRUE TRUE TRUE
sum(x > 5) # sum of a 0-1 variable is a count

## [1] 3

```

```
mean(x > 5) # mean of a 0-1 variable is a proportion
```

```
## [1] 0.6
```

Basic language features

```
#####
```

```
# Assignment:
```

```
# To assign values to a variable either use <- or =
```

```
# <- is the more common convention in R
```

```
x <- 1 + 1
```

```
x
```

```
## [1] 2
```

```
# = is the common assignment operator in other programming
```

```
# languages. It does work in R, but is not the convention.
```

```
y = 1 + 1
```

```
y
```

```
## [1] 2
```

```
#####
```

```
# Variable name rules:
```

```
# Variable names generally
```

```
# 1. Start with a letter (lower or uppercase)
```

```
# 2. Followed by letters, numbers, underscore (_), or period (.)
```

```
# 3. No spaces
```

```
# These do not work
```

```
# my variable <- 1234
```

```
# 1234variable <- 1234
```

```
# 1234 <- 1234
```

```
# This works
```

```
myvariable <- 1234
```

```
my_variable <- 1234
```

```
my_variable <- 1234
```

```
myvariable123 <- 1234
```

```
myVariable <- 1234
```

```
my.variable <- 1234
```

```
# R has many naming conventions
```

```
# As a matter of preference, style, and convenience, I prefer:
```

```
# 1. Short but descriptive names
```

```
# * Less than 8 characters for names of lists and data.frames
```

```
# * Less than 15 characters for variables names in data.frames
```

```
# 2. Use underscore to separate words within a variable name
```

```
# 3. Avoid upper case letters
```

```
# Names starting with a period are hidden
```

```
.myvariable <- 1234
```

```
ls()
```

```
## [1] "csurvey"      "my_variable"  "my.variable"  "myvariable"
## [5] "myVariable"      "myvariable123" "x"            "y"

ls(all.names = TRUE)

## [1] ".myvariable"  ".Random.seed" "csurvey"      "my_variable"
## [5] "my.variable"  "myvariable"   "myVariable"   "myvariable123"
## [9] "x"            "y"

#####
# Comments:
# Comments are any text on a line following a hash #
# 1. They often appear as the first character of a line
#    to present a whole line comment
# 2. At the end of a common on a line
mean(c(1,2,3,4)) # Example of end of line comment

## [1] 2.5

# 3. Half way through a command at the end of a line
c(1, # Example comment
  2,3, # Another comment
  4)

## [1] 1 2 3 4

#####
# Spaces:
# R will generally permits zero, one or more spaces between
# variables, operators, and other syntactic elements.
# However, appropriate and consistent spacing improves
# the readability of you scripts.
# See Hadley Wickham's style guide:
# http://adv-r.had.co.nz/Style.html

# This is bad but works
x<-c(1,2,3,400)*2
x<-  c ( 1,2,3, 400)* 2

# This is more readable:
# Add spaces after variables, operators, commas
x <- c(1, 2, 3, 400) * 2

#####
# Multiline line commands
# Commands can generally span multiple lines
# as long as R does not think the command has finished

# This works
x <- c("apple",
      "banana")
x

## [1] "apple" "banana"

y <- 10 +
  10 #this works
```

```

y

## [1] 20
# This does not work
y <- 10
  + 10

## [1] 10
y

## [1] 10
#####
# Multiple commands on one line
# You can include more than one command on one line
# by separat the commands by a semicolon.
# But generally, you should avoid doing this as it is not
# very readable.
x <- c(1, 2); y <- c(3, 4); z <- rnorm(10)
x;y;z

## [1] 1 2
## [1] 3 4

## [1] -0.09585945  0.73305089  0.84742665 -0.38502517  0.21117101
## [6] -0.71524025 -0.15603042 -2.00137618 -0.80437769 -0.71095037
#####
# # R is case sensitive
test <- "lower case"
TEST <- "upper case"
TEST

## [1] "upper case"
test # The original value was not lost

## [1] "lower case"
  # because test is different to TEST
Test # This variable does not exist

## Error in eval(expr, envir, enclos): object 'Test' not found
Test <- "title case"
Test

## [1] "title case"
# tip: It's often simpler to make variables lower case
# so that you don't have to think about case.

```

Understanding directories

```

# R has a working directory.
# This is important when loading and saving files to disk
getwd() # show the current working directory

```

```
## [1] "/Users/jeromy/teaching/r-training/acpid-2019-rtraining/materials/introduction-to-r/training-exe

# you can use setwd to change the working directory
# setwd("~/blah/myproject")

# Tip: Open RStudio with the Rproj file then the working directory
# will be the directory containing the Rproj file.

# Tips:
# * Try to avoid spaces in file names
#   (use hyphen or underscore instead)
# * If on Windows, then disable "hide extensions of
#   known file types" (see folder options )
# * If you do use spaces, then you'll need to escape the space with
#   a slash (e.g., ("my\ documents"))
# * Use backslash as the directory separator
# * Store all relevant files for a project within
#   the project working directory
```

The Workspace

```
#####
# R Sessions:

# Quitting R
# You can end the R Session using the q function
# q()

# But if you are in Rstudio, it is simpler to:
# * Just quit RStudio and this will quit the R session
# * Use the session menu in RStudio to Restart or Terminate
#   an R session

#####
# Workspaces and environments:
# list environments
search()

## [1] ".GlobalEnv"      "package:boot"      "package:metafor"
## [4] "package:lavaan"   "package:lme4"       "package:Matrix"
## [7] "package:dplyr"    "package:MASS"       "package:AER"
## [10] "package:sandwich" "package:lmtest"     "package:zoo"
## [13] "package:car"      "package:carData"    "package:Hmisc"
## [16] "package:ggplot2"  "package:Formula"    "package:survival"
## [19] "package:lattice"  "package:psych"      "package:stats"
## [22] "package:graphics" "package:grDevices"  "package:utils"
## [25] "package:datasets" "package:methods"    "Autoloads"
## [28] "package:base"

# Create some objects in the global environment
x <- 1:10
y <- 1:20
```

```

data(mtcars) # Add a built-in dataset mtcars

# Show objects in the global environment
ls()

## [1] "csurvey"      "mtcars"      "my_variable" "my.variable"
## [5] "myvariable"   "myVariable"  "myvariable123" "test"
## [9] "Test"        "TEST"       "x"           "y"
## [13] "z"

# or look at the environment pane in RStudio

#####
# Removing objects:
# Removing named objects with the rm function
rm(x)
ls()

## [1] "csurvey"      "mtcars"      "my_variable" "my.variable"
## [5] "myvariable"   "myVariable"  "myvariable123" "test"
## [9] "Test"        "TEST"       "y"           "z"

rm(y, mtcars)

# Remove all objects from global workspace
# Option 1. Use the following command
rm(list = ls())
# Option 2. Click the broom object in RStudio Environment pane

#####
# Saving objects
# Save all objects in the workspace
# save.image()
save.image(file = "output/everything.rdata")

x <- 30
y <- 1:10
# Save specific named objects using save function.
# rdata or RData is the standard file extension.
save(x, y, file = "output/y.rdata")

# Let's remove x and change y
rm(x)
y <- "changed"
y

## [1] "changed"

# load variables stored in rdata file
load(file = "output/y.rdata")
x

## [1] 30
y

## [1] 1 2 3 4 5 6 7 8 9 10

```

```
# Tips:
# * Try to avoid using save.image() to store temporary calculations
# * Instead, try to write scripts that can be run to return you to
#   your current state of analyses.
```

Data types: Logical, character, numeric

```
#####
# Basic data types
# The most common basic vector types are
x <- c(FALSE, TRUE) # logical vector
y <- c("a", "b", "cat", "dog") # character vector
z1 <- c(100, 1, 2, 3) # numeric integer vector
z2 <- c(100.2, 0.4, 0.9) # numeric real/double vector
class(x); typeof(x); mode(x)

## [1] "logical"
## [1] "logical"
## [1] "logical"
class(y); typeof(y); mode(y)

## [1] "character"
## [1] "character"
## [1] "character"
class(z1); typeof(z1); mode(z1)

## [1] "numeric"
## [1] "double"
## [1] "numeric"
class(z2); typeof(z2); mode(z2)

## [1] "numeric"
## [1] "double"
## [1] "numeric"
# Checking type of object
# there are a range of "is." functions for that return TRUE
# if object is of corresponding type
# apropos("~is\\.")
is.logical(c(TRUE, TRUE))

## [1] TRUE
is.numeric(c("a", "b"))

## [1] FALSE
is.character(c(1, 2, 3))
```



```
## [1] FALSE
#####
# Conversion of Types:
# R has functions that explicitly convert data types
# apropos("^as\\.")
as.character(c(1, 2, 3, 4))

## [1] "1" "2" "3" "4"
as.numeric(c("1", "2a", "3", "four"))

## Warning: NAs introduced by coercion
## [1] 1 NA 3 NA
as.numeric(c(FALSE, FALSE, TRUE, TRUE))

## [1] 0 0 1 1
# R often performs conversions implicitly
sum(c(FALSE, TRUE, TRUE)) # converts logical to 0, 1 numeric

## [1] 2
paste0("v", c(1, 2, 3)) # converts numeric vector to character

## [1] "v1" "v2" "v3"
```

Basic data structures: Vectors, Matrices, Lists, Data.frames

```
#####
# Vectors:
# In R, a single value (scalar) is a vector.
x <- 1 # I.e., x is a vector of length 1

# In addition to importing data,
# R has various functions for creating vectors.
c(1, 2, 3, 4) # c stands for combine

## [1] 1 2 3 4
1:10 # create an integer sequence 1 to 10

## [1] 1 2 3 4 5 6 7 8 9 10
seq(1, 10) # alternative way of creating a sequence

## [1] 1 2 3 4 5 6 7 8 9 10
seq(1, 10, by = 2) # The function has additional options

## [1] 1 3 5 7 9
rep(1, 5) # repeat a value a certain number of times

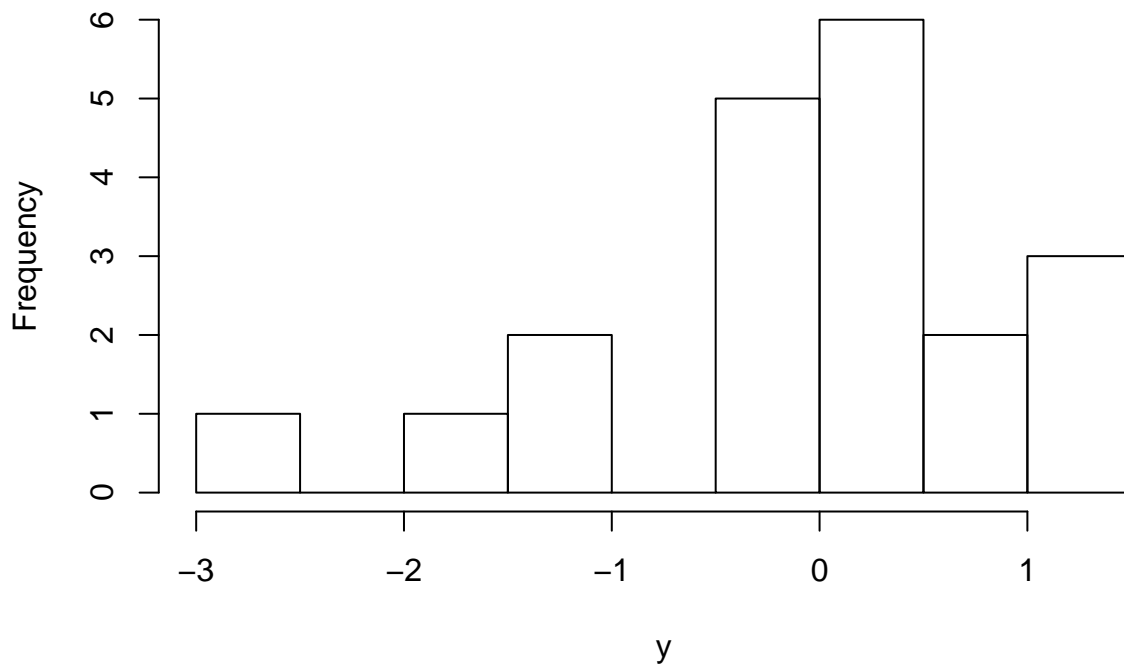
## [1] 1 1 1 1 1
```

```
rep(c(1,2,3), 5) # repeat a value a certain number of times

## [1] 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3
# as well as many simulation functions which we'll cover later
# Initial examples:
# Sample 10 items with replacement from
sample(x = c("happy", "funny", "silly"), size = 10, replace = TRUE)

## [1] "happy" "silly" "happy" "silly" "funny" "funny" "silly" "silly"
## [9] "happy" "happy"
# Sample 20 values from a normal distribution
y <- rnorm(n = 20, mean = 0, sd = 1)
hist(y) # show values in histogram
```

Histogram of y



```
# Vectors can have names
x <- c(1,2,3,4,5)
names(x) <- c("a", "b", "c", "d", "e")
x
```

```
## a b c d e
## 1 2 3 4 5
```

```
# Extracting vectors
x[c(1,2)] # by numeric position
```

```
## a b
## 1 2
```

```
x[x < 3] # by logical vector
```

```
## a b
```

```
## 1 2
x[c("b", "c")] # by name

## b c
## 2 3
#####
# Matrices:
# All data must be of same type (e.g., numeric, character, logical)
y <- matrix(c(1, 2,
              4, 5,
              7, 8 ),
            byrow = TRUE, ncol = 2)
y

##      [,1] [,2]
## [1,]    1    2
## [2,]    4    5
## [3,]    7    8
class(y)

## [1] "matrix"
# number of rows and columns
dim(y) # Number of rows and columns

## [1] 3 2
nrow(y) # Number of rows

## [1] 3
ncol(y) # Number of columns

## [1] 2
# Rows and columns can be given names
rownames(y) <- c("a", "b", "c")
colnames(y) <- c("col1", "col2")

# Rows and columns can be indexed
y["a", ] # By rowname

## col1 col2
##    1    2
y[, "col1"] # By column name

## a b c
## 1 4 7
y["a", "col1"] # By both

## [1] 1
y[c(1,2), ] # By row position

##    col1 col2
## a     1    2
```

```
## b      4      5
y[,1] # By column position

## a b c
## 1 4 7
y[c(2,3), 2] # By column position

## b c
## 5 8

#####
# Lists
# Store arbitrary structures of one or more named elements.
# Elements can be of different lengths
# Lists can contain lists can be nested to create tree like structures
# Lists are commonly used for representing results of analyses

w <- list(apple = c("a", "b", "c"),
          banana = c(1,2),
          carrot = FALSE,
          animals = list(dog = c("dog1", "dog2"),
                        cat = c(TRUE, FALSE)))

class(w)

## [1] "list"
# Accessing one element of list
w$apple # using dollar notation

## [1] "a" "b" "c"
w[[1]] # by position

## [1] "a" "b" "c"
w[["apple"]] # by name (double brackets)

## [1] "a" "b" "c"
# Accessing subset of list
w[c(1, 2)] # by position (single bracket)

## $apple
## [1] "a" "b" "c"
##
## $banana
## [1] 1 2
w[c("apple", "banana")] # by name

## $apple
## [1] "a" "b" "c"
##
## $banana
## [1] 1 2
w[c(FALSE, FALSE, TRUE, TRUE)] # by logical vector
```

```

## $carrot
## [1] FALSE
##
## $animals
## $animals$dog
## [1] "dog1" "dog2"
##
## $animals$cat
## [1] TRUE FALSE

# Quick illustration of a list object returned by
# a statistical function

# We'll simulate some data for two hypothetical groups x and y
# and perform an independent samples t-test.
x <- rnorm(10, mean = 0, sd = 1)
y <- rnorm(10, mean = 1, sd = 1)
fit <- t.test(x, y)

# The function
class(fit) # class does not say list, but it is a list

## [1] "htest"

mode(fit)

## [1] "list"

str(fit) # show structure of object

## List of 9
## $ statistic : Named num -2.29
## .. attr(*, "names")= chr "t"
## $ parameter : Named num 18
## .. attr(*, "names")= chr "df"
## $ p.value : num 0.0341
## $ conf.int : num [1:2] -2.882 -0.126
## .. attr(*, "conf.level")= num 0.95
## $ estimate : Named num [1:2] -0.483 1.021
## .. attr(*, "names")= chr [1:2] "mean of x" "mean of y"
## $ null.value : Named num 0
## .. attr(*, "names")= chr "difference in means"
## $ alternative: chr "two.sided"
## $ method : chr "Welch Two Sample t-test"
## $ data.name : chr "x and y"
## - attr(*, "class")= chr "htest"

names(fit) # show names of elements

## [1] "statistic" "parameter" "p.value" "conf.int" "estimate"
## [6] "null.value" "alternative" "method" "data.name"

# we can view particular elements
fit$statistic

## t
## -2.292472

```

```

fit$parameter

##          df
## 17.99844

fit$p.value

## [1] 0.03414527

# or extract subsets of the list
fit[c("statistic", "parameter", "p.value")]

## $statistic
##          t
## -2.292472
##
## $parameter
##          df
## 17.99844
##
## $p.value
## [1] 0.03414527

#####
# Data Frames:
# Data frames are the standard data structure used for storing
# data. If you have used other software (e.g., SPSS, Excel, etc.),
# this is what you may think of as a "dataset".
# Columns can be of different data types (e.g., character, numeric, logical, etc.)
z <- data.frame(var1 = 1:9, var2 = letters[1:9])
z

