

MAD – Data Analysis & Biostatistics in R

Getting to Work

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Administrative Notes

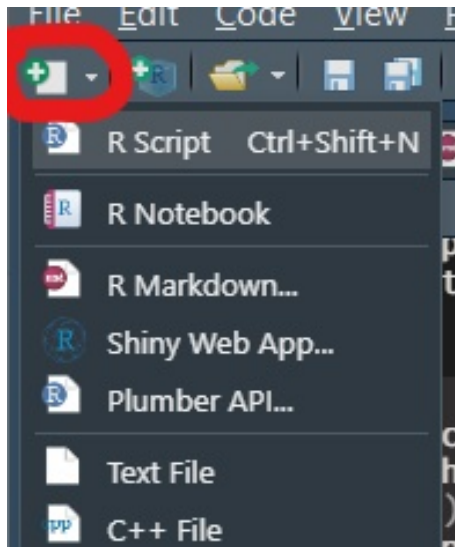
- Piazza is a **bust**. Not going to use it
 - ▶ Questions through all the traditional means
- Chapter 9 has had a major revision. If you downloaded before 5/9
 - ▶ Need a new download

Section 1

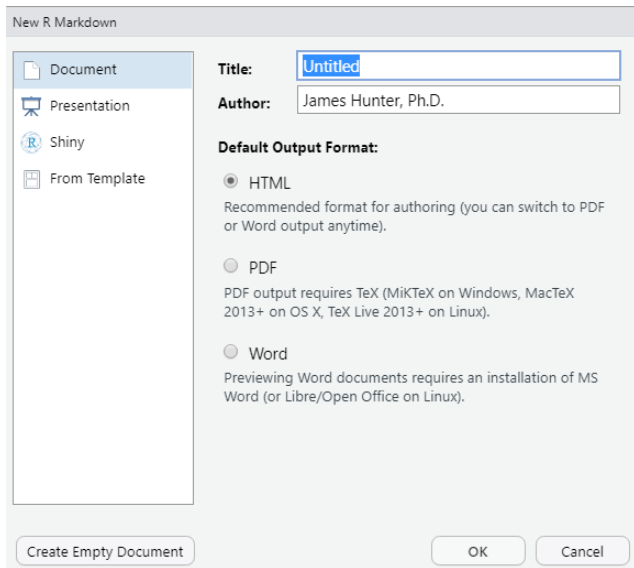
Submitting Homework and Other Documents

New R Markdown Document

- Go to upper left corner of screen (under File menu item)



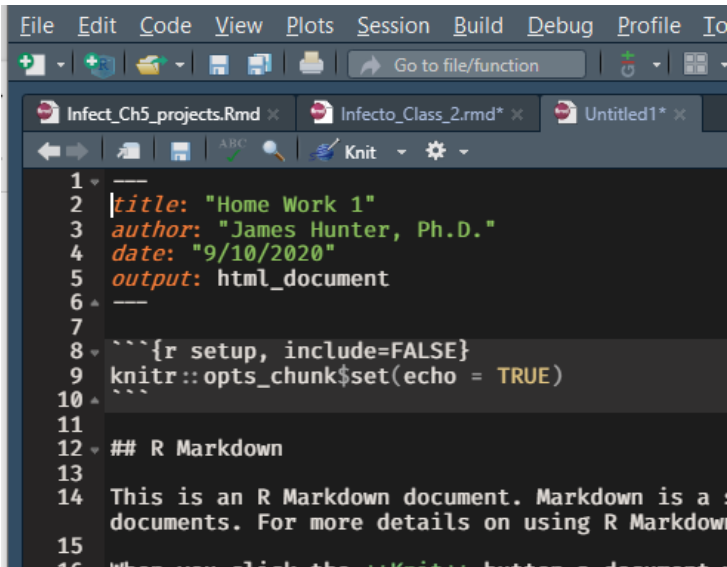
Click on “R Markdown. . .”



Next Step

- Click on “Document” in left column
- Fill in title: “HomeWork 1”
- Fill in Author: your name
- Select HTML or PDF
 - ▶ PDF *only* if you know you have LaTeX installed on machine
- Click OK button

Here is your RMD Document



The screenshot shows the RStudio interface with a dark theme. The menu bar at the top includes File, Edit, Code, View, Plots, Session, Build, Debug, Profile, and Tools. Below the menu bar is a toolbar with icons for creating new files, opening files, saving, and other standard operations. The tab bar shows three open files: 'Infect_Ch5_projects.Rmd', 'Infected_Class_2.rmd*', and 'Untitled1*'. The main editor area displays the content of 'Infect_Ch5_projects.Rmd'. The document is an R Markdown file with a YAML header, a knitr chunk, and a text block. The line numbers 1 through 16 are visible on the left side of the editor.

```
1 ---
2 |title: "Home Work 1"
3 |author: "James Hunter, Ph.D."
4 |date: "9/10/2020"
5 |output: html_document
6 ---
7
8 ```{r setup, include=FALSE}
9 knitr::opts_chunk$set(echo = TRUE)
10 ```
11
12 ## R Markdown
13
14 This is an R Markdown document. Markdown is a
15 documents. For more details on using R Markdown
16 When you click the Knit button a document
```


Working with .rmd Document

- Select and Erase all content from ## R Markdown down to end
- Copy question 1 from homework sheet to 2 lines after the end of block
- To show solution of question 1, open code block
 - ▶ Can type CTRL-ALT-I
 - ▶ Can type format characters (see delimiters of block below)

```
```${r setup, include=FALSE}  
knitr::opts_chunk$set(echo = TRUE)
library(tidyverse)
```
```

