R crash course - Introduction

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R Basics

Vectors

Matrices

Vector/Matrix Operations

Functions

Exercise

Packages

Dependencies

- Please download and install
 - ► Latest version R (free – GPL – from https://www.r-project.org/)
 - ► Latest version of Rstudio (also *free* AGPL from https://www.rstudio.com/)

A Question

You are given three sticks, each of a random length between 0 and 1.

What's the probability you can make a triangle?

- ▶ The answer is 1/2
- By the end of this session, we'll confirm this with a simulation in R

Rstudio Basics

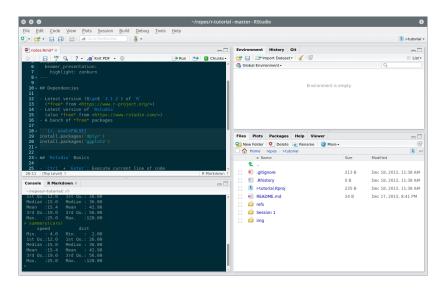


Figure 1: Screenshot of Rstudio

Rstudio Basics

- ▶ Save/load scripts as text files named *.R
- ► Save/load environment as .Rdata file
- ▶ Ctrl + L: Clear console
- ▶ Ctrl + #: Focus on panel #
- ► Ctrl + Enter: Execute line/selection
- ▶ Ctrl + Shift + A: Auto-format selected code
- ► Ctrl + Shift + C: Comment/Uncomment selection
- ▶ Ctrl + Shift + /: Reflow comments
- Many more (if you're willing to explore)

R Basics

R Basics: Working Directory

- Working directory (wd) is where your R session will load/save files
- ► To see where your current working directory is, run

getwd()

▶ To set the working directory to desired path, run

```
setwd("path")
```

R Basics: Working Directory (cont'd)

- Note that ~ is replaced with your HOME directory, e.g.
 C:\Users\Username\Documents in windows
- ▶ Use forward slashes (/), even on Windows!

setwd("~/R")

- Never set the working directory to an absolute path in your R script
 - Even better, organize your scripts such that you never have to change the working directory
 - setwd() is best reserved for that time when you started R from the wrong directory

R Basics: Math Operations

Simple math operations

```
3+11 # add stuff
3-11 # subtract stuff
3/11 # divide stuff
3*11 # multiply stuff
2^10 # raise to powers
```

R Basics: Assignments

- Convention for assigning values to variables is an arrow(<-)¹
- Direction of arrow indicates direction of assignment

```
A <- 12

A # 12

A + 3 -> B

B # 15

24 -> A

A # 24
```

► The equal sign (=) also works, but only for assignment to the left, e.g.

```
A = 12  # good
12 = A  # BAD
```

¹There are also double arrows (<<-) that do something slightly different. We won't talk about it, but if you're interested, see here

R Basics: Strings

➤ A String variable can be declared in either double quotes("") or single quotes ('')

```
s <- "This is a valid string"

## [1] "This is a valid string"

s <- 'and so is this'
s

## [1] "and so is this"</pre>
```

R Basics: Re-Assignments

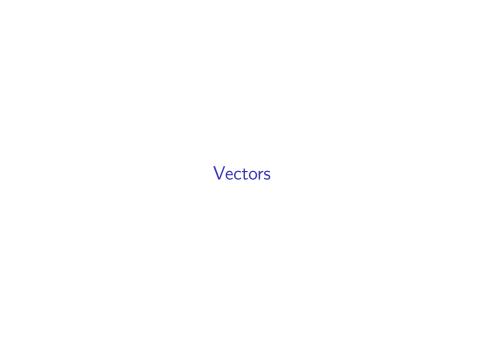
A variable can be re-assigned to anything

```
x <- 860306 # first x is assigned a number
x

## [1] 860306

x <- 'This is a variable!'
x # Now it is a string</pre>
```

```
## [1] "This is a variable!"
```



Vectors: c()

- ▶ Vectors are the building blocks of R even a single variable is actually an "atomic" vector (vector of size 1)
- ▶ Vectors in R are created by concatenating a series of elements

```
X <- c(1,2,3)
X # vector of numbers (1, 2, 3)</pre>
```

```
## [1] 1 2 3
```

```
Y <- c('this', 'that', 'those')
Y # this is a vector of Strings
```

```
## [1] "this" "that" "those"
```

Vectors: seq()

► Create a vector from a sequence with seq(from, to, by=1)

```
seq(1, 10)
```

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

```
seq(1, 10, 2)
```

```
## [1] 1 3 5 7 9
```