##   var1 var2
## 1    1   a
## 2    2   b
## 3    3   c
## 4    4   d
## 5    5   e
## 6    6   f
## 7    7   g
## 8    8   h
## 9    9   i

# Tip: Some functions work with matrices,
# some work with data.frames,
# and some work with both.
# * If you are wanting to store data like you might store in
# a database, then you'll generally want a data.frame.
# * If you are dealing with a mathematical object that you
# want to perform a mathematical operation on, then you generally
# want a matrix (e.g., correlation matrix, covariance matrix,
# distance matrix in MDS, matrices used for matrix algebra).

```

Working with data frames

```
# Let's use the built-in survey data.frame dataset
```

```
library(MASS)
```

```
data(survey)
```

```
?survey
```

```
mydata <- survey
```

```
#####
```

```
# Extracting observations (i.e., rows) and
```

```
# variables (i.e., columns).
```

```
# There are similarities to matrices and lists
```

```
# Select observations
```

```
mydata[1:5, ] # by row number
```

```
##      Sex Wr.Hnd NW.Hnd W.Hnd  Fold Pulse  Clap Exer Smoke Height
## 1 Female  18.5  18.0 Right R on L   92   Left Some Never  173.0
## 2 Male   19.5  20.5 Left  R on L  104   Left None Regul  177.8
## 3 Male   18.0  13.3 Right L on R   87 Neither None Occas    NA
## 4 Male   18.8  18.9 Right R on L   NA Neither None Never  160.0
## 5 Male   20.0  20.0 Right Neither  35   Right Some Never  165.0
##      M.I    Age
## 1 Metric 18.250
## 2 Imperial 17.583
## 3 <NA> 16.917
## 4 Metric 20.333
## 5 Metric 23.667
```

```
mydata[c(5,4,3,2,1), ] # re-order
```

```
##      Sex Wr.Hnd NW.Hnd W.Hnd  Fold Pulse  Clap Exer Smoke Height
## 5 Male   20.0  20.0 Right Neither  35   Right Some Never  165.0
## 4 Male   18.8  18.9 Right R on L   NA Neither None Never  160.0
## 3 Male   18.0  13.3 Right L on R   87 Neither None Occas    NA
## 2 Male   19.5  20.5 Left  R on L  104   Left None Regul  177.8
## 1 Female  18.5  18.0 Right R on L   92   Left Some Never  173.0
##      M.I    Age
## 5 Metric 23.667
## 4 Metric 20.333
## 3 <NA> 16.917
## 2 Imperial 17.583
## 1 Metric 18.250
```

```
mydata[ mydata$Sex == "Female", ] # by logical vector
```

```
##      Sex Wr.Hnd NW.Hnd W.Hnd  Fold Pulse  Clap Exer Smoke Height
## 1 Female  18.5  18.0 Right R on L   92   Left Some Never  173.00
## 6 Female  18.0  17.7 Right L on R   64   Right Some Never  172.72
## 8 Female  17.0  17.3 Right R on L   74   Right Freq Never  157.00
## 11 Female 17.0  17.2 Right L on R   80   Right Freq Never  156.20
## 13 Female 16.0  16.0 Right L on R   NA   Right Some Never  155.00
## 14 Female 19.5  20.2 Right L on R   66 Neither Some Never  155.00
## 16 Female 17.5  17.0 Right R on L   NA   Right Freq Never  156.00
## 17 Female 18.0  18.0 Right L on R   89 Neither Freq Never  157.00
```

## 25	Female	17.0	17.5	Right	R on L	64	Left	Some	Never	NA
## 31	Female	18.5	18.0	Right	R on L	76	Right	None	Occas	NA
## 33	Female	17.1	17.5	Right	R on L	72	Right	Freq	Heavy	166.40
## 37	Female	16.0	16.5	Right	L on R	NA	Right	Some	Never	168.00
## 41	Female	17.5	16.0	Right	L on R	NA	Right	Some	Never	169.00
## 42	Female	17.8	18.0	Right	R on L	72	Right	Some	Never	154.94
## 44	Female	20.1	20.2	Right	L on R	80	Right	Some	Never	176.50
## 45	Female	13.0	13.0	<NA>	L on R	70	Left	Freq	Never	180.34
## 49	Female	18.0	17.6	Right	R on L	60	Right	Some	Occas	168.00
## 50	Female	18.0	17.9	Right	R on L	50	Left	None	Never	165.00
## 57	Female	15.5	15.4	Right	R on L	70	Neither	None	Never	157.48
## 62	Female	18.5	18.2	Right	R on L	72	Neither	Freq	Never	167.64
## 63	Female	19.6	19.7	Right	L on R	70	Right	Freq	Never	178.00
## 64	Female	18.7	18.0	Left	L on R	NA	Left	None	Never	170.00
## 65	Female	17.3	18.0	Right	L on R	64	Neither	Freq	Never	164.00
## 67	Female	19.0	19.1	Right	L on R	NA	Neither	Freq	Never	172.00
## 68	Female	18.5	18.0	Right	R on L	64	Right	Freq	Never	NA
## 71	Female	18.0	17.5	Right	L on R	64	Left	Freq	Never	170.00
## 73	Female	17.0	16.6	Right	R on L	68	Right	Some	Never	171.00
## 74	Female	16.5	17.0	Right	L on R	40	Left	Freq	Never	167.64
## 75	Female	15.6	15.8	Right	R on L	88	Left	Some	Never	165.00
## 76	Female	17.5	17.5	Right	Neither	68	Right	Freq	Heavy	170.00
## 77	Female	17.0	17.6	Right	L on R	76	Right	Some	Never	165.00
## 78	Female	18.6	18.0	Right	L on R	NA	Neither	Freq	Heavy	165.10
## 79	Female	18.3	18.5	Right	R on L	68	Neither	Some	Never	165.10
## 83	Female	17.5	17.5	Right	R on L	98	Left	Freq	Never	NA
## 84	Female	17.0	17.4	Right	R on L	NA	Neither	Some	Never	NA
## 86	Female	17.7	17.0	Right	R on L	76	Right	Some	Never	167.00
## 87	Female	18.2	18.0	Right	L on R	70	Right	Some	Never	162.56
## 88	Female	18.3	18.5	Right	R on L	75	Left	Freq	Never	170.00
## 90	Female	18.0	17.7	Left	R on L	92	Left	Some	Never	NA
## 92	Female	17.5	18.0	Right	Neither	NA	Right	Some	Never	NA
## 93	Female	18.2	17.5	Right	L on R	70	Right	Some	Never	165.00
## 94	Female	18.2	18.5	Right	R on L	NA	Right	Some	Never	168.00
## 96	Female	19.0	18.8	Right	L on R	NA	Right	Some	Never	NA
## 98	Female	17.5	17.5	Right	R on L	60	Right	Freq	Never	166.50
## 100	Female	19.4	19.6	Right	R on L	68	Neither	Freq	Never	175.26
## 103	Female	16.0	16.0	Right	Neither	NA	Right	Some	Never	159.00
## 104	Female	17.5	17.3	Right	R on L	72	Right	Freq	Never	175.00
## 105	Female	17.5	17.0	Right	R on L	80	Left	Some	Heavy	163.00
## 106	Female	19.5	18.5	Right	R on L	80	Right	Some	Never	170.00
## 107	Female	16.2	16.4	Right	R on L	NA	Right	Freq	Occas	172.00
## 108	Female	17.0	15.9	Right	R on L	85	Right	Freq	Never	NA
## 111	Female	18.5	18.5	Right	R on L	76	Left	Freq	Never	175.00
## 113	Female	17.2	16.7	Right	R on L	75	Right	Freq	Never	170.18
## 115	Female	16.0	15.5	Right	L on R	60	Left	Freq	Never	162.56
## 116	Female	16.9	16.0	Right	L on R	70	Right	None	Never	158.00
## 117	Female	17.0	16.7	Right	R on L	70	Right	Some	Never	159.00
## 119	Female	18.5	18.0	Left	L on R	100	Neither	Some	Never	171.00
## 123	Female	18.5	18.0	Right	R on L	92	Right	Freq	Never	172.00
## 127	Female	16.0	16.0	Right	R on L	68	Right	Freq	Never	172.72
## 129	Female	17.5	17.0	Right	R on L	74	Right	Freq	Never	157.00
## 130	Female	16.4	16.5	Right	L on R	90	Right	Some	Never	152.00
## 133	Female	18.9	20.0	Right	R on L	86	Right	Some	Never	NA

## 134 Female	15.4	16.4	Left	L on R	80	Left	Freq	Occas	160.02
## NA <NA>	NA	NA	<NA>	<NA>	NA	<NA>	<NA>	<NA>	NA
## 140 Female	19.5	18.5	Right	L on R	68	Right	None	Never	167.00
## 141 Female	18.0	18.6	Right	R on L	84	Right	Some	Never	175.00
## 142 Female	18.3	19.0	Right	R on L	NA	Right	None	Never	165.00
## 143 Female	19.0	18.8	Right	R on L	65	Right	Freq	Never	172.72
## 145 Female	20.0	19.5	Left	R on L	68	Neither	Freq	Never	172.00
## 149 Female	18.0	18.0	Right	L on R	92	Neither	Freq	Never	165.00
## 150 Female	18.0	18.5	Right	R on L	64	Neither	Freq	Never	164.00
## 152 Female	13.0	12.5	Right	L on R	80	Right	Freq	Never	165.00
## 153 Female	16.3	16.2	Right	L on R	92	Right	Some	Regul	152.40
## 158 Female	18.9	19.2	Right	L on R	74	Right	Some	Never	167.64
## 161 Female	17.5	17.1	Right	R on L	80	Left	None	Never	167.00
## 164 Female	16.5	16.9	Right	R on L	60	Neither	Freq	Occas	169.20
## 166 Female	17.6	17.2	Right	R on L	81	Left	Some	Never	168.00
## 167 Female	19.5	19.2	Right	R on L	70	Right	Some	Never	170.00
## 168 Female	16.5	15.0	Right	L on R	65	Right	Some	Regul	160.02
## 171 Female	16.5	17.0	Right	L on R	NA	Right	Some	Never	168.00
## 173 Female	15.5	15.5	Right	Neither	50	Right	Some	Regul	NA
## 174 Female	18.0	17.5	Right	R on L	48	Neither	Freq	Never	165.00
## 175 Female	17.5	18.0	Right	R on L	68	Neither	Freq	Never	157.48
## 176 Female	19.0	18.5	Left	L on R	104	Left	Freq	Never	170.00
## 178 Female	16.7	17.0	Right	L on R	84	Left	Freq	Never	164.00
## 179 Female	20.5	20.5	Right	R on L	NA	Left	Freq	Regul	NA
## 180 Female	17.0	16.5	Right	R on L	70	Right	Some	Never	162.56
## 182 Female	14.0	13.5	Right	R on L	87	Neither	Freq	Occas	165.10
## 183 Female	17.5	17.6	Right	L on R	79	Right	Some	Never	162.50
## 187 Female	17.0	17.0	Right	L on R	79	Right	Some	Never	163.00
## 194 Female	17.6	17.8	Right	L on R	72	Left	Some	Never	160.02
## 195 Female	16.7	15.1	Right	Neither	NA	Right	None	Never	157.48
## 196 Female	17.0	17.6	Right	L on R	76	Right	Some	Never	165.00
## 197 Female	15.0	13.0	Right	R on L	80	Neither	Freq	Never	170.18
## 199 Female	19.1	19.0	Right	R on L	80	Right	Some	Occas	170.00
## 200 Female	17.5	16.5	Right	R on L	80	Neither	Some	Never	164.00
## 201 Female	16.2	15.8	Right	R on L	61	Right	Some	Occas	167.00
## 203 Female	18.8	17.8	Right	R on L	76	Right	Some	Never	NA
## 204 Female	18.5	18.0	Right	Neither	86	Right	None	Never	160.00
## 206 Female	17.5	17.0	Right	R on L	83	Neither	Freq	Occas	168.00
## 207 Female	17.5	17.6	Right	L on R	76	Right	Some	Never	153.50
## 210 Female	20.8	20.7	Right	R on L	NA	Neither	Freq	Never	171.50
## 211 Female	18.6	18.6	Right	L on R	74	Right	Some	Never	160.00
## 212 Female	17.5	17.5	Left	R on L	83	Neither	Some	Never	163.00
## 215 Female	18.0	17.8	Right	L on R	68	Right	Some	Never	168.90
## 217 Female	16.3	16.2	Right	L on R	NA	Right	None	Never	NA
## 219 Female	17.0	17.3	Right	L on R	NA	Neither	Freq	Never	173.00
## 222 Female	15.9	16.5	Right	R on L	70	Right	Freq	Never	167.64
## 223 Female	17.5	18.4	Right	R on L	88	Right	Some	Never	162.56
## 224 Female	17.5	17.6	Right	L on R	NA	Right	Freq	Never	150.00
## 225 Female	17.6	17.2	Right	L on R	NA	Right	Some	Never	NA
## 226 Female	17.5	17.8	Right	R on L	96	Right	Some	Never	NA
## 227 Female	18.8	18.3	Right	R on L	80	Right	Some	Heavy	170.18
## 229 Female	18.6	18.8	Right	L on R	70	Right	Freq	Regul	167.00
## 231 Female	18.8	18.5	Right	R on L	80	Right	Some	Never	169.00
## 233 Female	18.0	18.0	Right	L on R	85	Right	Some	Never	165.10

##	234	Female	18.5	18.0	Right	L on R	88	Right	Some	Never	160.00
##	235	Female	17.5	16.5	Right	R on L	NA	Right	Some	Never	170.00
##	237	Female	17.6	17.3	Right	R on L	85	Right	Freq	Never	168.50
##		M.I	Age								
##	1	Metric	18.250								
##	6	Imperial	21.000								
##	8	Metric	35.833								
##	11	Imperial	28.500								
##	13	Metric	18.750								
##	14	Metric	17.500								
##	16	Metric	17.167								
##	17	Metric	19.333								
##	25	<NA>	19.167								
##	31	<NA>	41.583								
##	33	Imperial	39.750								
##	37	Metric	19.000								
##	41	Metric	17.500								
##	42	Imperial	17.083								
##	44	Imperial	17.500								
##	45	Imperial	17.417								
##	49	Metric	18.417								
##	50	Metric	30.750								
##	57	Imperial	17.167								
##	62	Imperial	17.333								
##	63	Metric	17.500								
##	64	Metric	19.833								
##	65	Metric	18.583								
##	67	Metric	30.667								
##	68	<NA>	16.917								
##	71	Metric	17.583								
##	73	Metric	17.667								
##	74	Imperial	17.417								
##	75	Metric	17.750								
##	76	Metric	20.667								
##	77	Metric	23.583								
##	78	Imperial	17.167								
##	79	Imperial	17.083								
##	83	<NA>	17.667								
##	84	<NA>	17.167								
##	86	Metric	17.250								
##	87	Imperial	18.000								
##	88	Metric	18.750								
##	90	<NA>	17.583								
##	92	<NA>	18.000								
##	93	Metric	19.667								
##	94	Metric	17.083								
##	96	<NA>	17.083								
##	98	Metric	23.250								
##	100	Imperial	19.083								
##	103	Metric	20.833								
##	104	Metric	20.167								
##	105	Metric	17.667								
##	106	Metric	18.250								
##	107	Metric	17.000								

```

## 108      <NA> 18.500
## 111   Metric 24.167
## 113 Imperial 21.167
## 115 Imperial 17.417
## 116   Metric 20.500
## 117   Metric 22.917
## 119   Metric 18.917
## 123   Metric 17.500
## 127 Imperial 17.667
## 129   Metric 18.000
## 130   Metric 18.333
## 133      <NA> 19.083
## 134 Imperial 18.500
## NA      <NA>      NA
## 140   Metric 18.667
## 141   Metric 17.500
## 142   Metric 21.083
## 143 Imperial 17.250
## 145   Metric 19.167
## 149   Metric 20.000
## 150   Metric 20.167
## 152   Metric 18.167
## 153 Imperial 23.500
## 158 Imperial 44.250
## 161   Metric 18.417
## 164   Metric 29.083
## 166   Metric 18.500
## 167   Metric 18.167
## 168 Imperial 32.750
## 171   Metric 73.000
## 173      <NA> 18.500
## 174   Metric 18.667
## 175 Imperial 17.750
## 176   Metric 17.250
## 178   Metric 23.083
## 179      <NA> 19.250
## 180 Imperial 17.167
## 182 Imperial 17.083
## 183   Metric 17.250
## 187   Metric 24.667
## 194 Imperial 17.250
## 195 Imperial 18.167
## 196   Metric 26.500
## 197 Imperial 17.000
## 199   Metric 19.167
## 200   Metric 17.500
## 201   Metric 19.250
## 203      <NA> 18.583
## 204   Metric 20.167
## 206   Metric 17.083
## 207   Metric 17.417
## 210   Metric 18.500
## 211   Metric 17.167
## 212   Metric 17.250

```