Ready to Answer Questions

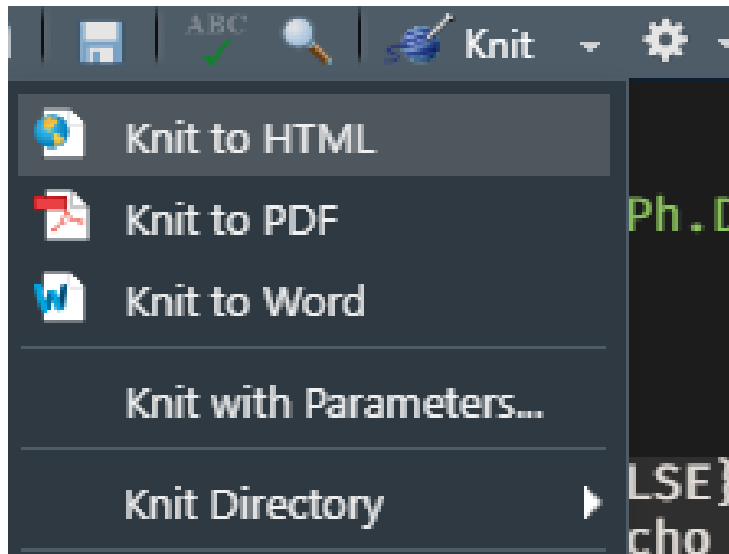
- Write code for your answer
 - ▶ To test your code, CTRL-Enter each line
 - ★ Or hit green arrow at top right of code block
- Go on to next question through to end
- Save the file (often)

Section 2

Transforming your .rmd to HTML

Transforming your .rmd to HTML

- Click on Knit above document



- Click on Knit to HTML
- If no major programming errors, it will do the calculations and give a result in web browser form

Raw .rmd File

```
1  ---
2  title: "Home Work 1"
3  author: "James Hunter, Ph.D."
4  date: "9/10/2020"
5  output: html_document
6  ---
7
8  ```{r setup, include=FALSE}
9  knitr::opts_chunk$set(echo = TRUE)
10 ```
11
12 ## Question 1
13
14 ```{r q1}
15
16 res <- ((log10(67000) * 92) + 36) / sqrt(759)
17 res
18 round(res, 1)
19
20 ```
21
22 [1] 17.42286
23 [1] 17.4
```

Home Work 1

James Hunter, Ph.D.

9/10/2020

Question 1

```
res <- ((log10(67000) * 92) + 36) / sqrt(759)
res
```

```
## [1] 17.42286
```

```
round(res, 1)
```

```
## [1] 17.4
```

Section 3

Homework from Week One

Question 1 - R as Calculator and PEMDAS

- Evaluate this expression in R; round to 1 decimal place

$$\frac{\log_{10}(67000) * 92 + 36}{\sqrt{759}}$$

```
res <- ((log10(67000) * 92) + 36) / sqrt(759)
res
```

```
## [1] 17.42286
```

```
round(res, 1)
```

```
## [1] 17.4
```

Question 2 – Indexing a Vector

- Vector of 20 random numbers created with

```
set.seed(42) # so everyone produces the same answer  
x <- round(rnorm(20, mean = 100, sd = 10), 2) # returns v  
x
```

```
## [1] 113.71  94.35 103.63 106.33 104.04  98.94 115.12  
## [11] 113.05 122.87  86.11  97.21  98.67 106.36  97.16
```

- What is the 2nd element of `x`?
 - ▶ `x[2] = 94.35`
- What is the class and type of `x`?
 - ▶ `class(x) = numeric`
 - ▶ `typeof(x) = double`
- What is the maximum value of `x`?
 - ▶ `max(x) = 122.87`

Question 3 - Information about Data Frame

- Based on data frame “einstein_soro.rds”
 - ▶ Binary stored form of data about COVID-19 sorological tests performed at Albert Einstein Hospital
 - ▶ From FAPESP Data Consortium files
- 1st Step – Load Data Frame

```
soro <- readRDS("C:/Users/james/OneDrive/Documents/MAD/MAD-Infecto-2020/einstein_soro.rds")
```

```
## 'data.frame':    200 obs. of  10 variables:
## $ pacid      : chr  "b6d668e4f818f7b3643ed593b8fb902bf9d2501e" "a090625661c06e9c
## $ dt_collect: chr  "28/05/2020" "11/05/2020" "16/06/2020" "10/06/2020" ...
## $ analysis   : chr  "IgM, COVID19" "IgG, COVID19" "IgG, COVID19" "COVID IgM Inte
## $ result     : chr  "0.74" "0.03" "0.02" "Não reagente" ...
## $ unit       : chr  "AU/ml" "AU/ml" "AU/ml" "NULL" ...
## $ reference  : chr  "<=0.90" "<=0.90" "<=0.90" "" ...
## $ sex        : Factor w/ 2 levels "female","male": 2 1 1 2 1 1 1 2 1 2 ...
## $ birth_yr   : num  1989 1975 1997 2006 1983 ...
## $ uf         : Factor w/ 25 levels "AC","AL","AM",...: 24 9 24 24 24 24 24 24 24
## $ city       : Factor w/ 21 levels "BARUERI","CAMPINAS",...: 19 NA 19 19 19 19 19
```

Questions about soro

- a. How many cases are in this data base?
 - ▶ First line of `str()` tells you: 200 obs. of 10 variables
 - ▶ Can also calculate as `nrow(soro) = 200`
- b. How many variables are in this data base?
 - ▶ First line of `str()`
 - ▶ Can also calculate as `length(soro) = 10`
- c. What is the class and type of soro?
 - ▶ `class(soro) = data.frame`
 - ▶ `typeof(soro) = list`
- d. What is the date of collection (`dt_collect`) of the 3rd case?
 - ▶ `soro$dt_collect[3] = 16/06/2020`

soro Questions e and f

- e. Print the 5th to the 10th birth years (`birth_yr`).

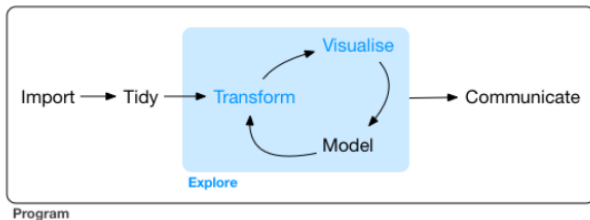
```
soro$birth_yr[5:10]
```

```
## [1] 1983 1963 1988 1971 1968 1976
```

- f. How many different cities are represented in this sample?
 - ▶ `levels(soro$city) = 21`

Section 4

Data Analysis Projects



Project Workflow

- Tidyverse diagram shows principal steps in organizing computational tasks
- **However**
 - ▶ Much planning necessary before we even start with *Import*
 - ▶ Right at start of project

Initial Issues in Project Planning – Biological/Medical

- What is my research topic? - What is the large-scale question?
- What has science said on this topic previously?
 - ▶ Literature search
 - ▶ What sources to use (PubMed, Google Scholar, etc.)?
 - ▶ When will I have read enough?