▶ Use short-hand from: to if you're incrementing by one

1:10

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

Vectors: rep()

▶ Use rep() to repeat values

[1] "Sat." "Sun." "Sat." "Sun."

```
rep(13, 4)
## [1] 13 13 13 13
rep('Yes!', 3)
## [1] "Yes!" "Yes!" "Yes!"
rep(c('Sat.', 'Sun.'), 2)
```

Vectors: rdist()

Generate vector of n samples from a specified distribution

```
runif(n = 10) # 10 samples from Unif(0, 1)
rnorm(n = 10) # 10 samples from Norm(0, 1)
rpois(n = 10) # 10 samples from Poisson(1)
rexp(n = 10) # 10 samples from Exp(1)
```

Distribution parameters can be specified as arguments, e.g.

```
# 100 samples from a Norm(20, 5) distribution
rnorm(n = 100, mean = 20, sd = 5)
```

Read documentation for available distributions

?Distributions

Vectors: Indexing

▶ Use square braces ([]) to index a vector (base 1)

```
X \leftarrow c(10, 11, 12, 13)
X[1]
## [1] 10
X[4]
## [1] 13
X[5]
## [1] NA
```

Vectors: Indexing (cont'd)

Negative indexing is used to exclude elements

X[-1]

```
## [1] 11 12 13
```

Index multiple objects by indexing with a vector

```
ind <- c(2, 4)
X[ind]</pre>
```

```
## [1] 11 13
```

Vectors: Re-assignment with Indices

▶ Replace elements by re-assigning with index

```
X[1] <- 101
X
```

```
## [1] 101 11 12 13
```

Replace multiple elements as well

```
X[2:3] <- c(22, 33)
X
```

```
## [1] 101 22 33 13
```

Vectors: Add Elements by Index

Add new elements to a vector by assigning

```
X[5]
## [1] NA

X[5] <- 555
X
```

```
## [1] 101 22 33 13 555
```

Vectors: Advanced Indexing

- Vectors can be indexed by a binary vector (TRUE/FALSE) of equal length
- ▶ i.e., you can index vectors by a specified condition, e.g.,

```
X <- 1:100
# create a binary vector with the same length of X
# where the element is TRUE if the element of X
# in the corresponding position satisfies condition
ind <- X > 95
tail(ind) # take a peek at the last few entries
```

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

```
X[ind]
```

```
## [1] 96 97 98 99 100
```

Matrices

Creating Matrices

A matrix is created from a vector, using matrix(), e.g.

```
X <- c(1:12)
# syntax: matrix(vector, # of rows, # of columns)
A <- matrix(X, 3, 4)
A</pre>
```

```
## [,1] [,2] [,3] [,4]
## [1,] 1 4 7 10
## [2,] 2 5 8 11
## [3,] 3 6 9 12
```

Notice that the matrix is created column-first

Matrix Indexing

- ► Similar to vectors, a matrix can be indexed with square braces, with the syntax [row #, col #], e.g.
- ► Leaving an entry empty will result in the full row/column

```
A[3,2]
## [1] 6
A[2,]
        # entire second row
## [1] 2 5 8 11
A[,4]
        # entire fourth column
   [1] 10 11 12
```

Vector/Matrix Operations

Vector Operations

```
X = c(1:4)
t(X) # transpose (column) vector X to row vector

## [,1] [,2] [,3] [,4]
## [1,] 1 2 3 4
```

X + X # element-wise summation

```
## [1] 2 4 6 8
```

X - X # element-wise subtraction

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

Vector Operations (cont'd)

```
X^3
## [1] 1 8 27 64
X * X # element-wise multiplication
## [1] 1 4 9 16
X %*% X # dot (inner) product
## [,1]
## [1.]
         30
```

Matrix Operations

```
A = matrix(1:4, 2, 2) # create 2x2 matrix
t(A) # transpose (column) vector A to row vector
## [,1] [,2]
## [1,] 1 2
## [2,] 3 4
A + A # element-wise summation
## [,1] [,2]
## [1,] 2 6
## [2,] 4 8
A - A # element-wise subtraction
## [,1] [,2]
## [1,] 0
## [2,]
```

Matrix Operations (cont'd)

```
A^3 # element-wise exponentiation
```

```
## [,1] [,2]
## [1,] 1 27
## [2,] 8 64
```

A * A # element-wise multiplication

```
## [,1] [,2]
## [1,] 1 9
## [2,] 4 16
```

A %*% A # dot (inner) product

```
## [,1] [,2]
## [1,] 7 15
## [2,] 10 22
```

Matrix Operations: Warning

Dimensions must make sense!