```
## 215 Imperial 17.083
## 217      <NA> 19.250
## 219   Metric 19.167
## 222 Imperial 17.333
## 223 Imperial 18.167
## 224   Metric 20.750
## 225      <NA> 19.917
## 226      <NA> 18.667
## 227 Imperial 18.417
## 229   Metric 20.333
## 231   Metric 18.167
## 233 Imperial 17.667
## 234   Metric 16.917
## 235   Metric 18.583
## 237   Metric 17.750
```

```
mydata[c("1", "2"), ] # by rownames
```

```
##      Sex Wr.Hnd NW.Hnd W.Hnd   Fold Pulse Clap Exer Smoke Height      M.I
## 1 Female   18.5   18.0 Right R on L    92 Left Some Never  173.0   Metric
## 2  Male    19.5   20.5  Left R on L   104 Left None Regul   177.8 Imperial
##      Age
## 1 18.250
## 2 17.583
```

```
# Select variables
```

```
mydata[, c(1,2)] # by position like a matrix
```

```
##      Sex Wr.Hnd
## 1  Female  18.5
## 2   Male   19.5
## 3   Male  18.0
## 4   Male  18.8
## 5   Male  20.0
## 6  Female  18.0
## 7   Male  17.7
## 8  Female  17.0
## 9   Male  20.0
## 10  Male  18.5
## 11  Female  17.0
## 12  Male  21.0
## 13  Female  16.0
## 14  Female  19.5
## 15  Male   16.0
## 16  Female  17.5
## 17  Female  18.0
## 18  Male   19.4
## 19  Male   20.5
## 20  Male   21.0
## 21  Male   21.5
## 22  Male   20.1
## 23  Male   18.5
## 24  Male   21.5
## 25  Female  17.0
## 26  Male   18.5
```

## 27	Male	21.0
## 28	Male	20.8
## 29	Male	17.8
## 30	Male	19.5
## 31	Female	18.5
## 32	Male	18.8
## 33	Female	17.1
## 34	Male	20.1
## 35	Male	18.0
## 36	Male	22.2
## 37	Female	16.0
## 38	Male	19.4
## 39	Male	22.0
## 40	Male	19.0
## 41	Female	17.5
## 42	Female	17.8
## 43	Male	NA
## 44	Female	20.1
## 45	Female	13.0
## 46	Male	17.0
## 47	Male	23.2
## 48	Male	22.5
## 49	Female	18.0
## 50	Female	18.0
## 51	Male	22.0
## 52	Male	20.5
## 53	Male	17.0
## 54	Male	20.5
## 55	Male	22.5
## 56	Male	18.5
## 57	Female	15.5
## 58	Male	19.5
## 59	Male	19.5
## 60	Male	20.6
## 61	Male	22.8
## 62	Female	18.5
## 63	Female	19.6
## 64	Female	18.7
## 65	Female	17.3
## 66	Male	19.5
## 67	Female	19.0
## 68	Female	18.5
## 69	Male	19.0
## 70	Male	21.0
## 71	Female	18.0
## 72	Male	19.4
## 73	Female	17.0
## 74	Female	16.5
## 75	Female	15.6
## 76	Female	17.5
## 77	Female	17.0
## 78	Female	18.6
## 79	Female	18.3
## 80	Male	20.0

## 81	Male	19.5
## 82	Male	19.2
## 83	Female	17.5
## 84	Female	17.0
## 85	Male	23.0
## 86	Female	17.7
## 87	Female	18.2
## 88	Female	18.3
## 89	Male	18.0
## 90	Female	18.0
## 91	Male	20.5
## 92	Female	17.5
## 93	Female	18.2
## 94	Female	18.2
## 95	Male	21.3
## 96	Female	19.0
## 97	Male	20.0
## 98	Female	17.5
## 99	Male	19.5
## 100	Female	19.4
## 101	Male	21.9
## 102	Male	18.9
## 103	Female	16.0
## 104	Female	17.5
## 105	Female	17.5
## 106	Female	19.5
## 107	Female	16.2
## 108	Female	17.0
## 109	Male	17.5
## 110	Male	19.7
## 111	Female	18.5
## 112	Male	19.2
## 113	Female	17.2
## 114	Male	20.5
## 115	Female	16.0
## 116	Female	16.9
## 117	Female	17.0
## 118	Male	23.0
## 119	Female	18.5
## 120	Male	21.0
## 121	Male	20.0
## 122	Male	22.5
## 123	Female	18.5
## 124	Male	19.8
## 125	Male	18.5
## 126	Male	19.3
## 127	Female	16.0
## 128	Male	18.8
## 129	Female	17.5
## 130	Female	16.4
## 131	Male	22.0
## 132	Male	19.0
## 133	Female	18.9
## 134	Female	15.4

## 135	Male	17.9
## 136	Male	23.1
## 137	<NA>	19.8
## 138	Male	22.0
## 139	Male	20.0
## 140	Female	19.5
## 141	Female	18.0
## 142	Female	18.3
## 143	Female	19.0
## 144	Male	21.4
## 145	Female	20.0
## 146	Male	18.5
## 147	Male	22.5
## 148	Male	19.5
## 149	Female	18.0
## 150	Female	18.0
## 151	Male	21.8
## 152	Female	13.0
## 153	Female	16.3
## 154	Male	21.5
## 155	Male	18.9
## 156	Male	20.5
## 157	Male	14.0
## 158	Female	18.9
## 159	Male	20.0
## 160	Male	18.5
## 161	Female	17.5
## 162	Male	18.1
## 163	Male	20.2
## 164	Female	16.5
## 165	Male	19.1
## 166	Female	17.6
## 167	Female	19.5
## 168	Female	16.5
## 169	Male	19.0
## 170	Male	19.0
## 171	Female	16.5
## 172	Male	20.5
## 173	Female	15.5
## 174	Female	18.0
## 175	Female	17.5
## 176	Female	19.0
## 177	Male	20.5
## 178	Female	16.7
## 179	Female	20.5
## 180	Female	17.0
## 181	Male	19.0
## 182	Female	14.0
## 183	Female	17.5
## 184	Male	18.5
## 185	Male	18.0
## 186	Male	20.5
## 187	Female	17.0
## 188	Male	18.5

```
## 189 Male 18.0
## 190 Male 18.5
## 191 Male 20.0
## 192 Male 22.0
## 193 Male 17.9
## 194 Female 17.6
## 195 Female 16.7
## 196 Female 17.0
## 197 Female 15.0
## 198 Male 16.0
## 199 Female 19.1
## 200 Female 17.5
## 201 Female 16.2
## 202 Male 21.0
## 203 Female 18.8
## 204 Female 18.5
## 205 Male 17.0
## 206 Female 17.5
## 207 Female 17.5
## 208 Male 17.5
## 209 Male 17.5
## 210 Female 20.8
## 211 Female 18.6
## 212 Female 17.5
## 213 Male 18.0
## 214 Male 17.0
## 215 Female 18.0
## 216 Male 19.5
## 217 Female 16.3
## 218 Male 18.2
## 219 Female 17.0
## 220 Male 23.2
## 221 Male 23.2
## 222 Female 15.9
## 223 Female 17.5
## 224 Female 17.5
## 225 Female 17.6
## 226 Female 17.5
## 227 Female 18.8
## 228 Male 20.0
## 229 Female 18.6
## 230 Male 18.6
## 231 Female 18.8
## 232 Male 18.0
## 233 Female 18.0
## 234 Female 18.5
## 235 Female 17.5
## 236 Male 21.0
## 237 Female 17.6
```

```
mydata[c(1,2)] # by position like a list
```

```
##      Sex Wr.Hnd
## 1 Female 18.5
## 2 Male 19.5
```


## 3	Male	18.0
## 4	Male	18.8
## 5	Male	20.0
## 6	Female	18.0
## 7	Male	17.7
## 8	Female	17.0
## 9	Male	20.0
## 10	Male	18.5
## 11	Female	17.0
## 12	Male	21.0
## 13	Female	16.0
## 14	Female	19.5
## 15	Male	16.0
## 16	Female	17.5
## 17	Female	18.0
## 18	Male	19.4
## 19	Male	20.5
## 20	Male	21.0
## 21	Male	21.5
## 22	Male	20.1
## 23	Male	18.5
## 24	Male	21.5
## 25	Female	17.0
## 26	Male	18.5
## 27	Male	21.0
## 28	Male	20.8
## 29	Male	17.8
## 30	Male	19.5
## 31	Female	18.5
## 32	Male	18.8
## 33	Female	17.1
## 34	Male	20.1
## 35	Male	18.0
## 36	Male	22.2
## 37	Female	16.0
## 38	Male	19.4
## 39	Male	22.0
## 40	Male	19.0
## 41	Female	17.5
## 42	Female	17.8
## 43	Male	NA
## 44	Female	20.1
## 45	Female	13.0
## 46	Male	17.0
## 47	Male	23.2
## 48	Male	22.5
## 49	Female	18.0
## 50	Female	18.0
## 51	Male	22.0
## 52	Male	20.5
## 53	Male	17.0
## 54	Male	20.5
## 55	Male	22.5
## 56	Male	18.5

## 57	Female	15.5
## 58	Male	19.5
## 59	Male	19.5
## 60	Male	20.6
## 61	Male	22.8
## 62	Female	18.5
## 63	Female	19.6
## 64	Female	18.7
## 65	Female	17.3
## 66	Male	19.5
## 67	Female	19.0
## 68	Female	18.5
## 69	Male	19.0
## 70	Male	21.0
## 71	Female	18.0
## 72	Male	19.4
## 73	Female	17.0
## 74	Female	16.5
## 75	Female	15.6
## 76	Female	17.5
## 77	Female	17.0
## 78	Female	18.6
## 79	Female	18.3
## 80	Male	20.0
## 81	Male	19.5
## 82	Male	19.2
## 83	Female	17.5
## 84	Female	17.0
## 85	Male	23.0
## 86	Female	17.7
## 87	Female	18.2
## 88	Female	18.3
## 89	Male	18.0
## 90	Female	18.0
## 91	Male	20.5
## 92	Female	17.5
## 93	Female	18.2
## 94	Female	18.2
## 95	Male	21.3
## 96	Female	19.0
## 97	Male	20.0
## 98	Female	17.5
## 99	Male	19.5
## 100	Female	19.4
## 101	Male	21.9
## 102	Male	18.9
## 103	Female	16.0
## 104	Female	17.5
## 105	Female	17.5
## 106	Female	19.5
## 107	Female	16.2
## 108	Female	17.0
## 109	Male	17.5
## 110	Male	19.7

## 111	Female	18.5
## 112	Male	19.2
## 113	Female	17.2
## 114	Male	20.5
## 115	Female	16.0
## 116	Female	16.9
## 117	Female	17.0
## 118	Male	23.0
## 119	Female	18.5
## 120	Male	21.0
## 121	Male	20.0
## 122	Male	22.5
## 123	Female	18.5
## 124	Male	19.8
## 125	Male	18.5
## 126	Male	19.3
## 127	Female	16.0
## 128	Male	18.8
## 129	Female	17.5
## 130	Female	16.4
## 131	Male	22.0
## 132	Male	19.0
## 133	Female	18.9
## 134	Female	15.4
## 135	Male	17.9
## 136	Male	23.1
## 137	<NA>	19.8
## 138	Male	22.0
## 139	Male	20.0
## 140	Female	19.5
## 141	Female	18.0
## 142	Female	18.3
## 143	Female	19.0
## 144	Male	21.4
## 145	Female	20.0
## 146	Male	18.5
## 147	Male	22.5
## 148	Male	19.5
## 149	Female	18.0
## 150	Female	18.0
## 151	Male	21.8
## 152	Female	13.0
## 153	Female	16.3
## 154	Male	21.5
## 155	Male	18.9
## 156	Male	20.5
## 157	Male	14.0
## 158	Female	18.9
## 159	Male	20.0
## 160	Male	18.5
## 161	Female	17.5
## 162	Male	18.1
## 163	Male	20.2
## 164	Female	16.5

##	165	Male	19.1
##	166	Female	17.6
##	167	Female	19.5
##	168	Female	16.5
##	169	Male	19.0
##	170	Male	19.0
##	171	Female	16.5
##	172	Male	20.5
##	173	Female	15.5
##	174	Female	18.0
##	175	Female	17.5
##	176	Female	19.0
##	177	Male	20.5
##	178	Female	16.7
##	179	Female	20.5
##	180	Female	17.0
##	181	Male	19.0
##	182	Female	14.0
##	183	Female	17.5
##	184	Male	18.5
##	185	Male	18.0
##	186	Male	20.5
##	187	Female	17.0
##	188	Male	18.5
##	189	Male	18.0
##	190	Male	18.5
##	191	Male	20.0
##	192	Male	22.0
##	193	Male	17.9
##	194	Female	17.6
##	195	Female	16.7
##	196	Female	17.0
##	197	Female	15.0
##	198	Male	16.0
##	199	Female	19.1
##	200	Female	17.5
##	201	Female	16.2
##	202	Male	21.0
##	203	Female	18.8
##	204	Female	18.5
##	205	Male	17.0
##	206	Female	17.5
##	207	Female	17.5
##	208	Male	17.5
##	209	Male	17.5
##	210	Female	20.8
##	211	Female	18.6
##	212	Female	17.5
##	213	Male	18.0
##	214	Male	17.0
##	215	Female	18.0
##	216	Male	19.5
##	217	Female	16.3
##	218	Male	18.2

```
## 219 Female 17.0
## 220 Male 23.2
## 221 Male 23.2
## 222 Female 15.9
## 223 Female 17.5
## 224 Female 17.5
## 225 Female 17.6
## 226 Female 17.5
## 227 Female 18.8
## 228 Male 20.0
## 229 Female 18.6
## 230 Male 18.6
## 231 Female 18.8
## 232 Male 18.0
## 233 Female 18.0
## 234 Female 18.5
## 235 Female 17.5
## 236 Male 21.0
## 237 Female 17.6
```

```
mydata[,c("Sex", "Fold")] # by name like a matrix
```

```
##      Sex      Fold
## 1  Female R on L
## 2   Male R on L
## 3   Male L on R
## 4   Male R on L
## 5   Male Neither
## 6  Female L on R
## 7   Male L on R
## 8  Female R on L
## 9   Male R on L
## 10  Male R on L
## 11  Female L on R
## 12  Male R on L
## 13  Female L on R
## 14  Female L on R
## 15  Male R on L
## 16  Female R on L
## 17  Female L on R
## 18  Male R on L
## 19  Male L on R
## 20  Male R on L
## 21  Male R on L
## 22  Male L on R
## 23  Male L on R
## 24  Male R on L
## 25  Female R on L
## 26  Male Neither
## 27  Male R on L
## 28  Male R on L
## 29  Male L on R
## 30  Male L on R
## 31  Female R on L
## 32  Male L on R
```

## 33	Female	R on L
## 34	Male	R on L
## 35	Male	L on R
## 36	Male	L on R
## 37	Female	L on R
## 38	Male	R on L
## 39	Male	R on L
## 40	Male	R on L
## 41	Female	L on R
## 42	Female	R on L
## 43	Male	R on L
## 44	Female	L on R
## 45	Female	L on R
## 46	Male	R on L
## 47	Male	L on R
## 48	Male	R on L
## 49	Female	R on L
## 50	Female	R on L
## 51	Male	R on L
## 52	Male	L on R
## 53	Male	L on R
## 54	Male	L on R
## 55	Male	R on L
## 56	Male	L on R
## 57	Female	R on L
## 58	Male	R on L
## 59	Male	L on R
## 60	Male	L on R
## 61	Male	R on L
## 62	Female	R on L
## 63	Female	L on R
## 64	Female	L on R
## 65	Female	L on R
## 66	Male	Neither
## 67	Female	L on R
## 68	Female	R on L
## 69	Male	L on R
## 70	Male	L on R
## 71	Female	L on R
## 72	Male	R on L
## 73	Female	R on L
## 74	Female	L on R
## 75	Female	R on L
## 76	Female	Neither
## 77	Female	L on R
## 78	Female	L on R
## 79	Female	R on L
## 80	Male	L on R
## 81	Male	R on L
## 82	Male	R on L
## 83	Female	R on L
## 84	Female	R on L
## 85	Male	L on R
## 86	Female	R on L

87 Female L on R
88 Female R on L
89 Male Neither
90 Female R on L
91 Male R on L
92 Female Neither
93 Female L on R
94 Female R on L
95 Male R on L
96 Female L on R
97 Male R on L
98 Female R on L
99 Male Neither
100 Female R on L
101 Male R on L
102 Male L on R
103 Female Neither
104 Female R on L
105 Female R on L
106 Female R on L
107 Female R on L
108 Female R on L
109 Male L on R
110 Male R on L
111 Female R on L
112 Male L on R
113 Female R on L
114 Male R on L
115 Female L on R
116 Female L on R
117 Female R on L
118 Male L on R
119 Female L on R
120 Male L on R
121 Male R on L
122 Male L on R
123 Female R on L
124 Male L on R
125 Male L on R
126 Male R on L
127 Female R on L
128 Male L on R
129 Female R on L
130 Female L on R
131 Male R on L
132 Male L on R
133 Female R on L
134 Female L on R
135 Male R on L
136 Male L on R
137 <NA> L on R
138 Male L on R
139 Male L on R
140 Female L on R

141 Female R on L
142 Female R on L
143 Female R on L
144 Male L on R
145 Female R on L
146 Male R on L
147 Male L on R
148 Male R on L
149 Female L on R
150 Female R on L
151 Male R on L
152 Female L on R
153 Female L on R
154 Male R on L
155 Male L on R
156 Male R on L
157 Male L on R
158 Female L on R
159 Male R on L
160 Male L on R
161 Female R on L
162 Male Neither
163 Male L on R
164 Female R on L
165 Male Neither
166 Female R on L
167 Female R on L
168 Female L on R
169 Male L on R
170 Male R on L
171 Female L on R
172 Male L on R
173 Female Neither
174 Female R on L
175 Female R on L
176 Female L on R
177 Male Neither
178 Female L on R
179 Female R on L
180 Female R on L
181 Male R on L
182 Female R on L
183 Female L on R
184 Male L on R
185 Male Neither
186 Male R on L
187 Female L on R
188 Male R on L
189 Male R on L
190 Male Neither
191 Male R on L
192 Male L on R
193 Male R on L
194 Female L on R


