Week 9 Cohort 4: R4DS Book Club

Chapters 8, 9, & 10

Workflow: projects; Tibbles

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5-minute ice breaker

• What types of data do you work with?

Quick housekeeping/reminders

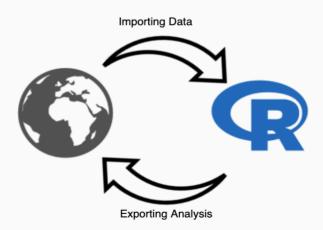
- Video camera is optional, but encouraged.
- If we need to slow down and discuss, let me know.
 - Most likely someone has the same question.
- Take time to learn the theory.
- Please attempt the chapter exercises.
- Please plan on teaching one of the lessons.

Tonight's discussion

- Chapter 8 Workflow: projects
- Chapter 9 Data wrangling introduction
- Chapter 10 Tibbles

Why care about a project-oriented workflow?

"One day you will need to bring data from the outside world into R and send numerical results and figures from R back into the world. ~Hadley Wickham & Garrett Grolemund (authors)"



- Data
- Plot(s)
- Model(s)
- Our code

Why care about a project-oriented workflow (cont.)?

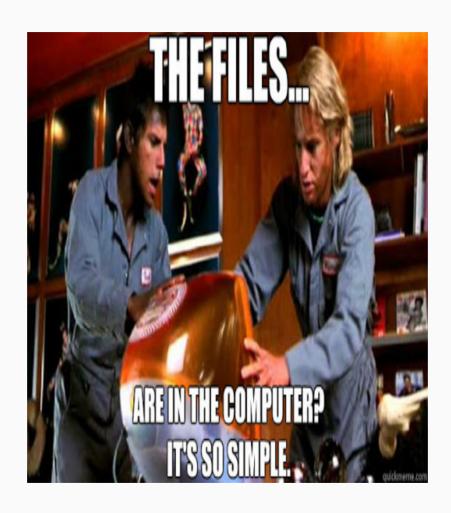
- Maximize effectiveness
- Reduce frustration
- Jenny Bryan on 'Project-oriented workflows'
- Sharla Gelfand on repeated reporting
 - Some of the concepts covered will be discussed in later chapters

Consider: What is real from our analysis?

"With your R scripts (and your data files), you can recreate the environment. It's much harder to recreate your R scripts from your environment! ~Hadley Wickham & Garrett Grolemund (authors)"

- Create well-commented, self-contained code and projects to aid replication.
- To facilitate this:
 - Change settings to NOT save your workspace.
 - Restart RStudio and rerun your script to check if everything is present
 - o Cmd/Ctrl + Shift + F10 Restarts R
 - Cmd/Ctrl + Shift + S Rerun the current script

Where does our analysis take place?



Okay, it's a little more complex.

- Paths and directories (AKA the location of your files).
- Three differences across operating systems (Linux/Mac vs. Windows)
 - 1. Slashes (e.g., /, Mac & Linux) vs. backslashes (e.g. \\, Windows).
 - Backslashes are special (e.g., \\), so use slashes
 - 2. Absolute paths, avoid them
 - /users/cberke/project-name/data 👎 VS /project-name/data 👍
 - Other users will have different directories.
 - Use relative paths.
 - 3. The ~ (AKA the home directory)
 - Windows points to Documents

Avoid setwd()

 You *can* set your working directory by doing something like this: setwd("/path/to/project/files")

- I will come into your office and SET YOUR COMPUTER ON FIRE () ~Jenny Bryan
 - Path's wont work for another user or on another computer.
 - Assumes all projects are done in one R process.
 - Avoid project leakage 🍐.

Rstudio projects 🎉

- Demo Rstudio project set up.
- Create notebook.
 - Highlight inserting a code chunk

Your turn 🗎

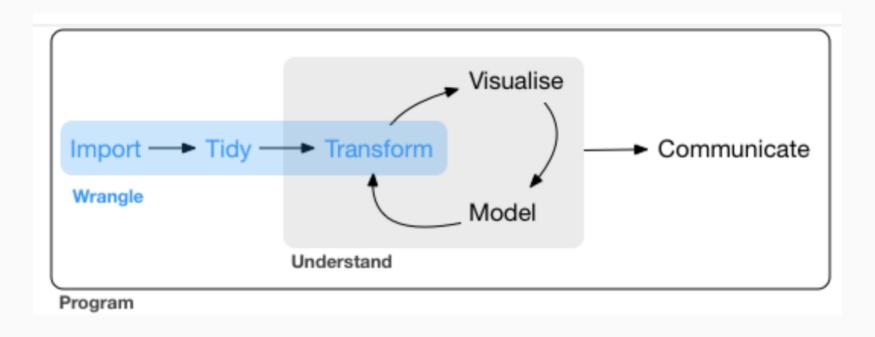
- 1. Open Rstudio and run getwd() (Close a project if you have one open)
 - What file path is returned?
- 2. Create a Rstudio project
- 3. Run getwd() again
 - What file path is returned?
- 4. Create a Rnotebook
- 5. Enter and run the following in a code chunk:

```
library(tidyverse)

# **This will write a file to your computer**
write_csv(mtcars, "mtcars.csv")
```

6. Can you find the file using your file explorer?

Chapter 9: Intro to data wrangling



Data science workflow ~Hadley Wickham & Garrett Grolemund (authors)

What does this section of the book cover?

