

# Welcome!

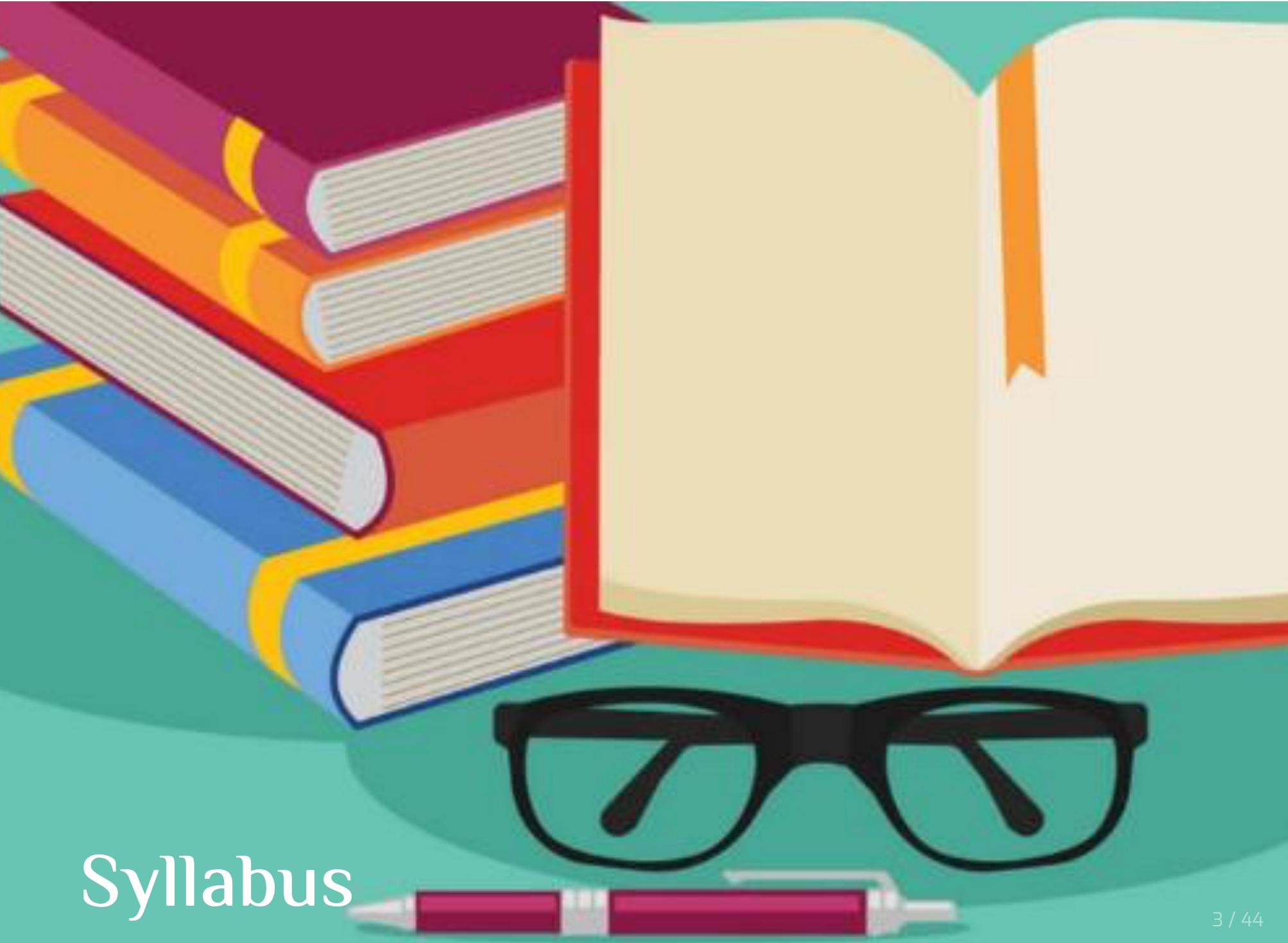
## An overview of the course

*Daniel Anderson  
Week 1, Class 1*



# Agenda

- Getting on the same page
- Syllabus
- Intro to data types



# Syllabus

# Course Website(s)

website repo

EDUC 610    Slides, Readings, & Schedule    Assignments

Syllabus    DataCamp    

## Functional Programming with R

This is the third course in the *data science in educational research* specialization. This course builds upon the content covered in the first two courses, with a specific focus on becoming a better programmer and improving workflows in statistical computing with R. Students will continue to work with the `tidyverse` suite of packages, with a emphasis on `{purrr}` for functional programming. At its core, functional programming is a technique to iterate a function over a vector, or set of vectors, to complete repetitive tasks. We will compare and contrast `{purrr}` functions with base R approaches, including `for` loops and the `apply` family of functions. Functional programming helps