A %*% A # 2x3 times 2x3: Nope

```
A <- matrix(1:6, 2, 3)
B <- matrix(1:6, 3, 2)
A %*% B # 2x3 times 3x2: OK

## [,1] [,2]
## [1,] 22 49
## [2,] 28 64
```

```
## Error in A %*% A: non-conformable arguments
```

Vector/Matrix Comparisons

Comparisons are all done element-wise

$$c(1, 2, 3) == c(1, 2, 4)$$

[1] TRUE TRUE FALSE

```
c(1, 2, 3) < c(1, 2, 4)
```

[1] FALSE FALSE TRUE

$$c(1, 2, 3) >= c(1, 2, 4)$$

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

► Note the double equal sign for comparing equality (one would be assignment!)

Helpful Vector Functions

▶ If possible, avoid loops by operating over the Vector/Matrix as a whole

```
mean(X)
                        # mean
sd(X)
                        # standard deviation
var(X)
                        # variance
max(X)
                        # maximum
min(X)
                        # minimum
median(X)
                        # median
sum(X)
                        # sum
prod(X)
                        # product
quantile(X,probs=0.5)
                        # quantile for specified probs
length(X)
                        # length of the vector
range(X)
```

Helpful Matrix Functions

▶ If possible, avoid loops by operating over the Vector/Matrix as a whole

```
rowSums(A) # Row sums
colSums(A) # Column sums
rowMeans(A) # Row means
colMeans(A) # Columns means
diag(A) # Diagonal of a matrix
solve(A) # Inverse of a matrix
cov(A) # Variance covariance matrix
cor(A) # Correlation matrix
```

Functions

Some more built-in functions

We've already seen many built-in functions, but here are some more!

```
log(X) # element-wise log
exp(X) # element-wise exponential
sqrt(X) # element-wise square root
```

Functions for Strings

```
# concatenate two (or more) strings
paste('one plus one equals', 1+1, '!')
## [1] "one plus one equals 2!"
paste('one plus one', 1+1, sep='=')
## [1] "one plus one=2"
# if you're into C-style formatting ...
sprintf('one plus one = %d', 1+1)
## [1] "one plus one = 2"
```

Functions for Strings (cont'd)

▶ Often, we want to concatenate strings with no spaces (e.g., when constructing filenames/paths at run-time)

```
# short-hand for concatenation w/o spaces
filename = 'some_file_name.csv'
pasteO('path/to/', filename)
```

```
## [1] "path/to/some_file_name.csv"
```

```
# function specifically for constructing file paths
file.path('path', 'to', filename)
```

```
## [1] "path/to/some_file_name.csv"
```

Functions for Strings (cont'd)

► To enforce upper/lower cases

```
s <- 'SoMe CraZY STRING'
tolower(s)</pre>
```

```
## [1] "some crazy string"
```

```
toupper(s)
```

```
## [1] "SOME CRAZY STRING"
```

Generic Functions

Some functions for exploring objects

```
obj <- 1:100
head(obj, n=5)  # display first n rows of obj

## [1] 1 2 3 4 5

tail(obj, n=5)  # display last n rows of obj

## [1] 96 97 98 99 100</pre>
```

Generic Functions (cont'd)

1.00

##

```
str(obj) # display structure of obj

## int [1:100] 1 2 3 4 5 6 7 8 9 10 ...

summary(obj) # display summary of obj

## Min. 1st Qu. Median Mean 3rd Qu. Max.
```

50.50

75.25

100.00

25.75 50.50

Control Flow

▶ if statements

```
if (condition) {
    # stuff to do when condition is TRUE
} else if(other_condition) { # (OPTIONAL)
    # stuff to do if other_condition is TRUE
} else { # (OPTIONAL)
    # stuff to do if all other conditions are FALSE
}
```

- ► Some consider the catch-all else clause to be dangerous
 - Some style guides even recommend avoiding the use of else for better readability
- Best practice is to make conditions explicit whenever possible

Loops: Note!

- ► Loops in R are inefficient²
- For many cases, there will be a much faster, vectorized alternative to looping
- ► We cover loops because there are some *rare* cases in which a loop might make more sense, but in general, loops should be avoided when writing R

²Not entirely true, but loops are still best avoided for other reasons too.