Initial Issues in Project Planning – Technique

- What is the type of study I will conduct?
 - ▶ Experimental
 - ▶ Observational
 - ▶ Case-Control
 - ▶ Other type?
- What is my research hypothesis?
- What experiments/field work/data gathering support that?
- What will be my primary analytic technique?
 - ▶ What have others done?
 - ★ Were they effective?
 - ★ Produce interesting results?

Data for My Project

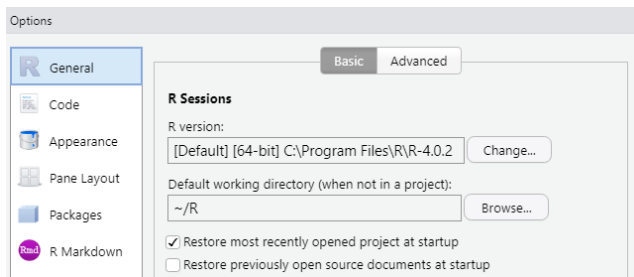
- What data do I need?
 - ▶ Categorical variables
 - ▶ Result variables
 - ★ Numeric
 - ★ Other Data Types
- How much data do I need?
 - ▶ Number of cases (**n**)
 - ▶ Statistical power to achieve significance

Section 5

Projects in RStudio

Active Directories

- When you start RStudio
 - ▶ Uses the default directory from your “Tools/Global Options” setting



- If you want a different directory for an analysis, need to use `setwd()` to direct R there

setwd() and getwd()

- To set a new working directory, need to use command `setwd()`
 - ▶ With full path name in quotes as argument
- Example - for my thesis research in HIVAIDS



```
setwd("C:/Users/james/OneDrive/Documents/HIVAIDS/Drug_")
```

- Complementary function `getwd()`
 - ▶ No argument
 - ▶ Shows what is current active directory

```
## [1] "C:/Users/james/OneDrive/Documents/MAD/MAD-Infecto"
```

Why Do We Need *Projects*?

- You will work on different projects that have
 - ▶ Different databases
 - ▶ Different scripts
 - ▶ Different documents
- Throwing all your files from all projects in 1 R directory will make you crazy
 - ▶ Many projects could have upwards of 500 files
- Planning project workflow implies organization
- Build an R data analysis project around a reserved workspace

RStudio Projects Allow Group Members to Work Together Better

- Everyone has a different directory structure on their machines
 - ▶ I have projects in my R directory
 - ▶ All my HIV projects are in my HIVAIDS directory
 - ▶ All the versions of this course are in my MAD directory
- What is your directory structure?
- Projects create their own working directory (active directory)
 - ▶ Same for everyone who uses the same project structure and file
 - ▶ No more constantly having to type km-long file paths
 - ★ And getting them wrong

If the first line of any of your R scripts is

```
setwd("C:\\Users\\jenny\\path\\that\\only\\I\\have")
```

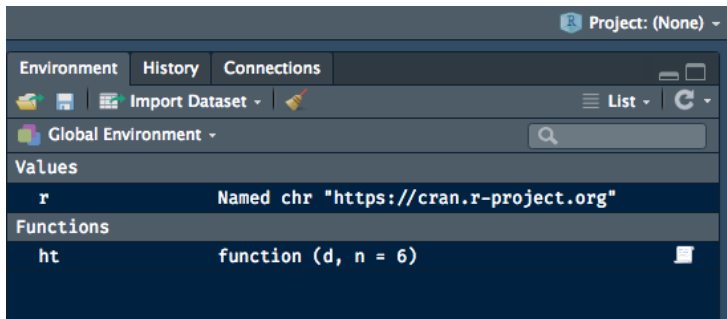
*Someone from the Tidyverse team will come to your office
and **set your computer on fire***

They have promised

The best way to avoid such a consequence is create a new RStudio project for each new activity

How to Create a RStudio Project

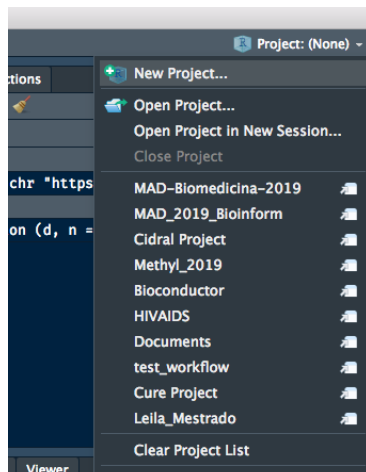
- Find “Project:(None)” text
 - ▶ Top right of screen above Environment, etc. tab



- So far, no active project

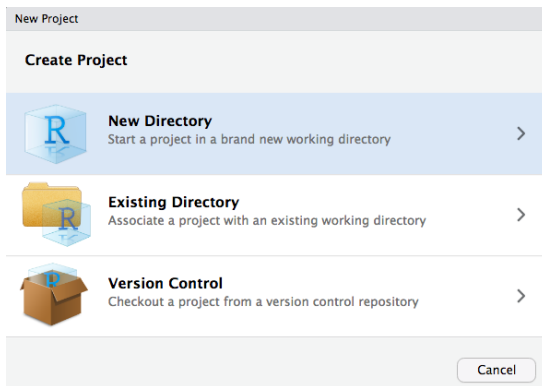
Create New Project

- Click on arrow to right of “(None)”
- Drop down menu will open
- Click “New Project ...” option



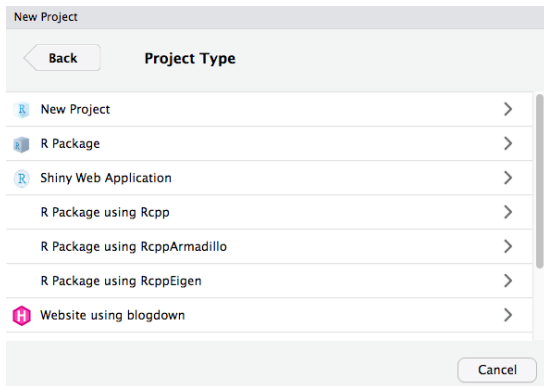
Create Project Window

- Options to choose a new or existing directory
- Version Control option more advanced
- For this case, choose “New Directory”