```

## 195 Female Neither
## 196 Female L on R
## 197 Female R on L
## 198 Male Neither
## 199 Female R on L
## 200 Female R on L
## 201 Female R on L
## 202 Male L on R
## 203 Female R on L
## 204 Female Neither
## 205 Male R on L
## 206 Female R on L
## 207 Female L on R
## 208 Male R on L
## 209 Male L on R
## 210 Female R on L
## 211 Female L on R
## 212 Female R on L
## 213 Male R on L
## 214 Male R on L
## 215 Female L on R
## 216 Male Neither
## 217 Female L on R
## 218 Male R on L
## 219 Female L on R
## 220 Male L on R
## 221 Male L on R
## 222 Female R on L
## 223 Female R on L
## 224 Female L on R
## 225 Female L on R
## 226 Female R on L
## 227 Female R on L
## 228 Male L on R
## 229 Female L on R
## 230 Male L on R
## 231 Female R on L
## 232 Male R on L
## 233 Female L on R
## 234 Female L on R
## 235 Female R on L
## 236 Male R on L
## 237 Female R on L

```

```
mydata[c("Sex", "Fold")] #
```

```

##      Sex      Fold
## 1  Female R on L
## 2   Male R on L
## 3   Male L on R
## 4   Male R on L
## 5   Male Neither
## 6  Female L on R
## 7   Male L on R
## 8  Female R on L

```

## 9	Male	R on L
## 10	Male	R on L
## 11	Female	L on R
## 12	Male	R on L
## 13	Female	L on R
## 14	Female	L on R
## 15	Male	R on L
## 16	Female	R on L
## 17	Female	L on R
## 18	Male	R on L
## 19	Male	L on R
## 20	Male	R on L
## 21	Male	R on L
## 22	Male	L on R
## 23	Male	L on R
## 24	Male	R on L
## 25	Female	R on L
## 26	Male	Neither
## 27	Male	R on L
## 28	Male	R on L
## 29	Male	L on R
## 30	Male	L on R
## 31	Female	R on L
## 32	Male	L on R
## 33	Female	R on L
## 34	Male	R on L
## 35	Male	L on R
## 36	Male	L on R
## 37	Female	L on R
## 38	Male	R on L
## 39	Male	R on L
## 40	Male	R on L
## 41	Female	L on R
## 42	Female	R on L
## 43	Male	R on L
## 44	Female	L on R
## 45	Female	L on R
## 46	Male	R on L
## 47	Male	L on R
## 48	Male	R on L
## 49	Female	R on L
## 50	Female	R on L
## 51	Male	R on L
## 52	Male	L on R
## 53	Male	L on R
## 54	Male	L on R
## 55	Male	R on L
## 56	Male	L on R
## 57	Female	R on L
## 58	Male	R on L
## 59	Male	L on R
## 60	Male	L on R
## 61	Male	R on L
## 62	Female	R on L

63 Female L on R
64 Female L on R
65 Female L on R
66 Male Neither
67 Female L on R
68 Female R on L
69 Male L on R
70 Male L on R
71 Female L on R
72 Male R on L
73 Female R on L
74 Female L on R
75 Female R on L
76 Female Neither
77 Female L on R
78 Female L on R
79 Female R on L
80 Male L on R
81 Male R on L
82 Male R on L
83 Female R on L
84 Female R on L
85 Male L on R
86 Female R on L
87 Female L on R
88 Female R on L
89 Male Neither
90 Female R on L
91 Male R on L
92 Female Neither
93 Female L on R
94 Female R on L
95 Male R on L
96 Female L on R
97 Male R on L
98 Female R on L
99 Male Neither
100 Female R on L
101 Male R on L
102 Male L on R
103 Female Neither
104 Female R on L
105 Female R on L
106 Female R on L
107 Female R on L
108 Female R on L
109 Male L on R
110 Male R on L
111 Female R on L
112 Male L on R
113 Female R on L
114 Male R on L
115 Female L on R
116 Female L on R

```

## 117 Female R on L
## 118 Male L on R
## 119 Female L on R
## 120 Male L on R
## 121 Male R on L
## 122 Male L on R
## 123 Female R on L
## 124 Male L on R
## 125 Male L on R
## 126 Male R on L
## 127 Female R on L
## 128 Male L on R
## 129 Female R on L
## 130 Female L on R
## 131 Male R on L
## 132 Male L on R
## 133 Female R on L
## 134 Female L on R
## 135 Male R on L
## 136 Male L on R
## 137 <NA> L on R
## 138 Male L on R
## 139 Male L on R
## 140 Female L on R
## 141 Female R on L
## 142 Female R on L
## 143 Female R on L
## 144 Male L on R
## 145 Female R on L
## 146 Male R on L
## 147 Male L on R
## 148 Male R on L
## 149 Female L on R
## 150 Female R on L
## 151 Male R on L
## 152 Female L on R
## 153 Female L on R
## 154 Male R on L
## 155 Male L on R
## 156 Male R on L
## 157 Male L on R
## 158 Female L on R
## 159 Male R on L
## 160 Male L on R
## 161 Female R on L
## 162 Male Neither
## 163 Male L on R
## 164 Female R on L
## 165 Male Neither
## 166 Female R on L
## 167 Female R on L
## 168 Female L on R
## 169 Male L on R
## 170 Male R on L

```

171 Female L on R
172 Male L on R
173 Female Neither
174 Female R on L
175 Female R on L
176 Female L on R
177 Male Neither
178 Female L on R
179 Female R on L
180 Female R on L
181 Male R on L
182 Female R on L
183 Female L on R
184 Male L on R
185 Male Neither
186 Male R on L
187 Female L on R
188 Male R on L
189 Male R on L
190 Male Neither
191 Male R on L
192 Male L on R
193 Male R on L
194 Female L on R
195 Female Neither
196 Female L on R
197 Female R on L
198 Male Neither
199 Female R on L
200 Female R on L
201 Female R on L
202 Male L on R
203 Female R on L
204 Female Neither
205 Male R on L
206 Female R on L
207 Female L on R
208 Male R on L
209 Male L on R
210 Female R on L
211 Female L on R
212 Female R on L
213 Male R on L
214 Male R on L
215 Female L on R
216 Male Neither
217 Female L on R
218 Male R on L
219 Female L on R
220 Male L on R
221 Male L on R
222 Female R on L
223 Female R on L
224 Female L on R

```
## 225 Female L on R
## 226 Female R on L
## 227 Female R on L
## 228 Male L on R
## 229 Female L on R
## 230 Male L on R
## 231 Female R on L
## 232 Male R on L
## 233 Female L on R
## 234 Female L on R
## 235 Female R on L
## 236 Male R on L
## 237 Female R on L
```

```
mydata$Sex # by name to get a single variable
```

```
## [1] Female Male Male Male Male Female Male Female Male Male
## [11] Female Male Female Female Male Female Female Male Male Male
## [21] Male Male Male Male Female Male Male Male Male Male
## [31] Female Male Female Male Male Male Female Male Male Male
## [41] Female Female Male Female Female Male Male Male Female Female
## [51] Male Male Male Male Male Male Female Male Male Male
## [61] Male Female Female Female Female Male Female Female Male Male
## [71] Female Male Female Female Female Female Female Female Female Male
## [81] Male Male Female Female Male Female Female Female Male Female
## [91] Male Female Female Female Male Female Male Female Male Female
## [101] Male Male Female Female Female Female Female Female Male Male
## [111] Female Male Female Male Female Female Female Male Female Male
## [121] Male Male Female Male Male Male Female Male Female Female
## [131] Male Male Female Female Male Male <NA> Male Male Female
## [141] Female Female Female Male Female Male Male Male Female Female
## [151] Male Female Female Male Male Male Male Female Male Male
## [161] Female Male Male Female Male Female Female Female Male Male
## [171] Female Male Female Female Female Female Male Female Female Female
## [181] Male Female Female Male Male Male Female Male Male Male
## [191] Male Male Male Female Female Female Female Male Female Female
## [201] Female Male Female Female Male Female Female Male Male Female
## [211] Female Female Male Male Female Male Female Male Female Male
## [221] Male Female Female Female Female Female Female Male Female Male
## [231] Female Male Female Female Female Male Female
## Levels: Female Male
```

```
#####
```

```
# Names
```

```
names(mydata) # get variable names
```

```
## [1] "Sex" "Wr.Hnd" "NW.Hnd" "W.Hnd" "Fold" "Pulse" "Clap"
## [8] "Exer" "Smoke" "Height" "M.I" "Age"
```

```
colnames(mydata) # but this also works
```

```
## [1] "Sex" "Wr.Hnd" "NW.Hnd" "W.Hnd" "Fold" "Pulse" "Clap"
## [8] "Exer" "Smoke" "Height" "M.I" "Age"
```

```
rownames(mydata) # rows can also have names
```

```
## [1] "1" "2" "3" "4" "5" "6" "7" "8" "9" "10" "11"
```

```
## [12] "12" "13" "14" "15" "16" "17" "18" "19" "20" "21" "22"
## [23] "23" "24" "25" "26" "27" "28" "29" "30" "31" "32" "33"
## [34] "34" "35" "36" "37" "38" "39" "40" "41" "42" "43" "44"
## [45] "45" "46" "47" "48" "49" "50" "51" "52" "53" "54" "55"
## [56] "56" "57" "58" "59" "60" "61" "62" "63" "64" "65" "66"
## [67] "67" "68" "69" "70" "71" "72" "73" "74" "75" "76" "77"
## [78] "78" "79" "80" "81" "82" "83" "84" "85" "86" "87" "88"
## [89] "89" "90" "91" "92" "93" "94" "95" "96" "97" "98" "99"
## [100] "100" "101" "102" "103" "104" "105" "106" "107" "108" "109" "110"
## [111] "111" "112" "113" "114" "115" "116" "117" "118" "119" "120" "121"
## [122] "122" "123" "124" "125" "126" "127" "128" "129" "130" "131" "132"
## [133] "133" "134" "135" "136" "137" "138" "139" "140" "141" "142" "143"
## [144] "144" "145" "146" "147" "148" "149" "150" "151" "152" "153" "154"
## [155] "155" "156" "157" "158" "159" "160" "161" "162" "163" "164" "165"
## [166] "166" "167" "168" "169" "170" "171" "172" "173" "174" "175" "176"
## [177] "177" "178" "179" "180" "181" "182" "183" "184" "185" "186" "187"
## [188] "188" "189" "190" "191" "192" "193" "194" "195" "196" "197" "198"
## [199] "199" "200" "201" "202" "203" "204" "205" "206" "207" "208" "209"
## [210] "210" "211" "212" "213" "214" "215" "216" "217" "218" "219" "220"
## [221] "221" "222" "223" "224" "225" "226" "227" "228" "229" "230" "231"
## [232] "232" "233" "234" "235" "236" "237"
```

Tip: Avoid row names.

Add another variable to the data.frame to store this information.

Examine first few rows

`head(mydata)` *# first 6 rows*

```
##      Sex Wr.Hnd NW.Hnd W.Hnd   Fold Pulse   Clap Exer Smoke Height
## 1 Female  18.5   18.0 Right R on L   92    Left Some Never 173.00
## 2  Male  19.5   20.5 Left  R on L  104    Left None Regul 177.80
## 3  Male  18.0   13.3 Right L on R   87 Neither None Occas    NA
## 4  Male  18.8   18.9 Right R on L   NA Neither None Never 160.00
## 5  Male  20.0   20.0 Right Neither  35    Right Some Never 165.00
## 6 Female  18.0   17.7 Right L on R   64    Right Some Never 172.72
##      M.I    Age
## 1  Metric 18.250
## 2 Imperial 17.583
## 3    <NA> 16.917
## 4  Metric 20.333
## 5  Metric 23.667
## 6 Imperial 21.000
```

`head(mydata, n = 10)` *# first 7 rows*

```
##      Sex Wr.Hnd NW.Hnd W.Hnd   Fold Pulse   Clap Exer Smoke Height
## 1 Female  18.5   18.0 Right R on L   92    Left Some Never 173.00
## 2  Male  19.5   20.5 Left  R on L  104    Left None Regul 177.80
## 3  Male  18.0   13.3 Right L on R   87 Neither None Occas    NA
## 4  Male  18.8   18.9 Right R on L   NA Neither None Never 160.00
## 5  Male  20.0   20.0 Right Neither  35    Right Some Never 165.00
## 6 Female  18.0   17.7 Right L on R   64    Right Some Never 172.72
## 7  Male  17.7   17.7 Right L on R   83    Right Freq Never 182.88
## 8 Female  17.0   17.3 Right R on L   74    Right Freq Never 157.00
## 9  Male  20.0   19.5 Right R on L   72    Right Some Never 175.00
```

```
## 10 Male 18.5 18.5 Right R on L 90 Right Some Never 167.00
## M.I Age
## 1 Metric 18.250
## 2 Imperial 17.583
## 3 <NA> 16.917
## 4 Metric 20.333
## 5 Metric 23.667
## 6 Imperial 21.000
## 7 Imperial 18.833
## 8 Metric 35.833
## 9 Metric 19.000
## 10 Metric 22.333
```

```
tail(mydata) # last few rows
```

```
## Sex Wr.Hnd NW.Hnd W.Hnd Fold Pulse Clap Exer Smoke Height
## 232 Male 18.0 16.0 Right R on L NA Right Some Never 180.34
## 233 Female 18.0 18.0 Right L on R 85 Right Some Never 165.10
## 234 Female 18.5 18.0 Right L on R 88 Right Some Never 160.00
## 235 Female 17.5 16.5 Right R on L NA Right Some Never 170.00
## 236 Male 21.0 21.5 Right R on L 90 Right Some Never 183.00
## 237 Female 17.6 17.3 Right R on L 85 Right Freq Never 168.50
## M.I Age
## 232 Imperial 20.750
## 233 Imperial 17.667
## 234 Metric 16.917
## 235 Metric 18.583
## 236 Metric 17.167
## 237 Metric 17.750
```

```
# View(mydata) # Rstudio function to open data in viewer
# or click on the icon in the Environment pane
```

```
# How many rows and columns?
```

```
dim(mydata) # rows and column counts
```

```
## [1] 237 12
```

```
nrow(mydata) # row count
```

```
## [1] 237
```

```
ncol(mydata) # column count
```

```
## [1] 12
```

```
# Examine structure
```

```
str(mydata)
```

```
## 'data.frame': 237 obs. of 12 variables:
## $ Sex : Factor w/ 2 levels "Female","Male": 1 2 2 2 2 1 2 1 2 2 ...
## $ Wr.Hnd: num 18.5 19.5 18 18.8 20 18 17.7 17 20 18.5 ...
## $ NW.Hnd: num 18 20.5 13.3 18.9 20 17.7 17.7 17.3 19.5 18.5 ...
## $ W.Hnd : Factor w/ 2 levels "Left","Right": 2 1 2 2 2 2 2 2 2 2 ...
## $ Fold : Factor w/ 3 levels "L on R","Neither",...: 3 3 1 3 2 1 1 3 3 3 ...
## $ Pulse : int 92 104 87 NA 35 64 83 74 72 90 ...
## $ Clap : Factor w/ 3 levels "Left","Neither",...: 1 1 2 2 3 3 3 3 3 3 ...
## $ Exer : Factor w/ 3 levels "Freq","None",...: 3 2 2 2 3 3 1 1 3 3 ...
```