Loops

for statements

```
for (ind in sequence/set) {
    # iterate over sequence or elements of a set
    # do stuff
}
```

while statements

```
while (condition) {
    # stuff to do while the condition is TRUE
    # the condition must become FALSE at some point!
}
```

Loops: Example

```
for (i in 1:3) {
  print(paste('iteration', i))
## [1] "iteration 1"
   [1] "iteration 2"
## [1] "iteration 3"
while (i \ge 0) {
  print(paste('de-iteration', i))
```

```
## [1] "de-iteration 3"
## [1] "de-iteration 2"
## [1] "de-iteration 1"
## [1] "de-iteration 0"
```

User Defined Functions

Write your own functions in the form

```
name_of_function <- function(arguments) {
    # do some stuff with arguments
    return(result)
}</pre>
```

▶ You can use your functions like any other function, e.g.,

```
name_of_function(arguments) # gives you the 'result'
```

User Defined Functions: Example

▶ Write a function that will take a vector in \mathbb{R}^3 and tell you if you can make a triangle or not

```
is good <- function(vec) {</pre>
  if (length(vec) != 3) {
      # it's always a good idea to make sure your
      stop('is good requires a vector of length 3')
 for (i in 1:3) {
    if (vec[i] > sum(vec[-i])) {
      return(FALSE)
 return(TRUE)
```

apply

- ▶ Loops in R are inefficient, and best avoided if possible
- Vectorize operations whenever possible.
- Use apply to loop over rows/columns of matrix/table

```
apply(DATA, MARGIN, FUNCTION)
# MARGIN = 1 applies FUNCTION to each row of DATA
# MARGIN = 2 applies FUNCTION to each column of DATA
# MARGIN = c(1,2) applies FUNCTION to both
# rows and columns of DATA
```

lacktriangle e.g., to get the variance of each row/column of a matrix X,

```
apply(X, 1, var) # variance of rows
apply(X, 2, var) # variance of columns
```

Exercise

The Question

You are given three sticks, each of a random length between 0 and 1.

What's the probability you can make a triangle?

- ► The answer is 1/2
- ▶ Use R to simulate 100,000 times and estimate the answer by
 - 1. generate 100,000 triplets of uniform (0, 1) random variables
 - find the portion that can be made into a triangle (hint: use the is_good function)

Answer 1: Using a for loop

```
system.time({  # measure execution time
  N <- 1e5;
 m < -0
  for (i in 1:N) {
   X \leftarrow runif(3)
    if (is_good(X)) {
        m < -m + 1
  cat(sprintf('\%.2f\%\n', m / N * 100))
```

```
## user system elapsed
## 1.456 0.000 1.478
```

49.83%

Answer 2: No loop - possibly faster but uses more memory

```
system.time({  # measure execution time
  N <- 1e5;
  X <- matrix(runif(N*3), nrow=3, ncol=N)
  m <- apply(X, 2, is_good)
  cat(sprintf('%.2f%%\n', mean(m) * 100 ))
})</pre>
```

```
## user system elapsed
## 1.240 0.016 1.292
```

49.87%

Packages

Installing R Packages

- R has many (MANY) packages created by other users that implement state-of-the-art tools (e.g., data manipulation, statistical models)
- ► These packages can be downloaded from the Comprehensive R Archive Network (CRAN)
- ▶ This is as simple as running a single line of code:

install.packages("package name")

- You will have to select one of many CRAN mirrors (copies across different servers) from which to download the package from
- ► For example, to install the tidyverse package, run

install.packages("tidyverse")

You only need to do this once for each machine

Loading Packages

- Once you've installed a package on a machine, you can load the package into your current workspace with the library() command
- ▶ For example, to use the tidyverse package, first load it with

library("tidyverse")

- You can also use specific functions from a package without loading it, by telling R which package the function belongs to with a namespace prefix, package_name::.
- For example, to use the round_any() function from the plyr package, without actually loading plyr, write

```
# Assuming plyr is installed
plyr::round_any()
```

Namespace collision

- ► One of the (unfortunately many) things that R is bad at is preventing namespace collisions
- ► For example, the packages plyr and dplyr³ have functions that are named the same (e.g, mutate(), summarize()), and if you ever load both, R will only "see" the function belonging to the package you loaded later
- ► So beware of what packages you load, and if you only intend to use a function or two, consider just specifying the namespace with ::, instead of loading the whole package.

³dplyr is part the tidyverse, and loaded when you load tidyverse