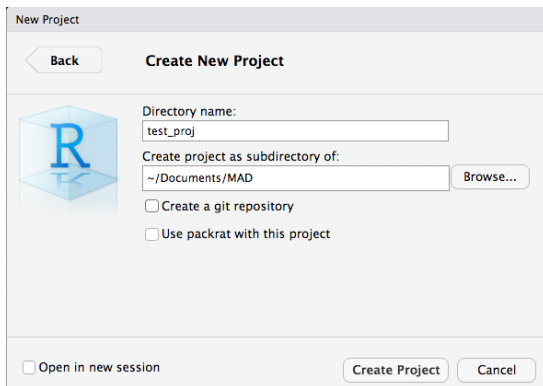
Types of New Projects

- RStudio has many types of new projects
- We want just a “New Project”: click on that



Final Creation Screen

- Give the project a name and a location
- Click Create Project button
- For now, ignore git and packrat boxes



New Project

Back

Create New Project

Directory name:
test_proj

Create project as subdirectory of:
~/Documents/MAD Browse...

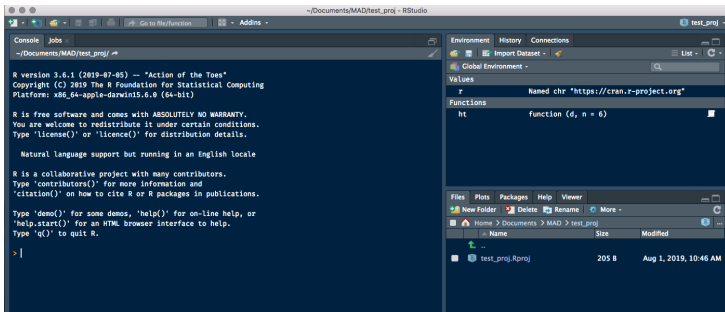
☐ Create a git repository

☐ Use packrat with this project

☐ Open in new session

Create Project Cancel

Here Is Your Project



Subdirectories

- Projects can become large, VERY LARGE
 - ▶ Thesis project: 1,978 files, 383 sub-directories, 425 MB
- Sub-directories for files of different types.
 - ▶ Like setting up a project
 - ▶ Helps keep files organized
- My preferred subdirectory structure for a new project
 - ▶ Data
 - ▶ Raw Data
 - ▶ Docs
 - ▶ Graphics
 - ▶ Programs/Scripts
 - ▶ Slides
- Keep *raw data* separate from any other data files
 - ▶ Cleaned, subsets, etc.
 - ▶ You may need it again!

Section 6

We're HERE!

Problem

- Various projects and various directories could have very different paths
- Path to test_proj on my old Mac Air:
 - ▶ `"/Users/jameshunter/Documents/MAD/test_proj"`
- Path to test_proj on my current Windows laptop:
 - ▶ `"C:/Users/james/OneDrive/Documents/MAD/test_proj"`
- If I wanted to use my path to go to a graphics subdirectory that both had
 - ▶ It would be a mess!

Solution: `here::here()`

- `here` package does one thing
 - ▶ Lets you know what is full path to working directory
- Simple form: no arguments

```
here::here()
```

```
## [1] "C:/Users/james/OneDrive/Documents/MAD/MAD-Infecto-2020"
```

Go to graphics Subdirectory of Current Project

- Give the function the argument "graphics"
- Thereafter, can use variable gr to refer to full path of location

```
gr <- here::here("graphics")  
gr
```

```
## [1] "C:/Users/james/OneDrive/Documents/MAD/MAD-Infecto-2020/graphics"
```

- Note new package::function notation
 - ▶ 2 colons call a function directly from a package without having to load package

Section 7

Scripts and Programming

Why Scripts? - A Review

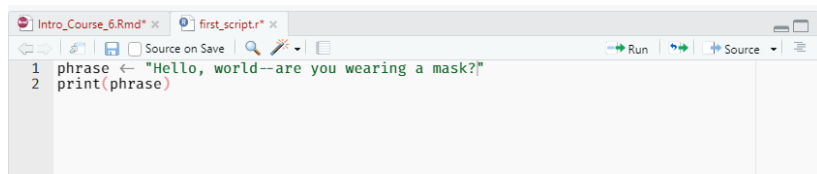
- Combine commands into coherent whole
- Store your commands
 - ▶ Reuse or recall at later date
- Facilitate correction of error and re-running program
- Scripts can be either
 - ▶ Separate documents (.r files)
 - ▶ Blocks of code in a larger .rmd document
 - ★ Like this one

Basic Scripts

- Open new script document (.r) with
 - ▶ CTRL-Shift-N
 - ▶ Plus icon in upper left of screen
- Editor is a basic text editor
- Example
 - ▶ Assign sentence: “Hello, world—wear a mask!” to the variable phrase
 - ▶ Print phrase to the screen

Program Text

```
phrase <- "Hello, world--wear a mask!"  
print(phrase)
```



- Execute program with Run or Source buttons

Section 8

Loops

- Way to execute series of commands repeatedly
- Executes until encounters condition that ends it
- 2 Flavors
 - ▶ **for** loops
 - ▶ **while** loops

for Loops

- Format

```
for(var in vector) {  
  code  
  code  
  code  
}
```

- for starts loop
- Number of repetitions of loop inside the parentheses
 - ▶ Repetitions:
 - ★ var – variable name
 - ★ vector – values variable can assume
- {} - pair of curly braces
- Inside braces: lines of code you want to execute

Example for Loop

- Square whole numbers from 5 to 10
- Denote variable as i
- Vector
 - ▶ `c(5, 6, 7, 8, 9, 10)`
 - ▶ `5:10`

```
for(i in 5:10){  
  print(i^2)  
}
```

```
## [1] 25  
## [1] 36  
## [1] 49  
## [1] 64  
## [1] 81  
## [1] 100
```

What the Loop Did

- 1st Iteration
 - ▶ Assigned 5 to i
 - ▶ Squared i (i^2)
 - ▶ Wrote i to screen (25)
- 2nd Iteration
 - ▶ Assigned 6 to i
 - ▶ Squared i
 - ▶ Wrote i to screen (36)
- Same for 3rd to 6th
- At end of 6th iteration ($i <- 10$)
 - ▶ Could not find a new value to give to i
 - ▶ Stopped