```
## $ Smoke : Factor w/ 4 levels "Heavy","Never",...: 2 4 3 2 2 2 2 2 2 ...
## $ Height: num 173 178 NA 160 165 ...
## $ M.I : Factor w/ 2 levels "Imperial","Metric": 2 1 NA 2 2 1 1 2 2 ...
## $ Age : num 18.2 17.6 16.9 20.3 23.7 ...
```

Getting help

```
# Use question mark (i.e., ?) followed by command name
# to lookup specific command
?mean
help(mean) # or use help function

# to look up package
help(package = "MASS")

# Press F1 in RStudio on the command name
# mean

# Use double question mark to do a full-text search on R help
??"factor analysis"

# Search google
# e.g., how to get the mean of a vector using r

# Ask question on Stackoverflow with the R tag
# http://stackoverflow.com/questions/tagged/r
```

Exercise 1

```
# 1. Working with vectors
# 1.1 Create a variable called x with 10 values drawn from a
# normal distribution (see rnorm)

# 1.2 Use the sum and > operator to work out how many values in x
# are larger than 1

# 3. Using the survey dataset in the MASS package
library(MASS)
data(survey)
# 3.1 Look up the help file on MASS

# 3.2 How many observations are there?

# 3.3 Show the first 10 rows of the cats data.frame

# 3.4 Show the structure of cats using the str function

# 3.5 Extract the female cats and assign to variable fcats
```

```
# 3.6 How many rows is in fcats?
```

Answers 1

```
# 1. Working with vectors
# 1.1 Create a variable called x with 10 values drawn from a
#     normal distribution (see rnorm)
x <- rnorm(10)

# 1.2 Use the sum and > operator to work out how many values in x
#     are larger than 1
sum(x > 1)
```

```
## [1] 0
```

```
# 3. Using the cats dataset in the MASS package
library(MASS)
data(survey)
# 3.1 Look up the help file on survey
?survey
```

```
# 3.2 How many observations are there?
nrow(survey)
```

```
## [1] 237
```

```
# 3.3 Show the first 10 rows of the survey data.frame
head(survey, 10)
```

```
##      Sex Wr.Hnd NW.Hnd W.Hnd  Fold Pulse  Clap Exer Smoke Height
## 1 Female  18.5  18.0 Right R on L   92   Left Some Never 173.00
## 2 Male    19.5  20.5 Left  R on L  104   Left None Regul 177.80
## 3 Male    18.0  13.3 Right L on R   87 Neither None Occas    NA
## 4 Male    18.8  18.9 Right R on L   NA Neither None Never 160.00
## 5 Male    20.0  20.0 Right Neither  35   Right Some Never 165.00
## 6 Female  18.0  17.7 Right L on R   64   Right Some Never 172.72
## 7 Male    17.7  17.7 Right L on R   83   Right Freq Never 182.88
## 8 Female  17.0  17.3 Right R on L   74   Right Freq Never 157.00
## 9 Male    20.0  19.5 Right R on L   72   Right Some Never 175.00
## 10 Male   18.5  18.5 Right R on L   90   Right Some Never 167.00
##      M.I    Age
## 1   Metric 18.250
## 2 Imperial 17.583
## 3    <NA> 16.917
## 4   Metric 20.333
## 5   Metric 23.667
## 6 Imperial 21.000
## 7 Imperial 18.833
## 8   Metric 35.833
## 9   Metric 19.000
## 10  Metric 22.333
```

```
# 3.4 Show the structure of survey using the str function
str(survey)
```

```
## 'data.frame':    237 obs. of  12 variables:
## $ Sex      : Factor w/ 2 levels "Female","Male": 1 2 2 2 2 1 2 1 2 2 ...
## $ Wr.Hnd: num  18.5 19.5 18 18.8 20 18 17.7 17 20 18.5 ...
## $ NW.Hnd: num  18 20.5 13.3 18.9 20 17.7 17.7 17.3 19.5 18.5 ...
## $ W.Hnd  : Factor w/ 2 levels "Left","Right": 2 1 2 2 2 2 2 2 2 2 ...
## $ Fold   : Factor w/ 3 levels "L on R","Neither",...: 3 3 1 3 2 1 1 3 3 3 ...
## $ Pulse  : int   92 104 87 NA 35 64 83 74 72 90 ...
## $ Clap   : Factor w/ 3 levels "Left","Neither",...: 1 1 2 2 3 3 3 3 3 3 ...
## $ Exer   : Factor w/ 3 levels "Freq","None",...: 3 2 2 2 3 3 1 1 3 3 ...
## $ Smoke  : Factor w/ 4 levels "Heavy","Never",...: 2 4 3 2 2 2 2 2 2 2 ...
## $ Height: num   173 178 NA 160 165 ...
## $ M.I    : Factor w/ 2 levels "Imperial","Metric": 2 1 NA 2 2 1 1 2 2 2 ...
## $ Age    : num   18.2 17.6 16.9 20.3 23.7 ...
```

```
# 3.5 Extract the female survey and assign to variable fsurvey
fsurvey <- survey[ survey$Sex == "F", ]
```

```
# 3.6 How many rows is in fsurvey?
nrow(fsurvey)
```

```
## [1] 1
```

Packages

```
# R has many additional packages
# To use a package it needs to be installed.
# You only need to install a package once.
# To use a package, you need to load the package each time
# you use R.
```

```
#####
# Installing an R package
# Option 1. Use the install.packages function.
# install.packages("psych")
```

```
# Option 2. Use the package tab in R Studio
# Click install and enter package details
```

```
#####
# Loading an installed package
# Option 1. Use the library function
library(psych) # I.e., put this at the start of your script
```

```
# Other options
# 2. We may talk about ProjectTemplate later
#
```

```

# 3. Put it in your R startup file
#   (not recommended as it reduces reproducibility)

#####
# Common errors
# Not having a package installed is a common error
# If you try to load a package that is not installed.
# e.g.,
# library(foo)
# You will get an error
# Error in library(foo) : there is no package called 'foo'
# This means either
# 1. You mistyped the name of the package, or
# 2. You need to install the package
#   install.packages("foo")
# Note foo is just an example. There is no package called foo.

# Trying to use a function from a package that is not loaded is a common error

# E.g., there is a function you want to use
detach(package:psych) # used for example to ensure psych is not attached
# say we wanted to use the fisherz function from the psych package
# but the psych package is not loaded
fisherz(.3)

```

```
## Error in fisherz(0.3): could not find function "fisherz"
```

```

# We get the error:
# "Error: object 'fisherz' not found"
# Thus we need to run
library(psych)

```

```

## Warning: package 'psych' was built under R version 3.5.2

##
## Attaching package: 'psych'

## The following object is masked from 'package:boot':
##
##   logit

## The following object is masked from 'package:lavaan':
##
##   cor2cov

## The following object is masked from 'package:car':
##
##   logit

## The following object is masked from 'package:Hmisc':
##
##   describe

## The following objects are masked from 'package:ggplot2':
##
##   %+%, alpha

```

```
fisherz(.3)
```

```
## [1] 0.3095196
```

```
#####
```

```
# Packages contain additional functions.
```

```
# Once the package is loaded, functions are added to the workspace
```

```
# list workspace
```

```
search()
```

```
## [1] ".GlobalEnv"      "package:psych"      "package:boot"
## [4] "package:metafor"   "package:lavaan"     "package:lme4"
## [7] "package:Matrix"    "package:dplyr"       "package:MASS"
## [10] "package:AER"        "package:sandwich"    "package:lmtest"
## [13] "package:zoo"        "package:car"         "package:carData"
## [16] "package:Hmisc"     "package:ggplot2"     "package:Formula"
## [19] "package:survival"   "package:lattice"     "package:stats"
## [22] "package:graphics"  "package:grDevices"   "package:utils"
## [25] "package:datasets"  "package:methods"     "Autoloads"
## [28] "package:base"
```

```
# To make it clear that a function comes from a particular package
```

```
# or to overcome the issue where two packages have functions with the same names
```

```
# use double colon (i.e., package::function).
```

```
# RStudio also permits auto-completion of function names.
```

```
psych::alpha # alpha is a function in the psych package
```

```
## function (x, keys = NULL, cumulative = FALSE, title = NULL, max = 10,
##     na.rm = TRUE, check.keys = FALSE, n.iter = 1, delete = TRUE,
##     use = "pairwise", warnings = TRUE, n.obs = NULL, impute = NULL)
## {
##     alpha.1 <- function(C, R) {
##         n <- dim(C)[2]
##         alpha.raw <- (1 - tr(C)/sum(C)) * (n/(n - 1))
##         sumR <- sum(R)
##         alpha.std <- (1 - n/sumR) * (n/(n - 1))
##         smc.R <- smc(R)
##         G6 <- (1 - (n - sum(smc.R))/sumR)
##         av.r <- (sumR - n)/(n * (n - 1))
##         R.adj <- R[lower.tri(R)]
##         var.r <- var(R.adj, na.rm = TRUE)
##         med.r <- median(R.adj, na.rm = TRUE)
##         mod1 <- matrix(av.r, n, n)
##         Res1 <- R - mod1
##         GF1 = 1 - sum(Res1^2)/sum(R^2)
##         Rd <- R - diag(R)
##         diag(Res1) <- 0
##         GF1.off <- 1 - sum(Res1^2)/sum(Rd^2)
##         sn <- n * av.r/(1 - av.r)
##         Q = (2 * n^2/((n - 1)^2 * (sum(C)^3))) * (sum(C) * (tr(C %*%
##             C) + (tr(C))^2) - 2 * (tr(C) * sum(C %*% C)))
##         result <- list(raw = alpha.raw, std = alpha.std, G6 = G6,
##             av.r = av.r, sn = sn, Q = Q, GF1, GF1.off, var.r = var.r,
##             med.r = med.r)
##         return(result)
##     }
## }
```

```

##     }
##     cl <- match.call()
##     if (!is.matrix(x) && !is.data.frame(x))
##       stop("Data must either be a data frame or a matrix")
##     if (class(x)[1] != "data.frame")
##       x <- fix.dplyr(x)
##     if (!is.null(keys)) {
##       if (is.list(keys)) {
##         select <- sub("-", "", unlist(keys))
##         x <- x[, select]
##         keys <- make.keys(x, keys)
##       }
##       else {
##         temp <- rep(1, ncol(x))
##         temp[(colnames(x) %in% keys)] <- -1
##         keys <- temp
##       }
##     }
##     nvar <- dim(x)[2]
##     nsub <- dim(x)[1]
##     scores <- NULL
##     response.freq <- NULL
##     raw <- FALSE
##     if (!isCorrelation(x)) {
##       raw <- TRUE
##       if (!is.null(impute)) {
##         if (impute == "median") {
##           item.impute <- apply(x, 2, median, na.rm = na.rm)
##         }
##         else {
##           item.impute <- apply(x, 2, mean, na.rm = na.rm)
##         }
##         for (i in 1:nsub) {
##           for (j in 1:nvar) {
##             x[i, is.na(x[i, j])] <- item.impute[j]
##           }
##         }
##       }
##       item.var <- apply(x, 2, sd, na.rm = na.rm)
##       bad <- which((item.var <= 0) | is.na(item.var))
##       if ((length(bad) > 0) && delete) {
##         for (baddy in 1:length(bad)) {
##           warning("Item = ", colnames(x)[bad][baddy], " had no variance and was deleted")
##         }
##         x <- x[, -bad]
##         nvar <- nvar - length(bad)
##       }
##       response.freq <- response.frequencies(x, max = max)
##       C <- cov(x, use = use)
##     }
##     else {
##       C <- x
##     }
##   }
##   if (is.null(colnames(x)))

```

```

##      colnames(x) <- paste0("V", 1:nvar)
##      p1 <- principal(x, scores = FALSE)
##      if (any(p1$loadings < 0)) {
##          if (check.keys) {
##              if (warnings)
##                  warning("Some items were negatively correlated with total scale and were automatical.
##              keys <- 1 - 2 * (p1$loadings < 0)
##          }
##          else {
##              if (is.null(keys) && warnings) {
##                  warning("Some items were negatively correlated with the total scale and probably \nsl
##                  if (warnings)
##                      cat("Some items (", rownames(p1$loadings)[(p1$loadings <
##                      0)], ") were negatively correlated with the total scale and \nprobably should be
##                      keys <- rep(1, nvar)
##              }
##          }
##      }
##      if (is.null(keys)) {
##          keys <- rep(1, nvar)
##          names(keys) <- colnames(x)
##      }
##      else {
##          keys <- as.vector(keys)
##          names(keys) <- colnames(x)
##          if (length(keys) < nvar) {
##              temp <- keys
##              keys <- rep(1, nvar)
##              names(keys) <- colnames(x)
##              keys[temp] <- -1
##          }
##      }
##      key.d <- diag(keys)
##      C <- key.d %*% C %*% key.d
##      signkey <- strtrim(keys, 1)
##      signkey[signkey == "1"] <- ""
##      colnames(x) <- paste(colnames(x), signkey, sep = "")
##      if (raw) {
##          if (any(keys < 0)) {
##              min.item <- min(x, na.rm = na.rm)
##              max.item <- max(x, na.rm = na.rm)
##              adjust <- max.item + min.item
##              flip.these <- which(keys < 0)
##              x[, flip.these] <- adjust - x[, flip.these]
##          }
##          if (cumulative) {
##              total <- rowSums(x, na.rm = na.rm)
##          }
##          else {
##              total <- rowMeans(x, na.rm = na.rm)
##          }
##          mean.t <- mean(total, na.rm = na.rm)
##          sdev <- sd(total, na.rm = na.rm)
##          raw.r <- cor(total, x, use = use)

```

```

##      t.valid <- colSums(!is.na(x))
##    }
##    else {
##      total <- NULL
##      totals <- TRUE
##    }
##    R <- cov2cor(C)
##    drop.item <- vector("list", nvar)
##    alpha.total <- alpha.1(C, R)
##    if (nvar > 2) {
##      for (i in 1:nvar) {
##        drop.item[[i]] <- alpha.1(C[-i, -i, drop = FALSE],
##          R[-i, -i, drop = FALSE])
##      }
##    }
##    else {
##      drop.item[[1]] <- drop.item[[2]] <- c(rep(R[1, 2], 2),
##        smc(R)[1], R[1, 2], NA, NA, NA, NA)
##    }
##    by.item <- data.frame(matrix(unlist(drop.item), ncol = 10,
##      byrow = TRUE))
##    if (max(nsub, n.obs) > nvar) {
##      by.item[6] <- sqrt(by.item[6]/(max(nsub, n.obs)))
##      by.item <- by.item[-c(7:8)]
##      colnames(by.item) <- c("raw_alpha", "std.alpha", "G6(smc)",
##        "average_r", "S/N", "alpha se", "var.r", "med.r")
##    }
##    else {
##      by.item <- by.item[-c(6:8)]
##      colnames(by.item) <- c("raw_alpha", "std.alpha", "G6(smc)",
##        "average_r", "S/N", "var.r", "med.r")
##    }
##    rownames(by.item) <- colnames(x)
##    Vt <- sum(R)
##    item.r <- colSums(R)/sqrt(Vt)
##    RC <- R
##    diag(RC) <- smc(R)
##    Vtc <- sum(RC)
##    item.rc <- colSums(RC)/sqrt(Vtc)
##    if (nvar > 1) {
##      r.drop <- rep(0, nvar)
##      for (i in 1:nvar) {
##        v.drop <- sum(C[-i, -i, drop = FALSE])
##        c.drop <- sum(C[, i]) - C[i, i]
##        r.drop[i] <- c.drop/sqrt(C[i, i] * v.drop)
##      }
##    }
##    item.means <- colMeans(x, na.rm = na.rm)
##    item.sd <- apply(x, 2, sd, na.rm = na.rm)
##    if (raw) {
##      Unidim <- alpha.total[7]
##      var.r <- alpha.total[[9]]
##      Fit.off <- alpha.total[8]
##      ase = sqrt(alpha.total$Q/nsub)

```