reduce redundancies in code, making it more efficient and, often, more readable. The course will also cover writing custom functions, which can also help in completing repetitive tasks, but can also extend the functionality of R, and is a key component of functional programming. The course concludes with a brief introduction to `shiny` for building interactive applications which, although somewhat outside of the scope of functional programming, requires using and writing functions.



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- Write effective and clear functions to continue with the mantra of "Don't Repeat Yourself"

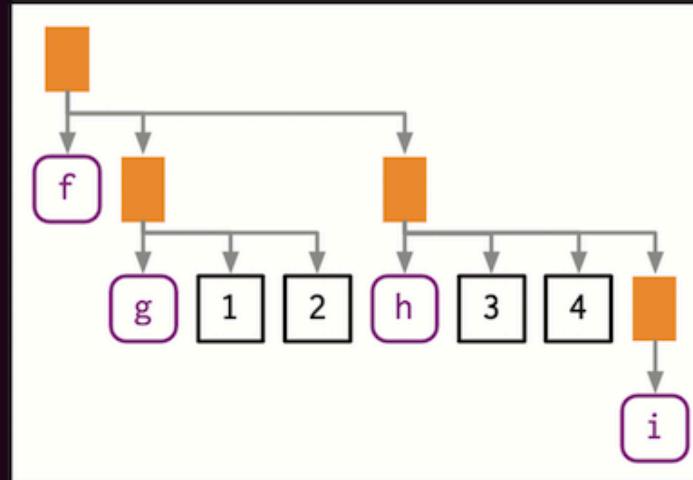
# This Week's learning objectives

- Understand the requirements of the course
- Understand the requirements of the final project

The R Series

# Advanced R

Second Edition



Hadley Wickham

 CRC Press  
Taylor & Francis Group  
A CHAPMAN & HALL BOOK

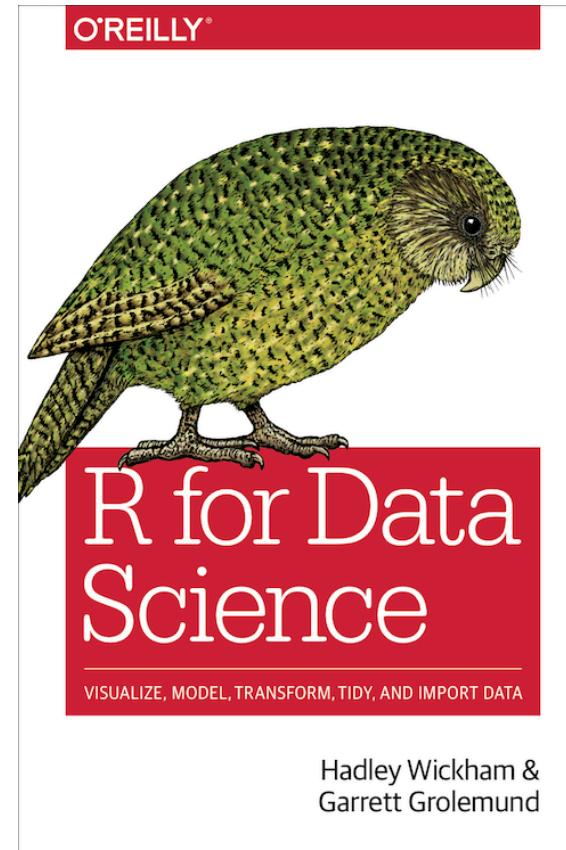
# Required Textbook

# Other books (also free)

Bryan (<http://happygitwithr.com>)