More Realistic Example

- Take a DNA string and transform it to a vector of letters
- Length of the string unknown
- Can be calculated with `nchar()` function: gives number of characters

```
library(tidyverse)
seq <- "CCTCAAATCACTCTTTGGCAACGACCCTTAGTCACAATAAAAGTAGGGGA"
seq_length <- nchar(seq)
seq_vector <- character(seq_length) # create vector to hold result
for (i in 1:nchar(seq)) {
  seq_vector[i] <- tolower(str_sub(seq, i, i))
}
seq_vector
```

```
## [1] "c" "c" "t" "c" "a" "a" "a" "t" "c" "a" "c" "t" "c" "t" "t" "t" "g" "g" "c"
## [20] "a" "a" "c" "g" "a" "c" "c" "c" "t" "t" "a" "g" "t" "c" "a" "c" "a" "a" "t"
## [39] "a" "a" "a" "a" "g" "t" "a" "g" "g" "g" "g" "a"
```

while Loops

- Similar structure to for loops but work differently
- Keeps repeating until condition you put in `while` statement turns `FALSE`
- Consult text
- Used much less than `for`

Loops vs. *Vectorized* Statements

- R is vectorized
 - ▶ Interpreter will act on all elements of a vector at once
 - ▶ Don't need to loop item by item
- Should we use loops at all?
- Computer speed today reduces the speed advantage of vectorized statements
 - ▶ Except for VERY large datasets, loops are practically equivalent
 - ▶ Millions of cases, hundreds of variables
- Don't give up on loops
 - ▶ Logic is usually easier to program than many vectorized functions
 - ▶ Can save much overall time in terms of programming

Section 9

Conditional Statements (*if... then... else*)

Thought Problem

- We are reading sequences of nucleotides
 - ▶ DNA alphabet: ACGT
 - ▶ RNA alphabet: ACGU
- We could say
 - ▶ *If the sequence is DNA then the alphabet is ACGT.*
- More complete version
 - ▶ *If the sequence is DNA then the alphabet is ACGT else the alphabet is ACGU.*

If Statements in R

- Basic structure

```
if(condition) {  
  code  
  code  
  code  
} else {  
  code  
  code  
  code  
}
```

Simple Example: No else Clause

```
x <- 0

if (x == 0) {
  print("x equals 0")
}
```

```
## [1] "x equals 0"
```

More Complete Logical Test

```
x <- 0

if (x == 6) {
  print("x equals 6")
} else {
  print("x does not equal 6")
}
```

```
## [1] "x does not equal 6"
```

3 Nested Conditions – DNA/RNA Example

```
seq_type <- "DNA"

if (seq_type == "DNA") {
  print("ACGT")
} else {
  if (seq_type == "RNA"){
    print("ACGU")
  } else {
    print("sequence neither DNA nor RNA")
  }
}
```

```
## [1] "ACGT"
```

ifelse() Function

- For simple logical tests
- `ifelse(test, true, false)`
- 3 Arguments
 - ▶ `test`: logical test
 - ▶ `true`: result if TRUE
 - ▶ `false`: result if FALSE
- If result of test is TRUE, returns true value
- If result of test is FALSE, returns false value

ifelse() Example

```
X <- 2
res <- ifelse(X > 10, "greater than", "less than or equal to 10")
paste("X is", res, "10")

## [1] "X is less than or equal to 10"
```

More Complicated if...then...else Logic

- Function in dplyr package: `case_when()`
- Can more comfortably handle large number of alternative cases
- More advanced

Section 10

Importing Files to R

Types We Will Work With

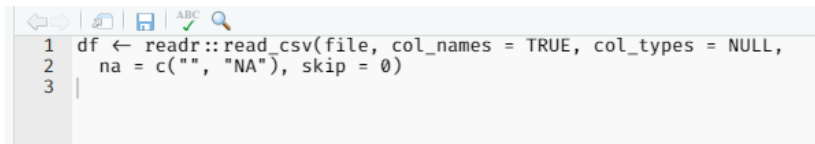
- Comma-Delimited Files (.csv)
- Excel Files (.xls or .xlsx)
- FASTA files

Section 11

Comma-Delimited Files (.csv)

Importing a .csv File

- A plain text file that is in rectangular form
- Commas separating the fields
- Interpretable by Excel
- Most common file type you will import
- Use `readr::read_csv()` to import it
 - ▶ Function reads file
 - ▶ Treats commas as Excel does: separators between columns
 - ▶ Imports it to a tibble



```
1 df <- readr::read_csv(file, col_names = TRUE, col_types = NULL,  
2   na = c("", "NA"), skip = 0)  
3 |
```

read_csv() Arguments

- `file` = file path to reach the file
 - ▶ Use of `here::here()`
- `col_names` = Is the first row the names of the columns or data?
 - ▶ TRUE means 1st row are the column names; `read_csv()` assumes so
 - ▶ FALSE means it has data rather than names

col_types = Argument

- read_csv() will try to guess the correct data type based on the column content
- If you want to let R guess, enter NULL or omit the argument
- If you want to specify column types, use a string of characters from list:
 - ▶ c = character
 - ▶ i = integer
 - ▶ n = number
 - ▶ d = double
 - ▶ l = logical
 - ▶ f = factor
 - ▶ D = date
 - ▶ T = date time
 - ▶ t = time
 - ▶ ? = guess
 - ▶ - to skip the column

Notes on `col_types` =

- Column string must have **exactly** the same number of characters as the data has columns
 - ▶ If you have a data set with five columns, the string must have five characters
 - ★ Ex. `col_types = "cfn-c"`
 - ★ Means character-factor-numeric-skip-character

Two Other Key Arguments

- Specifying missing data

- ▶ `read_csv()` assumes that missing data will be encoded with
 - ★ the string "NA" or a blank string
- ▶ If data uses other codes, need to specify them
 - ★ "99" is a common indicator of missing data in social science
 - ★ Must specify the `na =` argument as `na = c("", "NA", "99")`

- Skip rows: `skip =`

- ▶ Datasets with metadata, etc. in 1st few rows
- ▶ To skip these, set a positive value for the `skip =` argument
 - ★ Ex. `skip = 4` will start reading the database on the fifth row