```

##      alpha.total <- data.frame(alpha.total[1:5], ase = ase,
##                               mean = mean.t, sd = sdev, med.r = alpha.total[10])
##      colnames(alpha.total) <- c("raw_alpha", "std.alpha",
##                                "G6(smc)", "average_r", "S/N", "ase", "mean", "sd",
##                                "median_r")
##      rownames(alpha.total) <- ""
##      stats <- data.frame(n = t.valid, raw.r = t(raw.r), std.r = item.r,
##                          r.cor = item.rc, r.drop = r.drop, mean = item.means,
##                          sd = item.sd)
##    }
##  else {
##    if (is.null(n.obs)) {
##      Unidim <- alpha.total[7]
##      Fit.off <- alpha.total[8]
##      var.r <- alpha.total[9]
##      med.r <- alpha.total[10]
##      alpha.total <- data.frame(alpha.total[c(1:5, 10)])
##      colnames(alpha.total) <- c("raw_alpha", "std.alpha",
##                                "G6(smc)", "average_r", "S/N", "median_r")
##    }
##    else {
##      Unidim <- alpha.total[7]
##      Fit.off <- alpha.total[8]
##      var.r <- alpha.total[9]
##      alpha.total <- data.frame(alpha.total[1:5], ase = sqrt(alpha.total$Q/n.obs),
##                                alpha.total[10])
##      colnames(alpha.total) <- c("raw_alpha", "std.alpha",
##                                "G6(smc)", "average_r", "S/N", "ase", "median_r")
##    }
##    rownames(alpha.total) <- ""
##    stats <- data.frame(r = item.r, r.cor = item.rc, r.drop = r.drop)
##  }
##  rownames(stats) <- colnames(x)
##  if (n.iter > 1) {
##    if (!raw) {
##      message("bootstrapped confidence intervals require raw data")
##      boot <- NULL
##      boot.ci <- NULL
##    }
##    else {
##      boot <- vector("list", n.iter)
##      boot <- mclapply(1:n.iter, function(XX) {
##        xi <- x[sample.int(nsub, replace = TRUE), ]
##        C <- cov(xi, use = "pairwise")
##        if (!is.null(keys)) {
##          key.d <- diag(keys)
##          xi <- key.d %*% C %*% key.d
##        }
##        R <- cov2cor(C)
##        alpha.1(C, R)
##      })
##      boot <- matrix(unlist(boot), ncol = 10, byrow = TRUE)
##      colnames(boot) <- c("raw_alpha", "std.alpha", "G6(smc)",
##                          "average_r", "s/n", "ase", "Unidim", "Goodfit",

```

```
##             "var.r", "median.r")
##             boot.ci <- quantile(boot[, 1], c(0.025, 0.5, 0.975))
##         }
##     }
##     else {
##         boot = NULL
##         boot.ci <- NULL
##     }
##     names(Unidim) <- "Unidim"
##     names(Fit.off) <- "Fit.off"
##     result <- list(total = alpha.total, alpha.drop = by.item,
##         item.stats = stats, response.freq = response.freq, keys = keys,
##         scores = total, nvar = nvar, boot.ci = boot.ci, boot = boot,
##         Unidim = Unidim, var.r = var.r, Fit = Fit.off, call = cl,
##         title = title)
##     class(result) <- c("psych", "alpha")
##     return(result)
## }
## <bytecode: 0x7fad83e020c0>
## <environment: namespace:psych>
```

Missing data

```
# Missing data is represented in R by NA
```

```
x <- c(1, 2, NA, 4)
y <- c("a", "b", NA, "c")
x
```

```
## [1] 1 2 NA 4
```

```
y
```

```
## [1] "a" "b" NA "c"
```

```
# To see whether a value is missing
```

```
is.na(x)
```

```
## [1] FALSE FALSE TRUE FALSE
```

```
# If you have missing data, some functions will return NA by default
# rather than returning a value
```

```
mean(x)
```

```
## [1] NA
```

```
sd(x)
```

```
## [1] NA
```

```
# Many functions have a na.rm argument
# rm stands for "remove"
```

```
mean(x, na.rm = TRUE)
```

```
## [1] 2.333333
```

```
sd(x, na.rm = TRUE)
```

```
## [1] 1.527525
```

```

# or you remove the missing data
na.omit(x)

## [1] 1 2 4
## attr(,"na.action")
## [1] 3
## attr(,"class")
## [1] "omit"

mean(na.omit(x))

## [1] 2.333333

# na.omit also works on data frames performing listwise deletion
head(survey)

##      Sex Wr.Hnd NW.Hnd W.Hnd   Fold Pulse   Clap Exer Smoke Height
## 1 Female  18.5   18.0 Right R on L   92   Left Some Never 173.00
## 2 Male   19.5   20.5 Left  R on L  104   Left None Regul 177.80
## 3 Male   18.0   13.3 Right L on R   87 Neither None Occas   NA
## 4 Male   18.8   18.9 Right R on L   NA Neither None Never 160.00
## 5 Male   20.0   20.0 Right Neither 35   Right Some Never 165.00
## 6 Female 18.0   17.7 Right L on R   64   Right Some Never 172.72
##      M.I   Age
## 1 Metric 18.250
## 2 Imperial 17.583
## 3 <NA> 16.917
## 4 Metric 20.333
## 5 Metric 23.667
## 6 Imperial 21.000

dim(survey)

## [1] 237 12

cleaned_survey <- na.omit(survey)
dim(cleaned_survey)

## [1] 168 12

```

Getting summaries of data frames

```

library(MASS) # user survey data from MASS package
data(survey) # load an internal dataset
mydata <- survey

# Variable Names
names(mydata)

## [1] "Sex"    "Wr.Hnd" "NW.Hnd" "W.Hnd"  "Fold"   "Pulse"  "Clap"
## [8] "Exer"   "Smoke"  "Height" "M.I"    "Age"

# Show structure
str(mydata)

## 'data.frame': 237 obs. of 12 variables:

```

```
## $ Sex : Factor w/ 2 levels "Female","Male": 1 2 2 2 2 1 2 1 2 2 ...
## $ Wr.Hnd: num 18.5 19.5 18 18.8 20 18 17.7 17 20 18.5 ...
## $ NW.Hnd: num 18 20.5 13.3 18.9 20 17.7 17.7 17.3 19.5 18.5 ...
## $ W.Hnd : Factor w/ 2 levels "Left","Right": 2 1 2 2 2 2 2 2 2 ...
## $ Fold : Factor w/ 3 levels "L on R","Neither",...: 3 3 1 3 2 1 1 3 3 3 ...
## $ Pulse : int 92 104 87 NA 35 64 83 74 72 90 ...
## $ Clap : Factor w/ 3 levels "Left","Neither",...: 1 1 2 2 3 3 3 3 3 ...
## $ Exer : Factor w/ 3 levels "Freq","None",...: 3 2 2 2 3 3 1 1 3 3 ...
## $ Smoke : Factor w/ 4 levels "Heavy","Never",...: 2 4 3 2 2 2 2 2 2 ...
## $ Height: num 173 178 NA 160 165 ...
## $ M.I : Factor w/ 2 levels "Imperial","Metric": 2 1 NA 2 2 1 1 2 2 ...
## $ Age : num 18.2 17.6 16.9 20.3 23.7 ...
```

Useful summary of numeric and categorical variables

```
Hmisc::describe(mydata)
```

```
## mydata
##
## 12 Variables      237 Observations
## -----
## Sex
##      n missing distinct
##    236      1      2
##
## Value      Female      Male
## Frequency    118    118
## Proportion    0.5    0.5
## -----
## Wr.Hnd
##      n missing distinct      Info      Mean      Gmd      .05      .10
##    236      1      60    0.997    18.67    2.09    16.00    16.50
##      .25      .50      .75      .90      .95
##    17.50    18.50    19.80    21.15    22.05
##
## lowest : 13.0 14.0 15.0 15.4 15.5, highest: 22.5 22.8 23.0 23.1 23.2
## -----
## NW.Hnd
##      n missing distinct      Info      Mean      Gmd      .05      .10
##    236      1      68    0.998    18.58    2.184    15.50    16.30
##      .25      .50      .75      .90      .95
##    17.50    18.50    19.72    21.00    22.22
##
## lowest : 12.5 13.0 13.3 13.5 15.0, highest: 22.7 23.0 23.2 23.3 23.5
## -----
## W.Hnd
##      n missing distinct
##    236      1      2
##
## Value      Left Right
## Frequency    18    218
## Proportion 0.076 0.924
## -----
## Fold
##      n missing distinct
##    237      0      3
```

```

##
## Value      L on R Neither R on L
## Frequency      99      18      120
## Proportion    0.418    0.076    0.506
## -----
## Pulse
##      n missing distinct      Info      Mean      Gmd      .05      .10
##     192      45      43    0.997    74.15    13.07    59.55    60.00
##      .25      .50      .75      .90      .95
##     66.00    72.50    80.00    90.00    92.00
##
## lowest : 35 40 48 50 54, highest: 96 97 98 100 104
## -----
## Clap
##      n missing distinct
##     236      1      3
##
## Value      Left Neither Right
## Frequency      39      50      147
## Proportion    0.165    0.212    0.623
## -----
## Exer
##      n missing distinct
##     237      0      3
##
## Value      Freq None Some
## Frequency    115    24    98
## Proportion 0.485 0.101 0.414
## -----
## Smoke
##      n missing distinct
##     236      1      4
##
## Value      Heavy Never Occas Regul
## Frequency      11    189    19    17
## Proportion 0.047 0.801 0.081 0.072
## -----
## Height
##      n missing distinct      Info      Mean      Gmd      .05      .10
##     209      28      67    0.999    172.4    11.2    157.0    160.0
##      .25      .50      .75      .90      .95
##    165.0    171.0    180.0    185.4    189.6
##
## lowest : 150.00 152.00 152.40 153.50 154.94, highest: 191.80 193.04 195.00 196.00 200.00
## -----
## M.I
##      n missing distinct
##     209      28      2
##
## Value      Imperial Metric
## Frequency      68      141
## Proportion    0.325    0.675
## -----
## Age

```

```
##          n missing distinct      Info      Mean      Gmd      .05      .10
##        237         0       88    0.999    20.37    4.353    17.08    17.22
##        .25      .50      .75      .90      .95
##       17.67    18.58    20.17    23.58    30.68
##
## lowest : 16.750 16.917 17.000 17.083 17.167, highest: 41.583 43.833 44.250 70.417 73.000
## -----
```

```
# Common univariate statistics for numeric variables
psych::describe(mydata)
```

```
##      vars   n  mean    sd median trimmed  mad   min   max range  skew
## Sex*     1 236   1.50  0.50   1.50    1.50  0.74   1.00   2.0   1.00  0.00
## Wr.Hnd    2 236  18.67  1.88  18.50   18.61  1.48  13.00  23.2  10.20  0.18
## NW.Hnd    3 236  18.58  1.97  18.50   18.55  1.63  12.50  23.5   11.00  0.02
## W.Hnd*    4 236   1.92  0.27   2.00    2.00  0.00   1.00   2.0   1.00 -3.17
## Fold*     5 237   2.09  0.96   3.00    2.11  0.00   1.00   3.0   2.00 -0.18
## Pulse     6 192  74.15 11.69  72.50   74.02 11.12  35.00 104.0  69.00 -0.02
## Clap*     7 236   2.46  0.76   3.00    2.57  0.00   1.00   3.0   2.00 -0.98
## Exer*     8 237   1.93  0.95   2.00    1.91  1.48   1.00   3.0   2.00  0.14
## Smoke*    9 236   2.18  0.62   2.00    2.07  0.00   1.00   4.0   3.00  1.67
## Height   10 209 172.38  9.85 171.00  172.19 10.08 150.00 200.0  50.00  0.22
## M.I*     11 209   1.67  0.47   2.00    1.72  0.00   1.00   2.0   1.00 -0.74
## Age      12 237  20.37  6.47  18.58   18.99  1.61  16.75  73.0  56.25  5.16
##      kurtosis   se
## Sex*        -2.01 0.03
## Wr.Hnd         0.30 0.12
## NW.Hnd         0.44 0.13
## W.Hnd*         8.10 0.02
## Fold*        -1.89 0.06
## Pulse          0.33 0.84
## Clap*        -0.60 0.05
## Exer*        -1.88 0.06
## Smoke*         3.21 0.04
## Height       -0.44 0.68
## M.I*         -1.46 0.03
## Age          33.47 0.42
```

Summaries of numeric vectors (or data frame variables)

```
x <- c(1, 2, 3, 4,5)

# Total
sum(x) # sum of vector

## [1] 15

length(x) # length of vector (e.g., sample size)

## [1] 5

# Central tendency
mean(x) # mean of vector
```

```
## [1] 3
median(x) # median of vector

## [1] 3
# Spread
sd(x) # standard deviation

## [1] 1.581139
var(x) # variance

## [1] 2.5
range(x) # min and max of vector

## [1] 1 5
min(x) # minimum of vector

## [1] 1
max(x) # max of vector

## [1] 5
# Other distributional features
psych::skew(x) # skewness

## [1] 0
psych::kurtosi(x) # kurtosis

## [1] -1.912
dat <- data.frame(x = c(1, 2, 3, 4, 5),
                  y = c(0, 0, 1, 1, 1))
dat

##      x y
## 1 1 0
## 2 2 0
## 3 3 1
## 4 4 1
## 5 5 1
# Vector operations typically operate element wise
dat$z <- dat$x + dat$y
dat

##      x y z
## 1 1 0 1
## 2 2 0 2
## 3 3 1 4
## 4 4 1 5
## 5 5 1 6
dat$z <- dat$x * dat$y
dat

##      x y z
## 1 1 0 0
```

```
## 2 2 0 0
## 3 3 1 3
## 4 4 1 4
## 5 5 1 5

# A single value is recycled through the vector
dat$z <- dat$x + 10
dat

##   x y z
## 1 1 0 11
## 2 2 0 12
## 3 3 1 13
## 4 4 1 14
## 5 5 1 15
```

Exercise 2 - Data.frames

```
# For this exercise will use the GSS7402 dataset
library(AER)
help(package = AER)
data("GSS7402")
?GSS7402 # to learn about the dataset
# It might be easier to work with a shorter variable name
gss <- GSS7402

# 1. List the variable names in the gss dataset

# 2. Show the first few rows (hint: the head) of the dataset?

# 3. How many cases are there?

# 4. What is the mean, sd, and range age of the sample

# 5. Use the psych and Hmisc describe functions to describe the samples

# 6. Extract a data.frame with only people over the age of 80

# 7. Get the mean number of children ("kids") for participants
#    over the age of 80

# 8. Use the mean function to get the mean age at first birth.
#    Hint: there is missing data.
```

Answers Exercise 2 - Data.frames

```
# For this exercise will use the GSS7402 dataset
library(AER)
help(package = AER)
data("GSS7402")
?GSS7402 # to learn about the dataset
```



```
# It might be easier to work with a shorter variable name
gss <- GSS7402
```

```
# 1. List the variable names in the gss dataset
```

```
names(gss)
```

```
## [1] "kids"      "age"      "education" "year"
## [5] "siblings"  "agefirstbirth" "ethnicity" "city16"
## [9] "lowincome16" "immigrant"
```

```
# 2. Show the first few rows (hint: the head) of the dataset?
```

```
head(gss)
```

```
## kids age education year siblings agefirstbirth ethnicity city16
## 1 0 25 14 2002 1 NA cauc no
## 2 1 30 13 2002 4 19 cauc yes
## 3 1 55 2 2002 1 27 cauc no
## 4 2 57 16 2002 1 22 cauc no
## 5 2 71 12 2002 6 29 cauc yes
## 6 0 19 13 2002 1 NA other yes
## lowincome16 immigrant
## 1 no no
## 2 no no
## 3 no yes
## 4 no no
## 5 no no
## 6 no no
```

```
# 3. How many cases are there?
```

```
nrow(gss)
```

```
## [1] 9120
```

```
# 4. What is the mean, sd, and range age of the sample
```

```
mean(gss$age)
```

```
## [1] 46.08202
```

```
sd(gss$age)
```

```
## [1] 17.92389
```

```
range(gss$age)
```

```
## [1] 18 89
```

```
# 5. Use the psych and Hmisc describe functions to describe the samples
```

```
psych::describe(gss)
```

```
## vars n mean sd median trimmed mad min max range
## kids 1 9120 2.08 1.81 2 1.86 1.48 0 8 8
## age 2 9120 46.08 17.92 43 44.94 19.27 18 89 71
## education 3 9120 12.64 2.96 12 12.70 2.97 0 20 20
## year 4 9120 1990.29 9.10 1994 1990.79 11.86 1974 2002 28
## siblings 5 9120 4.05 3.25 3 3.60 2.97 0 35 35
## agefirstbirth 6 3312 22.63 4.86 22 22.18 4.45 9 42 33
## ethnicity* 7 9120 1.80 0.40 2 1.88 0.00 1 2 1
## city16* 8 9120 1.42 0.49 1 1.41 0.00 1 2 1
```

```
## lowincome16*      9 9120      1.21  0.41      1      1.14  0.00      1      2      1
## immigrant*       10 9120      1.11  0.31      1      1.01  0.00      1      2      1
##                  skew kurtosis  se
## kids             1.00      1.03 0.02
## age              0.48     -0.78 0.19
## education        -0.26      1.03 0.03
## year            -0.36     -1.16 0.10
## siblings         1.67      4.78 0.03
## agefirstbirth    0.87      0.59 0.08
## ethnicity*      -1.53      0.35 0.00
## city16*          0.30     -1.91 0.01
## lowincome16*     1.41     -0.02 0.00
## immigrant*       2.50      4.26 0.00
```