Wickham & Grolemund  
(<https://r4ds.had.co.nz>)



# Structure of the course

- First 5 weeks - mostly iteration
  - Data types
  - Base R iterations
  - {purrr}
  - Batching processes and working with list columns
  - Parallel iterations (and a few extras)

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- First 5 weeks - mostly iteration
  - Data types
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  - Parallel iterations (and a few extras)
- Second 5 weeks - mostly writing functions
  - Writing functions 1
  - Writing functions 2
  - Shiny 1
  - Shiny 2
  - Code performance & packages (briefly)

# Labs

16%

4 @ 10 points each

Two labs during each 5 week segment

- Iteration w/Base R
  - Mostly `for` loops
- Iteration w/{purrr}
  - Mostly `map_*`
- Writing functions 1
  - Basic functions
- Writing functions 2
  - Making functions "pure"
  - Building up functionality (one function calling another)

# DataCamp

DataCamp  
*One required module (10 points)*

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*One module available for extra credit (5 points)*

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*Split by 5 week sections*

Must complete one (due dates differ by section)

If you want the extra credit, one must come from each section

# First 5 week DataCamp courses

*Due by 11:59 PM 5/6/19*

- Intermediate R
- Foundations of functional programming with purrr
- Intermediate functional programming with purrr
- Machine learning in the tidyverse

# Second 5 week DataCamp courses

*Due by 11:59 PM 6/12/19*

- Writing functions in R
- Building web applications in R with Shiny
- Building web applications in R with Shiny: Case studies
- Developing R packages
- Writing efficient R code

# Midterm

*50 points (20%)*

- Final format still yet to be determined - happy to take input on what it should look like

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- Will include you writing new code
- May include you
  - Identifying bugs in code
  - Automating an existing approach
  - Multiple choice/fill-in the blank questions

# Final Project

*150 points total (60%)*

5 parts

Component	Points	Due
Groups Finalized	0	4/10/19
Outline	10	5/1/19
Draft	20	5/27/19
Peer review	20	6/3/19
Product	100	6/12/19 (11:59 pm)