Simple .csv Example

- `classe_simples.csv` records info on a small school class
 - ▶ Name, home state, age and scores on a standardized test (1 - 10)

| | A | B | C | D | |
|----|---------|--------|-------|------|--|
| 1 | nome | estado | idade | nota | |
| 2 | Pedro | DF | 34 | 4.12 | |
| 3 | Joana | PE | 59 | 6.1 | |
| 4 | Isabel | SP | 26 | 9.34 | |
| 5 | Pedro | GO | 67 | 6.75 | |
| 6 | Bia | RJ | 53 | 5.85 | |
| 7 | Pedro | RJ | 51 | 7.49 | |
| 8 | Tomas | TO | 46 | 5.78 | |
| 9 | Isabel | PE | 66 | 4.2 | |
| 10 | Tomas | GO | 21 | 8.78 | |
| 11 | Ricardo | TO | 51 | 8.58 | |
| 12 | | | | | |

Importing classe_simples.csv

- Assign the function to a name for the data frame classe1
- Run a simplified command with no column spec
 - ▶ R will guess

```
classe1 <- read_csv(here::here("classe_simples.csv"))  
classe1
```

Result

```
## Parsed with column specification:
## cols(
##   nome = col_character(),
##   estado = col_character(),
##   idade = col_double(),
##   nota = col_double()
## )
```

```
## # A tibble: 10 x 4
```

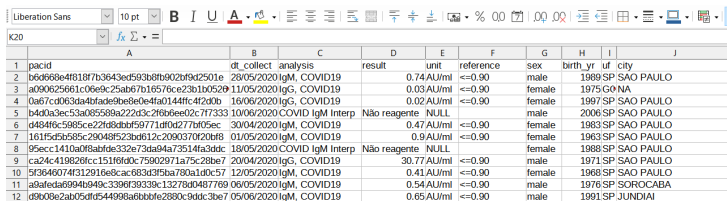
| | nome | estado | idade | nota |
|-------|---------|--------|-------|-------|
| | <chr> | <chr> | <dbl> | <dbl> |
| ## 1 | Pedro | DF | 34 | 4.12 |
| ## 2 | Joana | PE | 59 | 6.1 |
| ## 3 | Isabel | SP | 26 | 9.34 |
| ## 4 | Pedro | GO | 67 | 6.75 |
| ## 5 | Bia | RJ | 53 | 5.85 |
| ## 6 | Pedro | RJ | 51 | 7.49 |
| ## 7 | Tomas | TO | 46 | 5.78 |
| ## 8 | Isabel | PE | 66 | 4.2 |
| ## 9 | Tomas | GO | 21 | 8.78 |
| ## 10 | Ricardo | TO | 51 | 8.58 |

classe_simples.csv Results

- First line contained variable names; R used them
- If we had wanted to specify columns, we could have used
 - ▶ `col_types = "ccnn"`

More Complex Data Set

- Example we will use throughout data cleaning process
- 99 case version of Einstein Hospital COVID-19 serological tests
 - ▶ `einstein_soro_tests.csv`
- First look at “messy”, ie. real data



| | A | B | C | D | E | F | G | H | I | J |
|----|--|------------|------------------|--------------|-------------|-----------|--------|----------|----|-----------|
| 1 | pacid | dt_collect | analysis | result | unit | reference | sex | birth_yr | uf | city |
| 2 | b6d668e4f810f7b3643ed593b8fb902bf9d2501e | 28/05/2020 | IgM, COVID19 | | 0.74 AU/ml | <=0.90 | male | 1989 | SP | SAO PAULO |
| 3 | a090625661c06e9c25ab67b16576ce23b1b0529 | 11/05/2020 | IgG, COVID19 | | 0.03 AU/ml | <=0.90 | female | 1975 | GO | NA |
| 4 | 0a67cd063da4bfade9be0e4fa0144ffc4f2d0b | 16/06/2020 | IgG, COVID19 | | 0.02 AU/ml | <=0.90 | female | 1997 | SP | SAO PAULO |
| 5 | b4d0a3ec53a085589a222d3c2f6b6ee02c7f7333 | 10/06/2020 | COVID IgM Interp | Não reagente | NULL | | male | 2006 | SP | SAO PAULO |
| 6 | d484f6c5985ce22f8dbbf59771df0d277bf05ec | 30/04/2020 | IgM, COVID19 | | 0.47 AU/ml | <=0.90 | female | 1983 | SP | SAO PAULO |
| 7 | 161f5d5b585c29048f523bd612c2090370f20bf8 | 01/05/2020 | IgM, COVID19 | | 0.9 AU/ml | <=0.90 | female | 1963 | SP | SAO PAULO |
| 8 | 95ecc1410a0f8abfde332e73da94a73514fa3ddc | 18/05/2020 | COVID IgM Interp | Não reagente | NULL | | female | 1988 | SP | SAO PAULO |
| 9 | ca24c419826fcc151f6fd0c75902971a75c28be7 | 20/04/2020 | IgG, COVID19 | | 30.77 AU/ml | <=0.90 | male | 1971 | SP | SAO PAULO |
| 10 | 5f3646074f312916e8cac683d3f5ba780a1d0c57 | 12/05/2020 | IgM, COVID19 | | 0.41 AU/ml | <=0.90 | female | 1968 | SP | SAO PAULO |
| 11 | a9afeda6994b949c3396f39339c13278d0487769 | 06/05/2020 | IgM, COVID19 | | 0.54 AU/ml | <=0.90 | male | 1976 | SP | SOROCABA |
| 12 | d9b08e2ab05dff544998a6b6bfe2880c9ddc3be7 | 05/06/2020 | IgM, COVID19 | | 0.65 AU/ml | <=0.90 | male | 1991 | SP | JUNDIAI |

Constructing the Import Command

- Variable names in first line
 - ▶ `col_names` = can be omitted
- No need to use `skip` =
- NA argument
 - ▶ `result` and `unit` variables use “Não reagente” and “NULL”
 - ▶ `birth_yr` has 3 NA values
 - ▶ Unclear if they mean 0 or could not get a meaningful result
 - ▶ Create argument: `na = c("Não reagente", "NULL", "NA")`

col_types = Argument

- pacid: awkward, but obviously a string
- dt_collect: Date, but in non-standard format
 - ▶ R will parse as character type
 - ▶ We can reformat it in tidying phase
- analysis: character string
- result: should be numeric, but has text entries
 - ▶ R will parse as character type
- unit & reference fields: character strings
 - ▶ reference could be numeric, but leave to tidying
- sex: character, but better as factor: only 2 values
- birth_yr: numeric
- uf & city: character
- Using codes, full argument: `col_types = "ccccccfncc"`