```
Hmisc::describe(gss)
```

```
## gss
##
## 10 Variables      9120 Observations
## -----
## kids
##      n missing distinct      Info      Mean      Gmd
##    9120         0         9    0.961     2.076     1.941
##
## Value      0      1      2      3      4      5      6      7      8
## Frequency  2127  1544  2338  1474   790   376   208   100   163
## Proportion 0.233 0.169 0.256 0.162 0.087 0.041 0.023 0.011 0.018
## -----
## age
##      n missing distinct      Info      Mean      Gmd      .05      .10
##    9120         0        72      1    46.08    20.38        22        25
##      .25      .50      .75      .90      .95
##      31      43      59      73      79
##
## lowest : 18 19 20 21 22, highest: 85 86 87 88 89
## -----
## education
##      n missing distinct      Info      Mean      Gmd      .05      .10
##    9120         0        21    0.957    12.64    3.178         8         9
##      .25      .50      .75      .90      .95
##      12      12      14      16      18
##
## lowest : 0 1 2 3 4, highest: 16 17 18 19 20
## -----
## year
##      n missing distinct      Info      Mean      Gmd
##    9120         0         8    0.979    1990    10.3
##
## Value      1974  1978  1982  1986  1990  1994  1998  2002
## Frequency   785   877  1064   842   767  1688  1580  1517
## Proportion 0.086 0.096 0.117 0.092 0.084 0.185 0.173 0.166
## -----
## siblings
##      n missing distinct      Info      Mean      Gmd      .05      .10
##    9120         0        27    0.984     4.051    3.359         1         1
```

```
##      .25      .50      .75      .90      .95
##        2        3        6        8       10
##
## lowest :  0  1  2  3  4, highest: 22 23 25 27 35
## -----
## agefirstbirth
##      n missing distinct      Info      Mean      Gmd      .05      .10
##    3312    5808      33    0.995    22.63    5.345      16      17
##      .25      .50      .75      .90      .95
##       19      22      25      30      32
##
## lowest :  9 11 12 13 14, highest: 38 39 40 41 42
## -----
## ethnicity
##      n missing distinct
##    9120      0      2
##
## Value      other  cauc
## Frequency  1785  7335
## Proportion 0.196 0.804
## -----
## city16
##      n missing distinct
##    9120      0      2
##
## Value      no  yes
## Frequency  5246 3874
## Proportion 0.575 0.425
## -----
## lowincome16
##      n missing distinct
##    9120      0      2
##
## Value      no  yes
## Frequency  7182 1938
## Proportion 0.788 0.212
## -----
## immigrant
##      n missing distinct
##    9120      0      2
##
## Value      no  yes
## Frequency  8122  998
## Proportion 0.891 0.109
## -----
```

```
# 6. Extract a data.frame with only people over the age of 80
```

```
gss_over80 <- gss[ gss$age > 80, ]
```

```
# 7. Get the mean number of children ("kids") for participants
```

```
#   over the age of 80
```

```
mean(gss[ gss$age > 80, "kids"])
```

```
## [1] 2.394737
```

```
# 8. Use the mean function to get the mean age at first birth.
# Hint: there is missing data.
mean(gss$agefirstbirth) # doesn't work because there is missing data

## [1] NA

mean(gss$agefirstbirth, na.rm = TRUE) # doesn't work because there is missing data

## [1] 22.63074
```

String functions

```
paste("hello", "how", "are", "You") # defaults to space separator

## [1] "hello how are You"

paste0("hello", "how", "are", "You") # no separator

## [1] "hellohowareYou"

paste("apple", "banana", "carrot", "date", sep = ", ") # specify arbitrary separator

## [1] "apple, banana, carrot, date"

paste0("v", 1:10) # paticularly useful with vectors

## [1] "v1" "v2" "v3" "v4" "v5" "v6" "v7" "v8" "v9" "v10"

# Extract substring
substr("abcdefghijklmnop", 4, 6)

## [1] "def"

# Change case
toupper("abcd") # make upper case

## [1] "ABCD"

tolower("ABCD") # make lower case

## [1] "abcd"

mystring <- c("apple", "banana", "carrot", "date", "egg", "fig")
# Identify which strings match a pattern
grep("a", mystring) # index of objects with "a"

## [1] 1 2 3 4

grep("a", mystring, value = TRUE) # value of objects with "a"

## [1] "apple" "banana" "carrot" "date"

# get count of number of characters
nchar(mystring)

## [1] 5 6 6 4 3 3

data.frame(mystring, nchar(mystring))

## mystring nchar.mystring.
```

```
## 1    apple      5
## 2    banana     6
## 3    carrot     6
## 4     date     4
## 5     egg      3
## 6     fig      3
```

```
# Substitute a mystringreplacement text that matches a pattern
questions <- c("How are you?", "What is going on?")
gsub(" ", "_", questions) # replace space with underscore
```

```
## [1] "How_are_you?"      "What_is_going_on?"
```

```
# R string manipulation tools are very powerful
# For more information see
?grep
?"regular expression"
```

```
# see also Hadley Wickham's package for string manipulation
# It attempts to introduce greater consistency in notation.
# install.packages("stringr")
library(stringr)
```

```
## Warning: package 'stringr' was built under R version 3.5.2
```

```
help(package = "stringr")
# all functions begin with str_
str_length(mystring) # see nchar
```

```
## [1] 5 6 6 4 3 3
```

```
str_sub(mystring, start = 1, end = 3)
```

```
## [1] "app" "ban" "car" "dat" "egg" "fig"
```

```
# writing output to the console
cat("Hello World!")
```

```
## Hello World!
```

```
# Tab is \t and new line is \n
cat("Hello\t World\nSome more text")
```

```
## Hello      World
## Some more text
```

Importing data

```
# A simple option is to export data from your external data
# in csv format and then import the data using csv
# csv
medals <- read.csv("data/practice/medals.csv")
head(medals)
```

```
##   Year   City      Sport  Discipline NOC      Event Event.gender
## 1 1924 Chamonix  Skating Figure skating AUT  individual      M
```

```
## 2 1924 Chamonix    Skating Figure skating AUT      individual      W
## 3 1924 Chamonix    Skating Figure skating AUT      pairs          X
## 4 1924 Chamonix    Bobsleigh      Bobsleigh BEL      four-man        M
## 5 1924 Chamonix    Ice Hockey      Ice Hockey CAN      ice hockey       M
## 6 1924 Chamonix    Biathlon      Biathlon FIN      military patrol  M
## Medal
## 1 Silver
## 2 Gold
## 3 Gold
## 4 Bronze
## 5 Gold
## 6 Silver
```

```
tail(medals)
```

```
##      Year  City Sport Discipline NOC      Event Event.gender Medal
## 2306 2006 Turin Skiing  Snowboard USA      Half-pipe      M Gold
## 2307 2006 Turin Skiing  Snowboard USA      Half-pipe      M Silver
## 2308 2006 Turin Skiing  Snowboard USA      Half-pipe      W Gold
## 2309 2006 Turin Skiing  Snowboard USA      Half-pipe      W Silver
## 2310 2006 Turin Skiing  Snowboard USA Snowboard Cross      M Gold
## 2311 2006 Turin Skiing  Snowboard USA Snowboard Cross      W Silver
```

```
dim(medals)
```

```
## [1] 2311      8
```

```
# Other delimited formats
```

```
medals <- read.table("data/practice/medals.tsv", sep = "\t")
```

```
# Read Excel
```

```
library(readxl)
```

```
## Warning: package 'readxl' was built under R version 3.5.2
```

```
# note that readxl returns a tibble,
```

```
# which is almost but not quite the same as a data.frame
```

```
medals <- data.frame(readxl::read_excel("data/practice/medals.xls"))
```

```
head(medals)
```

```
##      Year  City Sport Discipline NOC      Event Event.gender
## 1 1924 Chamonix    Skating Figure skating AUT      individual      M
## 2 1924 Chamonix    Skating Figure skating AUT      individual      W
## 3 1924 Chamonix    Skating Figure skating AUT      pairs          X
## 4 1924 Chamonix    Bobsleigh      Bobsleigh BEL      four-man        M
## 5 1924 Chamonix    Ice Hockey      Ice Hockey CAN      ice hockey       M
## 6 1924 Chamonix    Biathlon      Biathlon FIN      military patrol  M
## Medal
## 1 Silver
## 2 Gold
## 3 Gold
## 4 Bronze
## 5 Gold
## 6 Silver
```

```
# SPSS
```

```
library(foreign)
```

```
cas <- foreign::read.spss("data/practice/cas.sav", to.data.frame = TRUE)
```

```
attr(cas, "variable.labels")
```

```
##                                district
##                                "District code"
##                                school
##                                "School name"
##                                county
##                                "County"
##                                grades
##                                "grade span of district"
##                                students
##                                "Total enrollment"
##                                teachers
##                                "Number of teachers"
##                                calworks
## "Percent qualifying for CalWorks (income assistance)"
##                                lunch
## "Percent qualifying for reduced-price lunch"
##                                computer
##                                "Number of computers"
##                                expenditure
##                                "Expenditure per student"
##                                income
##                                "District average income (in USD 1,000)"
##                                english
##                                "Percent of English learners"
##                                read
##                                "Average reading score"
##                                math
##                                "Average math score"
```

```
# tip: You may need to think about value labels in your SPSS file
# Specifically, if you have numeric variables that have variable labels, you may
# want to remove the value labels in SPSS or
```

```
# import stata, sas
?read.dta
?read.sas
```

```
## No documentation for 'read.sas' in specified packages and libraries:
## you could try '??read.sas'
```

```
# Use ProjectTemplate to auto-import (see discussion later)
```

Exporting data

```
mydata <- data.frame(a = c(1,2,3), b = c("a", "b", "c"))
```

```
# Internal R format
# Good option if you need to re-open data in R
save(mydata, file="output/mydata.rdata")
# load("output/mydata.rdata")
```

```

# csv
# Good option if you need to get data into other software
# This should open in almost all other software (e.g. Excel, SPSS, etc.)
write.csv(mydata, file = "output/mydata.csv")
write.csv(mydata, file = "output/mydata-2.csv", row.names = FALSE) # exclude row.names

# If you need more flexibility in terms of delimiters, etc.
write.table(mydata, file = "output/mydata.tsv", sep = "\t") # e.g., tab delimiter

# write excel
library(openxlsx)

## Warning: package 'openxlsx' was built under R version 3.5.2
openxlsx::write.xlsx(mydata, file = "output/mydata.xlsx")

## Note: zip::zip() is deprecated, please use zip::zipr() instead

# Exporting to other formats
# There are a range of options for exporting to other formats
# Functionality is often spread around
# Given that the csv option is usually sufficient
library(foreign)
?foreign::write.foreign # options for exporting to SAS, SPSS, and Stata directly

```

Exercise 3

```

# 1. Open medals.csv in the data/practice/ directory
#    and assign to variable medals

# 2. Check that the file imported correctly
#    (a) look at the first few rows,
#    (b) look at the last few rows,
#    (c) check the structure (i.e., str),
#    (c) Use the Hmisc describe function to check basic properties

# 3. Create a new variable in medals that indicates
#    whether the medals was Gold (TRUE) or Silver/Bronze (FALSE)
#    and call it isgold

# 4. Calculate the total number of gold medals

# 5. Export the medals data.frame to the output folder
#    (a) as a csv file
#    (b) as a native rdata file

# 6. Remove the medals dataset from the workspace
#    and then load it again from the csv file.
#    Check that it imported correctly.
#    Then remove medals and repeat for the rdata file

```


Answers for Exercise 3

```
# 1. Open medals.csv in the data/practice/ directory
# and assign to variable medals
medals <- read.csv("data/practice/medals.csv")
```

```
# 2. Check that the file imported correctly
# (a) look at the first few rows,
# (b) look at the last few rows,
# (b) check the structure (i.e., str),
# (c) Use the Hmisc describe function to check basic properties
head(medals)
```

```
##   Year   City   Sport   Discipline NOC      Event Event.gender
## 1 1924 Chamonix   Skating Figure skating AUT   individual      M
## 2 1924 Chamonix   Skating Figure skating AUT   individual      W
## 3 1924 Chamonix   Skating Figure skating AUT     pairs      X
## 4 1924 Chamonix Bobsleigh   Bobsleigh BEL   four-man      M
## 5 1924 Chamonix Ice Hockey   Ice Hockey CAN   ice hockey      M
## 6 1924 Chamonix Biathlon   Biathlon FIN military patrol      M
##   Medal
## 1 Silver
## 2  Gold
## 3  Gold
## 4 Bronze
## 5  Gold
## 6 Silver
```

```
tail(medals)
```

```
##      Year City Sport Discipline NOC      Event Event.gender Medal
## 2306 2006 Turin Skiing  Snowboard USA   Half-pipe      M   Gold
## 2307 2006 Turin Skiing  Snowboard USA   Half-pipe      M Silver
## 2308 2006 Turin Skiing  Snowboard USA   Half-pipe      W   Gold
## 2309 2006 Turin Skiing  Snowboard USA   Half-pipe      W Silver
## 2310 2006 Turin Skiing  Snowboard USA Snowboard Cross      M   Gold
## 2311 2006 Turin Skiing  Snowboard USA Snowboard Cross      W Silver
```

```
str(medals)
```

```
## 'data.frame':   2311 obs. of  8 variables:
## $ Year      : int  1924 1924 1924 1924 1924 1924 1924 1924 1924 1924 ...
## $ City      : chr  "Chamonix" "Chamonix" "Chamonix" "Chamonix" ...
## $ Sport     : chr  "Skating" "Skating" "Skating" "Bobsleigh" ...
## $ Discipline : chr  "Figure skating" "Figure skating" "Figure skating" "Bobsleigh" ...
## $ NOC       : chr  "AUT" "AUT" "AUT" "BEL" ...
## $ Event     : chr  "individual" "individual" "pairs" "four-man" ...
## $ Event.gender: chr  "M" "W" "X" "M" ...
## $ Medal     : chr  "Silver" "Gold" "Gold" "Bronze" ...
```

```
Hmisc::describe(medals)
```

```
## medals
##
## 8 Variables      2311 Observations
## -----
```

```

## Year
##      n missing distinct      Info      Mean      Gmd      .05      .10
##    2311      0      20    0.995    1980    24.32    1932    1948
##      .25      .50      .75      .90      .95
##    1968    1988    1998    2006    2006
##
## Value      1924  1928  1932  1936  1948  1952  1956  1960  1964  1968
## Frequency      49   41   42   51   68   67   72   81  103  106
## Proportion 0.021 0.018 0.018 0.022 0.029 0.029 0.031 0.035 0.045 0.046
##
## Value      1972  1976  1980  1984  1988  1992  1994  1998  2002  2006
## Frequency     105   111   115   117   138   171   183   205   234   252
## Proportion 0.045 0.048 0.050 0.051 0.060 0.074 0.079 0.089 0.101 0.109
## -----
## City
##      n missing distinct
##    2311      0      17
##
## Albertville (171, 0.074), Calgary (138, 0.060), Chamonix (49, 0.021),
## Cortina d'Ampezzo (72, 0.031), Garmisch-Partenkirchen (51, 0.022),
## Grenoble (106, 0.046), Innsbruck (214, 0.093), Lake Placid (157, 0.068),
## Lillehammer (183, 0.079), Nagano (205, 0.089), Oslo (67, 0.029), Salt Lake
## City (234, 0.101), Sapporo (105, 0.045), Sarajevo (117, 0.051), Squaw
## Valley (81, 0.035), St. Moritz (109, 0.047), Turin (252, 0.109)
## -----
## Sport
##      n missing distinct
##    2311      0      7
##
## Value      Biathlon  Bobsleigh      Curling Ice Hockey      Luge
## Frequency      162      133          21          69      108
## Proportion      0.070      0.058      0.009      0.030      0.047
##
## Value      Skating      Skiing
## Frequency      758      1060
## Proportion      0.328      0.459
## -----
## Discipline
##      n missing distinct
##    2311      0      15
##
## Alpine Skiing (367, 0.159), Biathlon (162, 0.070), Bobsleigh (115, 0.050),
## Cross Country S (399, 0.173), Curling (21, 0.009), Figure skating (207,
## 0.090), Freestyle Ski. (54, 0.023), Ice Hockey (69, 0.030), Luge (108,
## 0.047), Nordic Combined (84, 0.036), Short Track S. (96, 0.042), Skeleton
## (18, 0.008), Ski Jumping (114, 0.049), Snowboard (42, 0.018), Speed
## skating (455, 0.197)
## -----
## NOC
##      n missing distinct
##    2311      0      45
##
## lowest : AUS AUT BEL BLR BUL, highest: UKR URS USA UZB YUG
## -----

```