• Data wrangling:

- Tibbles
- Data import
- Tidy data

• Data transformation:

- Relational data
- Strings
- Factors
- Dates and times

Chapter 10: Tibbles

- What is a tibble?
 - We've already worked with them!
 - A data structure adapted from Base R's data.frame.
- How are tibbles different from a data.frame?
 - Tibbles are data frames, but they tweak some older behaviors to make life a little easier. ~Hadley Wickham & Garrett Grolemund (authors)
 - tibbles provide stricter checking and better formatting than the traditional data frame. ~Hadley Wickham & Garrett Grolemund (authors)
 - The general ethos is that tibbles are lazy and surly: they do less and complain more than the base data.frame S. ~ tbl_df-class documentation.
 - Tibbles are opinionated

Let's observe these properties

• Example 1:

```
mtcars # A data.frame
class(mtcars) # Check the class
tbl_sum(mtcars)
```

• Example 2:

```
mtcars_tbl ← as_tibble(mtcars) # using as_tibble() to create a tibble
mtcars_tbl
class(mtcars)
tbl_sum(mtcars)
```

• What differences do you see?

Creating tibbles

• Have a data.frame? Use as_tibble().

```
(iris ← as_tibble(iris))
```

• Creating a tibble from vectors.

```
# Notice the recycling rules of inputs of length 1
tibble(
    x = 1:5,
    y = 1,
    z = x ^ 2 + y
)
```

• Use tribble() for data entry in code

```
tribble(
    ~x, ~y, ~z,
    #--|--|----
    "a", 2, 3.6,
    "b", 1, 8.5
)
```

When creating/using tibbles, know that...

- tibble() never changes the type of the inputs
- tibble() never changes the names of variables
- tibble() never creates row names
- tibble() allows the use of non-syntactic names

```
(tb ← tibble(
    `:)` = "smile",
    ` = "space",
    `2000` = "number"
))
```

```
## # A tibble: 1 x 3
## `:)` ` `2000`
## <chr> <chr> ## 1 smile space number
```

• tibbles are at the core of the tidyverse, so get comfortable using them

Tibbles vs. data.frame

1. Printing

- Tibbles have a refined print method.
- Rows and columns fit the screen.
- Data types are reported.
- You have flexibility to change the output, package?tibble

2. Subsetting

New tools: \$ and [[

Tibble subsetting

```
# Create a tibble
tibble_df ← tibble(
  abc = runif(5),
 xyz = rnorm(5)
# Create a data.frame
df \leftarrow data.frame(
 abc = runif(5),
 xyz = rnorm(5)
```

Tibble subsetting (cont.)

```
tibble_df$<u>abc</u>
## [1] 0.79014970 0.21065367 0.13461285 0.57002023 0.06602784
df$abc
## [1] 0.96659906 0.72045568 0.16218410 0.50333947 0.02519313
tibble df$a
## Warning: Unknown or uninitialised column: `a`.
## NULL
df$a
## [1] 0.96659906 0.72045568 0.16218410 0.50333947 0.02519313
```

Tibble subsetting (cont.)

```
# Using [[]], by position and name
tibble_df[[1]]
```

[1] 0.79014970 0.21065367 0.13461285 0.57002023 0.06602784

```
tibble_df[["abc"]]
```

[1] 0.79014970 0.21065367 0.13461285 0.57002023 0.06602784

df[[1]]

[1] 0.96659906 0.72045568 0.16218410 0.50333947 0.02519313

df[["abc"]]

[1] 0.96659906 0.72045568 0.16218410 0.50333947 0.02519313

Tibble subsetting (cont.)

```
# Using the pipe, you need the `.` place holder
tibble_df %>% .$abc
```

[1] 0.79014970 0.21065367 0.13461285 0.57002023 0.06602784

```
tibble_df %>% .[["xyz"]]
```

[1] -0.44601287 0.02497698 0.19212898 0.37532214 -0.83882670

Converting back to a data.frame

- Older functions may not work well with tibbles.
 - o Use as.data.frame()

```
class(as.data.frame(tibble_df))
```

```
## [1] "data.frame"
```

Questions/Discussion