Ready to Import

```
einstein_soro <- read_csv(here::here("einstein_soro_tests.csv"),  
                          col_types = "ccccccfncc",  
                          na = c("Não reagente", "NULL", "NA"))  
glimpse(einstein_soro)
```

```
## Rows: 99  
## Columns: 10  
## $ pacid      <chr> "b6d668e4f818f7b3643ed593b8fb902bf9d2501e", "a090625661c...  
## $ dt_collect <chr> "28/05/2020", "11/05/2020", "16/06/2020", "10/06/2020", ...  
## $ analysis   <chr> "IgM, COVID19", "IgG, COVID19", "IgG, COVID19", "COVID I...  
## $ result     <chr> "0.74", "0.03", "0.02", "Não reagente", "0.47", "0.9", "...  
## $ unit       <chr> "AU/ml", "AU/ml", "AU/ml", NA, "AU/ml", "AU/ml", NA, "AU...  
## $ reference  <chr> "<=0.90", "<=0.90", "<=0.90", "", "<=0.90", "<=0.90", "...  
## $ sex       <fct> male, female, female, male, female, female, female, male...  
## $ birth_yr   <dbl> 1989, 1975, 1997, 2006, 1983, 1963, 1988, 1971, 1968, 19...  
## $ uf        <chr> "SP", "GO", "SP", "SP", "SP", "SP", "SP", "SP", "SP", "S...  
## $ city      <chr> "SAO PAULO", NA, "SAO PAULO", "SAO PAULO", "SAO PAULO", ...
```

Section 12

Working with Excel Files

Advice on Working *within* Excel - 1

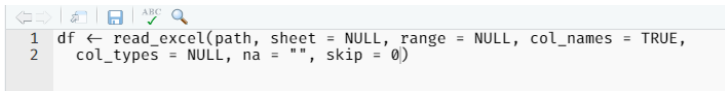
- If you are preparing .csv files,
 - ▶ Save different types of data in different files
- If you are working with .x/sx files,
 - ▶ Save different types of data in different sheets (tabs)
- Data should be placed in a block of data
 - ▶ Without any blank lines
- First line should have variable names, if you are using them
- Each column should contain **only** one class of data
 - ▶ Numeric, character, logical, ...

Advice on Working *within* Excel - 2

- Zeros are always “0”
 - ▶ Never “-”, “ ” (blank space) or other character
- Missing data should always be “NA”
 - ▶ Never “0”, “99” or other text
- Start making the data conform to the tidy data rules here
 - ▶ Each column should be a **variable**
 - ▶ Each line should be a **case**
- **Never** use colors or design elements

Importing Excel Files

- Do not have to save files as “.csv” first
- readxl package
 - ▶ read_excel()
 - ▶ Works with both .xls and .xlsx formats
- Differs from read_csv() primarily in handling of column types
- Also has arguments for
 - ▶ Specifying tabs within workbook
 - ▶ Specifying specific region on a given sheet

A screenshot of the RStudio interface. The top toolbar shows icons for undo, redo, save, and search. Below the toolbar, a code editor displays two lines of R code. The first line is 'df <- read_excel(path, sheet = NULL, range = NULL, col_names = TRUE,' and the second line is 'col_types = NULL, na = "", skip = 0)'. The code is highlighted in a light blue background.

```
1 df <- read_excel(path, sheet = NULL, range = NULL, col_names = TRUE,  
2   col_types = NULL, na = "", skip = 0)
```

Arguments for `read_excel()`

- Specifying what to download
 - ▶ Path for file
 - ▶ Sheet name
 - ▶ range in normal Excel format: "A1:B25"
- `col_names` = functions same as `read_csv()`
- `na` = and `skip` = function the same

Column Types in `read_excel()`

- Uses whole words instead of single letters
- If you want to make a factor, need to do so in tidying process
- Column types:
 - ▶ date
 - ▶ guess (trust R to make a good choice)
 - ▶ list
 - ▶ logical
 - ▶ numeric
 - ▶ skip
 - ▶ text

Spreadsheet Example

- R standard data set: Palmer Penguins
 - ▶ 8 characteristics of 344 penguins
 - ▶ `penguins.xlsx`

| | | | | | | | | |
|-----|---------|-----------|----------------|---------------|-------------------|-------------|--------|------|
| L20 | | | | | | | | |
| | A | B | C | D | E | F | G | H |
| 1 | species | island | bill_length_mm | bill_depth_mm | flipper_length_mm | body_mass_g | sex | year |
| 2 | Adelie | Torgersen | 39.1 | 18.7 | 181 | 3750 | male | 2007 |
| 3 | Adelie | Torgersen | 39.5 | 17.4 | 186 | 3800 | female | 2007 |
| 4 | Adelie | Torgersen | 40.3 | 18 | 195 | 3250 | female | 2007 |
| 5 | Adelie | Torgersen | NA | NA | NA | NA | NA | 2007 |
| 6 | Adelie | Torgersen | 36.7 | 19.3 | 193 | 3450 | female | 2007 |
| 7 | Adelie | Torgersen | 39.3 | 20.6 | 190 | 3650 | male | 2007 |
| 8 | Adelie | Torgersen | 38.9 | 17.8 | 181 | 3625 | female | 2007 |
| 9 | Adelie | Torgersen | 39.2 | 19.6 | 195 | 4675 | male | 2007 |
| 10 | Adelie | Torgersen | 34.1 | 18.1 | 193 | 3475 | NA | 2007 |
| 11 | Adelie | Torgersen | 42 | 20.2 | 190 | 4250 | NA | 2007 |
| 12 | Adelie | Torgersen | 37.8 | 17.1 | 186 | 3300 | NA | 2007 |
| 13 | Adelie | Torgersen | 37.8 | 17.3 | 180 | 3700 | NA | 2007 |