```
## Event
##      n missing distinct
##    2311      0      67
##
## lowest : 10000m      1000m      10km      10km pursuit      12,5km mass start
## highest: super-G      Team      Team pursuit      Team sprint      two-man
## -----
## Event.gender
##      n missing distinct
##    2311      0      3
##
## Value      M      W      X
## Frequency  1386   802   123
## Proportion 0.600 0.347 0.053
## -----
## Medal
##      n missing distinct
##    2311      0      3
##
## Value      Bronze   Gold Silver
## Frequency    764    774   773
## Proportion  0.331  0.335  0.334
## -----
```

```
# 3. Create a new variable in medals that indicates
#   whether the medals was Gold (TRUE) or Silver/Bronze (FALSE)
#   and call it isgold
medals$isgold <- medals$Medal == "Gold"
```

```
# 4. Calculate the total number of gold medals
sum(medals$isgold)
```

```
## [1] 774
```

```
# 5. Export the medals data.frame to the output folder
#   (a) as a csv file
#   (b) as a native rdata file
write.csv(medals, "output/medals.csv")
# or technically you may want to do
write.csv(medals, "output/medals.csv", row.names = FALSE)
save(medals, file = "output/medals.rdata")
```

```
# 6. Remove the medals dataset from the workspace
#   and then load it again from the csv file.
#   Check that it imported correctly.
# Then remove medals and repeat for the rdata file
rm(medals)
medals <- read.csv("output/medals.csv")
head(medals)
```

##	Year	City	Sport	Discipline	NOC	Event	Event.gender
## 1	1924	Chamonix	Skating	Figure skating	AUT	individual	M
## 2	1924	Chamonix	Skating	Figure skating	AUT	individual	W
## 3	1924	Chamonix	Skating	Figure skating	AUT	pairs	X
## 4	1924	Chamonix	Bobsleigh	Bobsleigh	BEL	four-man	M
## 5	1924	Chamonix	Ice Hockey	Ice Hockey	CAN	ice hockey	M

```
## 6 1924 Chamonix    Biathlon    Biathlon FIN military patrol    M
##      Medal isgold
## 1 Silver  FALSE
## 2   Gold   TRUE
## 3   Gold   TRUE
## 4 Bronze  FALSE
## 5   Gold   TRUE
## 6 Silver  FALSE

rm(medals)
load("output/medals.rdata")
head(medals)
```

```
##   Year      City      Sport      Discipline NOC      Event Event.gender
## 1 1924 Chamonix    Skating Figure skating AUT      individual      M
## 2 1924 Chamonix    Skating Figure skating AUT      individual      W
## 3 1924 Chamonix    Skating Figure skating AUT      pairs        X
## 4 1924 Chamonix  Bobsleigh    Bobsleigh BEL      four-man      M
## 5 1924 Chamonix Ice Hockey    Ice Hockey CAN      ice hockey      M
## 6 1924 Chamonix    Biathlon    Biathlon FIN military patrol      M
##      Medal isgold
## 1 Silver  FALSE
## 2   Gold   TRUE
## 3   Gold   TRUE
## 4 Bronze  FALSE
## 5   Gold   TRUE
## 6 Silver  FALSE
```

Random variables and distributions

```
# In statistics, we often want to generate random data with certain properties
# or looking up features of statistical distributions.
# See the following help for list of common distributions in base R
?Distributions

# and see http://cran.r-project.org/web/views/Distributions.html for many more distributions

# Each distribution has four functions that differ in terms of the first letter
# For example, for the normal distribution, you have
dnorm(1) # Density of the value 1 of a standard normal distribution

## [1] 0.2419707

pnorm(1) # Cumulative distribution function for value of 1 on standard normal distribution

## [1] 0.8413447

qnorm(.975) # Inverse cumulative distribution function for value of .975

## [1] 1.959964

rnorm(5) # Generate 5 random draws from normal distribution

## [1] -1.428656760  0.008139295 -0.203252620 -2.308346829 -1.063667774
```

```

#
dunif(1) # Density of the value 1 of a uniform distribution (0, 1)

## [1] 1
punif(.5) # Cumulative distribution function for value of 1 on uniform distribution

## [1] 0.5
qunif(.975) # Inverse cumulative distribution function for value of .975

## [1] 0.975
runif(5) # Generate 5 random draws from uniform distribution

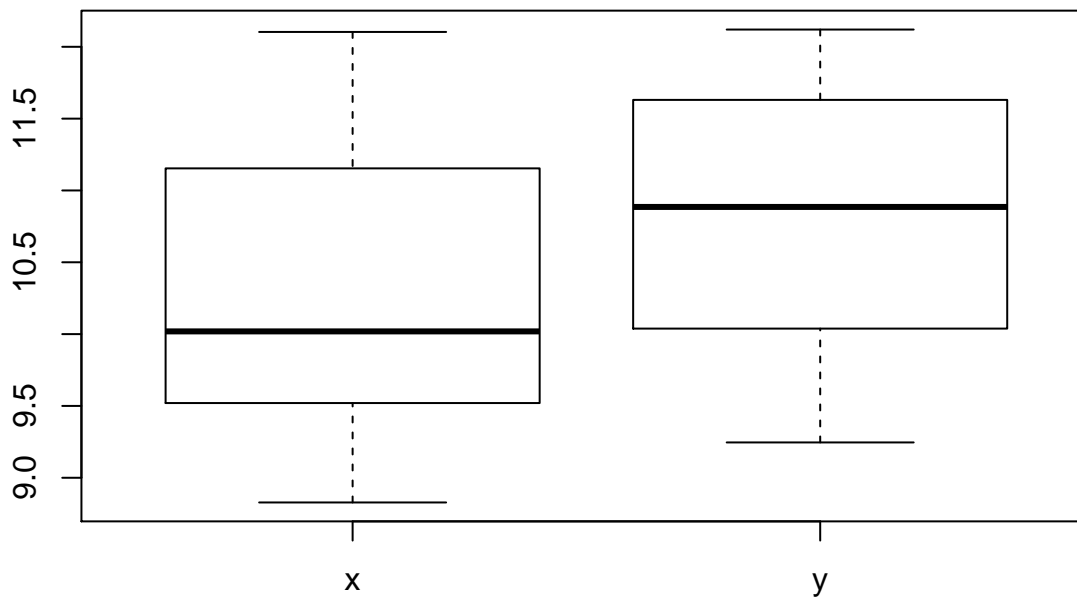
## [1] 0.7419174 0.0461430 0.7963427 0.3642507 0.2666787

# Distributions have parameters that can be specified
x <- rnorm(10, mean = 10, sd = 1) # draw 10 from mean of 10
y <- rnorm(10, mean = 11, sd = 1) # draw 10 from mean of 11
dat <- data.frame(x=x, y=y)
dat

##           x           y
## 1  10.336496 11.020999
## 2   9.200158  9.497520
## 3   9.530175 11.630426
## 4   8.828091 11.267821
## 5  12.103554 12.120114
## 6   9.701318 10.749551
## 7   9.520142  9.245939
## 8  10.590699 10.725930
## 9  11.153680 11.937980
## 10 11.751954 10.038433

boxplot(dat)

```



Functions

```
# You can write functions and these are generally the same as
# the functions you use in R

# For example, I could create a function that printed some text
print_some_text <- function(x = "Hello World") {
  print(x)
}
```

```
# If I run the above command, I can then use it
print_some_text() # using the default argument
```

```
## [1] "Hello World"
```

```
print_some_text("blah blah blah") # or to print some other text
```

```
## [1] "blah blah blah"
```

```
# Anatomy of a function
# Functions have a name
# They take one or more arguments
# Arguments may have default values
```

```
# Let's take a more interesting example: Power analysis
# The following data simulates data for two groups and
# examines whether there is a significant difference at .05
# It repeats the process 1000 times and calculates the
# proportion of times it is statistically significant
# (i.e., simulation estimate of the statistical power)
```

```
significant <- NULL
for (i in 1:1000) {
  x <- rnorm(30, mean = 0.0, sd = 1)
  y <- rnorm(30, mean = 0.3, sd = 1)
  fit <- t.test(x, y)
  fit
  significant[i] <- (fit$p.value < .05)
}
statistical_power <- mean(significant)
statistical_power
```

```
## [1] 0.203
```

```
# we could convert this to a function
power_group_dif1 <- function() {
  significant <- NULL
  for (i in 1:1000) {
    x <- rnorm(30, mean = 0.0, sd = 1)
    y <- rnorm(30, mean = 0.3, sd = 1)
    fit <- t.test(x, y)
    fit
    significant[i] <- (fit$p.value < .05)
  }
}
```

```

    statistical_power <- mean(significant)
    statistical_power
}

power_group_dif1()

## [1] 0.186

# but the beauty of function is that they can make things general
# Let's make the mean of group 2 an argument that can be specified
power_group_dif2 <- function(mean2 = 0.3) {
  significant <- NULL
  for (i in 1:1000) {
    x <- rnorm(30, mean = 0.0, sd = 1)
    y <- rnorm(30, mean = mean2, sd = 1)
    fit <- t.test(x, y)
    fit
    significant[i] <- (fit$p.value < .05)
  }
  statistical_power <- mean(significant)
  statistical_power
}

# now we can specify different values
power_group_dif2(0)

## [1] 0.052

power_group_dif2(.3)

## [1] 0.203

power_group_dif2(.5)

## [1] 0.475

power_group_dif2(.8)

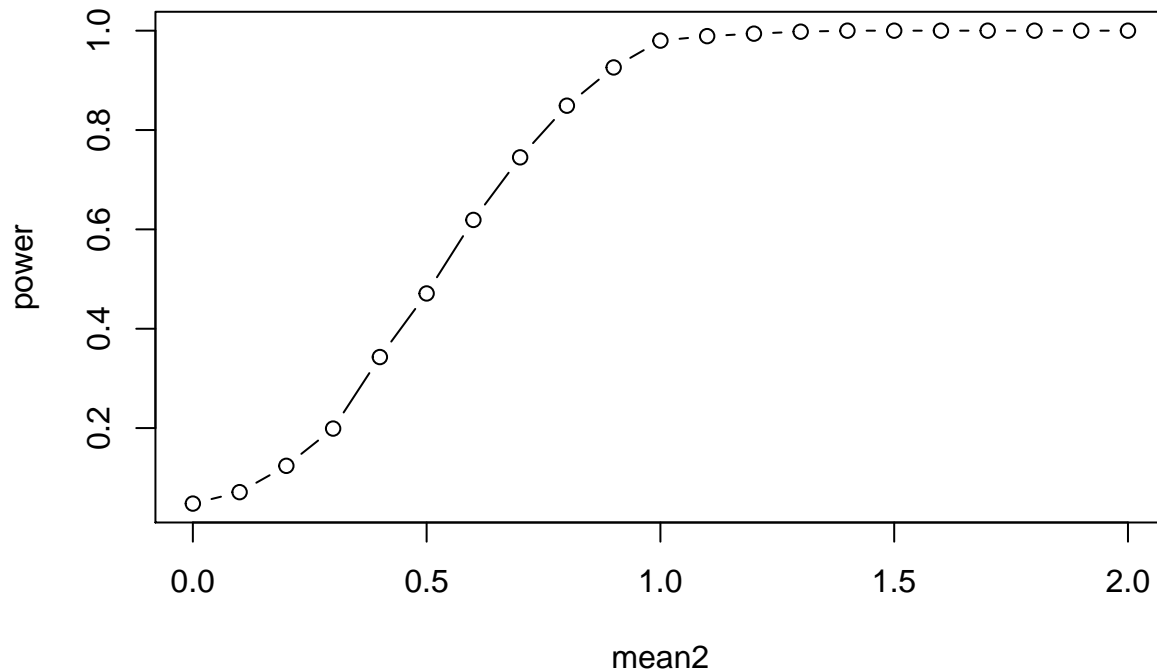
## [1] 0.846

power_group_dif2(1)

## [1] 0.97

settings <- seq(from = 0, to = 2, by = .1)
results <- data.frame(mean2= settings)
results$power <- sapply(results$mean2, function(X) power_group_dif2(X))
plot(results, type = "b")

```



```
# obviously it could be made a whole lot more general
power_group_dif3 <- function(mean1 = 0, mean2 = 0.3, sd1 = 1, sd2 = 1,
                             n1 = 30 , n2 = 30, ksimulations = 1000,
                             alpha_criterion = .05) {
  significant <- NULL
  for (i in 1:ksimulations) {
    x <- rnorm(30, mean = mean1, sd = sd1)
    y <- rnorm(30, mean = mean2, sd = sd2)
    fit <- t.test(x, y)
    fit
    significant[i] <- (fit$p.value < alpha_criterion)
  }
  statistical_power <- mean(significant)
  statistical_power
}

power_group_dif3(mean1 = 10, mean2 = 11, sd1 = 1, sd2 = 1,
                  n1 = 100 , n2 = 100, ksimulations = 10000,
                  alpha_criterion = .01)
```

```
## [1] 0.8867
```

Debugging functions

```
# debugging functions
print_some_text <- function(x = "Hello World") {
  print(x)
}
debugonce(print_some_text) # activates debugging on the function
print_some_text()
```



```
## debugging in: print_some_text()
## debug at <text>#2: {
##   print(x)
## }
## debug at <text>#3: print(x)
## [1] "Hello World"
## exiting from: print_some_text()

# many other useful functions
?traceback # provide further information when an error occurs
?browser # place in function
```

Viewing source code for internal functions

```
# Option 1: type function name
t.test
cor
power.t.test

# Option 2:
# S3 Methods
# Some functions are generic and operate differently depending
# on the class of the first argument
# mean
# print
# summary

# Methods will list the actual function names called
methods(mean)
methods(print)
methods(summary)

mean.default
summary.table

# Option 3:
# Some functions are part of packages but are not exported
# I.e., they are intended for internal use, but
# they are often quite useful
library(ProjectTemplate)

## Warning: package 'ProjectTemplate' was built under R version 3.5.2

# Double colon shows the functions exported from a package
# i.e., packagename::function
ProjectTemplate::run.project

# Triple colon shows internal functions
# i.e., packagename:::function
ProjectTemplate:::xls.reader
```

```
# Also, see the getAnywhere function  
xls.reader # this doesn't work
```

```
## Error in eval(expr, envir, enclos): object 'xls.reader' not found
```

```
getAnywhere(xls.reader) # this does work
```

Exercise 4

```
library(MASS)  
data(mammals)  
?mammals  
head(mammals)
```

```
##           body brain  
## Arctic fox    3.385 44.5  
## Owl monkey   0.480 15.5  
## Mountain beaver 1.350 8.1  
## Cow          465.000 423.0  
## Grey wolf    36.330 119.5  
## Goat         27.660 115.0
```

```
# 1. Create a function that takes a single argument x  
#     and prints that value twice.  
#     use the function to print "hello world" twice
```

```
# 2. Divide mammall brain weight (g) by body weight (kg) and  
#     get the mean of this value
```

```
# 3. Write a function that takes arguments x and y  
#     and returns the mean of x divided by y
```

```
# 4. Apply the function to get the mean ratio of brain to body size
```

```
# 5. Modify the ratio function to return a list with  
#     (a) the mean of x divided by y, and  
#     (b) the sd of x divided by y.  
#     Then apply to mammals data as above.
```

```
# 6. Step through the code for the correlation function
```

```
# 7. Show the source code for  
#     (a) the t.test function,  
#     (b) the summary method for lm objects  
#     (c) the alpha function in the psych package
```

Answers 4

```
library(MASS)
data(mammals)
?mammals
head(mammals)
```

```
##              body brain
## Arctic fox    3.385  44.5
## Owl monkey    0.480  15.5
## Mountain beaver 1.350   8.1
## Cow           465.000 423.0
## Grey wolf     36.330 119.5
## Goat          27.660 115.0
```

```
# 1. Create a function that takes a single argument x
# and prints that value twice.
# use the function to print "hello world" twice
```

```
print_twice <- function(x) {
  print(x)
  print(x)
}
print_twice("hello world")
```

```
## [1] "hello world"
## [1] "hello world"
```

```
# 2. Divide mammal brain weight (g) by body weight (kg) and
# get the mean of this value
mean(mammals$brain / mammals$body )
```

```
## [1] 9.624214
```

```
# 3. Write a function that takes arguments x and y
# and returns the mean of x divided by y
```

```
mean_ratio <- function(x, y) {
  mean(x / y)
}
```

```
# 4. Apply the function to get the mean ratio of brain to body size
mean_ratio(mammals$brain, mammals$body)
```

```
## [1] 9.624214
```

```
# 5. Modify the ratio function to return a list with
# (a) the mean of x divided by y, and
# (b) the sd of x divided by y.
# Then apply to mammals data as above.
```

```
mean_ratio <- function(x, y) {
  ratioxy <- x / y
  list(mean_ratio = mean(ratioxy),
       sd_ratio = sd(ratioxy))
}
```

```
# 6. Step through the code for the correlation function
# debugonce(cor)
```

```
cor(mammals$brain, mammals$body, method = "spearman")
```

```
## [1] 0.9534986
```

```
# 7. Show the source code for  
# (a) the t.test function,  
# (b) the summary method for lm objects  
# (c) the alpha function in the psych package  
# t.test  
# summary.lm  
# psych::alpha
```