R Tutorial

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Special thanks to Sam Maher who wrote an R Tutorial in 2018 that influenced this.

Goal

This is R Studio. The goal of this lecture is to see some basics about R so that we can dive into more exciting things tomorrow.

Overview of R Studio

Layout of RStudio:

- Script
 - Where you will be writing your own programs
- Environment/../Git
 - Mostly just environment and Git
 - Show which data objects you have loaded in memory
 - Eventually where you'll do version control with GitHub (last day)
- Files/Plots/Packages/Help/Viewer
 - Helps you load packages and other files to load
 - Default window when you're trying to get with a function from a package (we will get to this)
- Console/Terminal
 - where the code actually run
 - Scripts executes in console
 - Code disappear in the console, whereas a script saves your code
 - To run something in the console, type it in and hit "enter"
 - Important to know there is a "terminal" in R studio. Again, just know it's there.

Console

- We can type code straight into the console and run it.
- The console in R Studio knows your running R code.

```
# Type the following in the console, then press enter 2+3
```

[1] 5

```
# variable assignment happens with an arrow (on a mac can do option + -) a <- 2 b <- 3 a + b
```

[1] 5

How to write and execute a script

Writing Scripts and Running Code

- New Script: File > New File > R Script (or R Markdown, which is what I'm using now), or just the New Document script in the upper left hand corner of the screen > R Script (or R Markdown)
- Keyboard shortcut to run a section of code: highlight or put your cursor on that line and hit Ctrl + Enter (Windows) or Command + Enter (Mac)
- Executing code: the run button at the top right corner of the script
- For R Markdowns, we also have the Knit button at the top of the script.
 - That will run all code and "knit" it together into a pdf or HTML or doc
 - File type is specified at the top of the R Markdown file

R Markdown

- I wrote this pdf using an R Markdown (show them quickly).
- I love these because I'm able to write lots of notes to myself while I'm coding
 - it produces a nice shareable file
 - can share my notes, code and results with others.
- You can do the minipsets with a script or r markdown.

Mini psets

- If you have no experience with R, start with a script.
- If you want to do a R Markdown and am used to them, that's fine.

You can "clean up" the Environment after you've executed code by clicking the broom icon. This will delete everything in your environment.

Basic Data Types and COMMENTS

```
# This is a comment, you can use '#' to write notes to yourself in your code
# - Comments are what make or break good coders, and coders who can collaborate with others.
# - If you ever thing you writing "too" many comments, you almost always not.
# - The things you think are obvious in your code wont be to others (nor yourself in a year when you ge
# Numeric -- integer: no decimal points
```

```
myInt <- 1

# Numeric -- double: decimal points
myNum <- 2.4

# logical (Boolean): a true/false statement. Use () to evaluate if something is true or false
myBool_1 <- (3 < 4)
myBool_2 <- (3 > 4)

# character (string)
myChar_a <- "a"
myChar_b <- 'b'</pre>
```

Ways to store datatypes

```
# vector: can only be a vector of one data type (numeric, logical, string)
myVec_n \leftarrow c(1, 2, 3, 4, 5)
myVec_s <- c(str, "b", "c")</pre>
# matrix: can only be a matrix of one data type
myMat_n <- matrix(c(myVec_n,</pre>
                6, 7, 8, 9, 10),
              nrow = 2,
              ncol = 5)
# Lists: Very powerful, but somewhat confusing. For now, just know they exist
myList <- list(2, "c", myMat_n)</pre>
myList[[1]] # returns numeric
## [1] 2
myList[[2]] # returns string
## [1] "c"
myList [[3]] # returns matrix
        [,1] [,2] [,3] [,4] [,5]
## [1,]
                           7
          1
                3
                      5
## [2,]
                      6
                               10
# data frame: can have multiple data types
myDF <- as.data.frame(myMat_n)</pre>
colnames(myDF) # these don't mean anything to me
```

```
colnames(myDF) <- c("age_yr", "weight_lb", "income_$", "height_ft", "height_in")</pre>
```

Dataframes:

- Like matrices
- Can have different data types in each column
- Reference specific columns using the "\$" operator, followed by the name of the column

```
# investigate one column
myDF$age_yr

## [1] 1 2

#create a new column
myDF$nonsense <- myDF$age_yr + myDF$weight_lb</pre>
```

- For the most part, you'll be loading new data by reading a csv
- You might have to create one at some point.
- By looking at how they're created we can get a better sense of what goes into them

```
# Create the data frame
BMI <- data.frame(
    gender = c("Male", "non-binary", "Female"),
    male = c(T, F, F),
    height = c(152, 171.5, 165),
    weight = c(81, 93, 78),
    Age = c(42,38,26)
)
# Try referencing one column
BMI$male # version 1</pre>
```

[1] TRUE FALSE FALSE

```
BMI[,2] #version 2
```

[1] TRUE FALSE FALSE

```
# Try referencing one row
BMI[1,]
```

```
## gender male height weight Age
## 1 Male TRUE 152 81 42
```

```
# Try referencing one cell
BMI$height[1] # version 1
```

[1] 152

```
BMI[1,3] # version 2
## [1] 152
```

A word of caution

• Make sure you don't over write your variables by accident.

```
# assigning new value to same variable (something to do carefully)
a <- 5
a <- a + 1 # If you run this line more than one, you will NOT get six
a

## [1] 6

# assigning new value to new variable
a <- 5
a_new <- a + 1 # If you run this line more than one, you WILL get six
a_new
## [1] 6</pre>
```

Functions

Functions: once you have initialized them, they take in an input, perform a set of operations on them, and then give you some return value.

Example on board

- consider the function: $myF(x) \{ y < -x + 3; return(y) \}$
- what does myF(3) return? 6

Points:

- These are helpful when you have something that you do often
- Rule of thumb: if you're copying and pasting code 3 times or more, make function
- (i say if you are going to copy past ever, because even if you think it'll only be twice it'll probably be more)
- Recent example for me:
 - wrote a function to take a date and return the season
 - Wrote a function to get kelvin and return Fahrenheit

```
myF <- function(x){
    y <- x - x^2
    return(y)
}
myF(.5)</pre>
```

```
## [1] 0.25
```

```
myF(.25)
## [1] 0.1875
myF(.7)
## [1] 0.21
Loops
   • for loops: iterates through a task for a set number of times
   • Consider these loops (psuedo code):
       - For (i in 1 through 4) { print i }
       - For (i in 1 through 4) { print i / 4}
   • Can be helpful when
       - Iterate through a column of data and do something to each row
       - Construct a new column and want to construct each row by scratch
#simple
for (i in 1:4){
  print(i)
## [1] 1
## [1] 2
## [1] 3
## [1] 4
# more involved
for (i in 1:4){
  print(i/4)
}
## [1] 0.25
## [1] 0.5
## [1] 0.75
## [1] 1
# combining loop and function
for (i in 1:4){
  y = myF(i/4)
  print(y)
## [1] 0.1875
## [1] 0.25
## [1] 0.1875
## [1] 0
```

```
# manipulating a column (from above: version one of referencing a cell)
for (i in 1:length(BMI$Age)) {
   BMI$Age[i] <- BMI$Age[i] + 1 # everyone aged on year
}

# Making a new column
for (i in 1:length(BMI$Age)) {
   BMI$Age[i] <- BMI$Age[i] + 1 # everyone aged one year
}</pre>
```

If statements

- sometimes you want to execute a task ONLY if a certain condition is met.
- Open the BMI df:
 - Our RA did not record men's ages right
 - All men are actually 3 years younger than what's recorded
 - What would the correct DF look like?
- If statements let you fix a mistake like this
- Also demonstrates why the Boolean (true/false or indicator) variable is so powerful

```
# goes through each row and changes age if someone is male
for (i in 1:length(BMI$male)) {
   if (BMI$male[i] == TRUE) {
     BMI$Age_new[i] <- BMI$Age[i] - 3
}else{
     BMI$Age_new[i] <- BMI$Age[i]
}
}</pre>
```

Other R Tutorials

Princeton Getting Started with R UCLA Getting Started with R

Some specific packages

We haven't covered packages yet, but a few good resources for the tomorrow.

ggplot dplyr and tidyr