Example Arguments

- Column types

- ▶ Argument is unnecessary as R will parse values correctly
- ▶ If we want to specify it
 - ★ `col_types = c("text", "text", "numeric", "numeric", "numeric", "numeric", "text", "numeric")`
- ▶ NA needed for measurement columns since missings exist
 - ★ `na = "NA"`

Import File

```
penguin <- readxl::read_excel(here::here("penguins.xlsx"), na = "NA")
glimpse(penguin)
```

```
## Rows: 344
## Columns: 8
## $ species      <chr> "Adelie", "Adelie", "Adelie", "Adelie", "Adelie",...
## $ island       <chr> "Torgersen", "Torgersen", "Torgersen", "Torgersen...
## $ bill_length_mm <dbl> 39.1, 39.5, 40.3, NA, 36.7, 39.3, 38.9, 39.2, 34....
## $ bill_depth_mm <dbl> 18.7, 17.4, 18.0, NA, 19.3, 20.6, 17.8, 19.6, 18....
## $ flipper_length_mm <dbl> 181, 186, 195, NA, 193, 190, 181, 195, 193, 190, ...
## $ body_mass_g   <dbl> 3750, 3800, 3250, NA, 3450, 3650, 3625, 4675, 347...
## $ sex           <chr> "male", "female", "female", NA, "female", "male",...
## $ year          <dbl> 2007, 2007, 2007, 2007, 2007, 2007, 2007, 2007, 2...
```

Section 13

FASTA Files

Read a FASTA File

- Most common storage format for nucleotide (DNA or RNA) and amino acid sequences
- Text-based representation using single-letter codes (alphabets)
- Each sequence can be preceded by a line of sequence name and comments
- Like a .csv file: pure text (uses text editor)
- To prepare sequences for analysis
 - ▶ Use specialized programs for reading and parsing
- New package (bioseq) provides easy and complete functions
 - ▶ Not in original set of downloaded packages
 - ▶ Install on your machine with `install.packages("bioseq")` in Console
- `bioseq::read_fasta()` takes 2 arguments
 - ▶ file name and path
 - ▶ Type of sequence: "DNA" (default), "RNA" or "AA"

FASTA File Example

- DNA sequences of the *gag* polyprotein of the HIV-1 virus
 - ▶ Reference genome HXB-2
- Sequence stored in file HIVHXB.fa
- Sequence is 1,503 base pairs long.

```
HIVHXB.fa x
1 >HIVHXB2CG (790 .. 2292) (1503 bp)
2 ATGGGTGCGAGAGCGTCAGTATTAAGCGGGGGAGAATTAGATCGATGGGAAAAAATTCGGTTAAGGCCAGGGGGAAAGAA
3 AAAATATAAATTAACATATAGTATGGGCAAGCAGGGAGCTAGAACGATTCGCAGTTAATCTGGCCTGTAGAAACAT
4 CAGAAGGCTGTAGACAAATACTGGGACAGCTACAACCATCCCTTCAGACAGGATCAGAAGAAGCTTAGATCATTATATAAT
5 ACAGTAGCAACCTCTATTGTGTCATCAAGGATAGAGATAAAAGACACCAAGGAAGCTTTAGACAAGATAGAGGAAGA
6 GCAAAACAAAAGTAAGAAAAAGCAGCAAGCAGCAGCTGACACAGGACACAGCAATCAGGTCAGCCAAAATTACCTTA
7 TAGTGCAGAACATCCAGGGGCAATGGTACATCAGGCCATATCACTAGAACTTTAAATGCATGGGTAAAAGTAGTAGAA
8 GAGAAGGCTTTCAGCCAGAAAGTGATACCCATGTTTTTCAGCATTATCAGAAGGAGGCCACCCACAAGATTTAAACACCAT
9 GCTAAACACAGTGGGGGGACATCAAGCAGCCATGCAATGTAAAAGAGACCATCAATGAGGAAGCTGCAGAAATGGGATA
10 GAGTGCATCCAGTGCATGCAGGGCCTATTGCACAGGCCAGATGAGAGAACCAAGGGGAAGTGACATAGCAGGAACTACT
11 AGTACCCTTCAGGAACAAATAGGATGGATGACAAATAATCCACCTATCCAGTAGGAGAAATTTATAAAAGATGGATAAT
12 CCTGGGATTAATAAAATAGTAAGAAATGTATAGCCCTACCAGCATTCTGGACATAAGACAAGGACCAAGGAACCCCTTA
13 GAGACTATGTAGACCGGTTCTATAAACTCTAAGAGCCGAGCAAGCTTCACAGGAGGTAAAAAATTGGATGACAGAAACC
14 TTGTTGGTCCAAAATGCGAACCCAGATTGTAAGACTATTTTAAAGCATTGGGACCAGCGGCTACACTAGAAGAAATGAT
15 GACAGCATGTCAAGGAGTAGGAGGACCCGGCCATAAGGCAAGAGTTTGGCTGAAGCAATGAGCCAGTAACAAATTCAG
16 CTACCATAATGATGCAGAGAGGCAATTTTAGGAACCAAAGAAAGATTGTTAAGTGTTCATTGTGGCAAGGAAGGGCAC
17 ACAGCCAGAAATTCAGGGGCCCTAGGAAAAAGGGCTGTTGGAATGTGGAAGGAGGACCAACAAATGAAGATTGTAC
18 TGAGAGACAGGCTAATTTTTTAGGGAAGATCTGGCCTTCCTACAAGGGAAGGCCAGGGAATTTCTTCAGAGCAGACGAC
19 AGCCACAGCCCCACCAAGAGAGAGCTTCAGGTCTGGGGTAGAGACAACAACTCCCTCCAGGAAGCAGGAGCCGATAGAC
20 AAGGAAGCTGTATCCCTTAACCTTCCTCAGGTCACTCTTTGGCAACGACCCCTCGTCACATAAA
```

Import the File

```
seq <- bioseq::read_fasta(here::here("HIVHXB.fa"), "DNA")
stringr::str_sub(seq, 1, 60)
```

```
## [1] "ATGGGTGCGAGAGCGTCAGTATTAAGCGGGGAGAATTAGATCGATGGGAAAAAATTCGG"
```

- Result a character string vector
- First 60 nucleotides using the subsetting function of stringr package.

Section 14

[Go to Presentation Class_2b](#)