# What is it?

*Two basic options*

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*Two basic options*

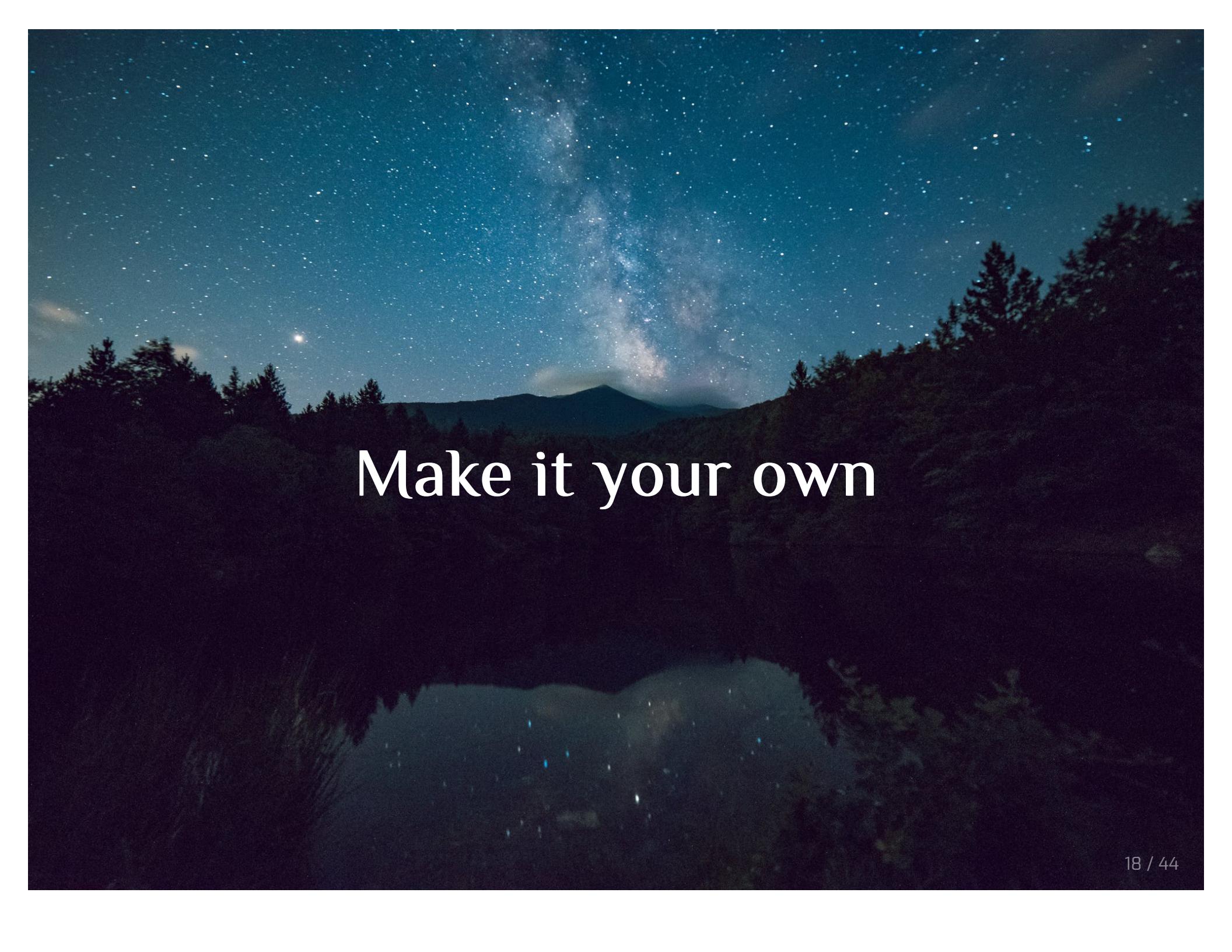
## Data Product

Similar to first class

- Brief research manuscript (can be APA or not, I don't really care 🤷‍♀️)
- Shiny app
- Dashboard
  - Probably unlikely to work well though, unless you make it a shiny dashboard
- Blog post
- For the ambitious - a documented R package

# Tutorial

- Probably best done through a blog post or series of blog posts
- Approach as if you're teaching others about the content I'll ask you to cover
- BONUS: You can actually release the blog post(s) and may get some traffic 



Make it your own

# What you have to have

- Everything on GitHub
- Publicly available dataset
- Team of 2-4

# What you have to cover

Unfortunately, this still is a class assignment. I have to be able to evaluate that you can actually apply the content within a messy, real-world setting.

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Unfortunately, this still is a class assignment. I have to be able to evaluate that you can actually apply the content within a messy, real-world setting.

The grading criteria (which follow) may force you into some use cases that are a bit artificial. This is okay.

# Grading criteria

- No code is used repetitively (no more than twice) **10 points**

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- At least one instance of parallel iteration (e.g., `map2_*`, `pmap_*`)  
**10 points**
- At least one use case of `purrr::nest %>% mutate()` **10 points**

- At least two custom functions **20 points; 10 points each**
  - Each function must be "pure" unless it is clear that it should have side effects (e.g., a plotting function)
  - Each function must do exactly one thing
  - The functions **may** replicate the behavior of a base function - as noted above this is about practicing the skills you learn in class

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- Code is fully reproducible and housed on GitHub **10 points**
- No obvious errors in chosen output format **10 points**
- Deployed on the web and shareable through a link **10 points**

# Outline

*Due 5/1/19*

Four components:

- Description of data source (must be publicly available)
- Purpose (tutorial or substantive)
- Chosen format
- Lingering questions
  - How can I help?

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- Description of data source (must be publicly available)
- Purpose (tutorial or substantive)
- Chosen format
- Lingering questions
  - How can I help?

Please include all components - including the question(s) section!

# Draft

*Due 5/27/19, before class*

- Expected to still be a work in progress
  - This means some of your code may be rough and/or incomplete. However:
- Direction should be obvious
- Most, if not all, grading elements should be present
- Provided to your peers so they can learn from you as much as you can learn from their feedback

# Peer Review

- Exact same process we've used before
- It has always gone very well - let's keep it going
- If, during your peer review, you find grading elements not present, definitely note them

# Utilizing GitHub (required)

- You'll be assigned two groups to review (10 points each)
- Fork their repo
- Embed comments, suggest changes to their code
- Submit a PR
  - Summarize your overall review in the PR
- I WILL cover this this term, barring a tornado or similar

# Grading

250 points total

- 4 labs at 10 points each (40 points; 16%)
- 1 DataCamp modules (10 points; 4%)
- Midterm (50 points; 20%)
- Final Project (150 points; 60%)
  - Proposal (10 points; 4%)
  - Draft (20 points; 8%)
  - Peer review (20 points; 8%)
  - Product (100 points; 40%)

# Grading

Lower percent	Lower point range	Grade	Upper point range	Upper percent
0.97	(242 pts)	A+	(250 pts)	
0.93	(232 pts)	A	(242 pts)	0.97
0.90	(225 pts)	A-	(232 pts)	0.93
0.87	(218 pts)	B+	(225 pts)	0.90
0.83	(208 pts)	B	(218 pts)	0.87
0.80	(200 pts)	B-	(208 pts)	0.83
0.77	(192 pts)	C+	(200 pts)	0.80
0.73	(182 pts)	C	(192 pts)	0.77
0.70	(175 pts)	C-	(182 pts)	0.73
		F	(175 pts)	0.70



# COMMUNICATION

# Missing class



I will be out Monday next week

- Trying to get a guest speaker
- May end up having to adjust the schedule a bit

Any time left?

Any time left?  
Basic data types

# Vectors

## *Pop quiz*

Discuss with your neighbor

- What are the four basic types of atomic vectors?
- What function creates a vector?
- T/F: A list (an R list) is not a vector.
- What is the fundamental difference between a matrix and a data frame?
- What does *coercion* mean, and when does it come into play?

# Vector types

*4 basic types*

Note there are two others (complex and raw), but we don't care about them (I've never even seen them used).

- Integer
- Double
- Logical
- Character

# Vector types

## *4 basic types*

Note there are two others (complex and raw), but we don't care about them (I've never even seen them used).

- Integer
- Double
- Logical
- Character

Integer and double vectors are both numeric.

# Creating vectors

Vectors are created with `c`. Below are examples of each of the four main types of vectors.

```
integer <- c(5L, 7L, 3L, 94L) # L explicitly an integer, not double
double <- c(3.27, 8.41, Inf, -Inf)
logical <- c(TRUE, TRUE, TRUE, FALSE, TRUE, FALSE, FALSE)
character <- c("red", "orange", "yellow", "green", "blue", "violet", "rainbo
```

# Coercion

- Vectors **must** be of the same type.
- If you try to mix types, implicit coercion will occur
- Implicit coercion defaults to the most flexible type
  - which is... ?

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  - which is... ?

```
c(7L, 3.25)
```

```
## [1] 7.00 3.25
```

```
c(3.24, TRUE, "April")
```

```
## [1] "3.24"  "TRUE"  "April"
```

```
c(TRUE, 5)
```

```
## [1] 1 5
```

# Explicit coercion

- You can alternatively define the coercion to occur

```
as.integer(c(7L, 3.25))
```

```
## [1] 7 3
```

```
as.logical(c(3.24, TRUE, "April"))
```

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

```
as.character(c(TRUE, 5))
```

```
## [1] "1" "5"
```

# Checking types

- Use `typeof` to verify the type of vector

```
typeof(c(7L, 3.25))
```

```
## [1] "double"
```

```
typeof(as.integer(c(7L, 3.25)))
```

```
## [1] "integer"
```

# Piping

- Although traditionally used within the tidyverse (not what we're doing here), it can still be useful. The following are equivalent

```
library(magrittr)

typeof(as.integer(c(7L, 3.25)))

## [1] "integer"

c(7L, 3.25) %>%
  as.integer() %>%
  typeof()

## [1] "integer"
```

# Pop quiz

Without actually running the code, talk with your neighbor: Which will each of the following coerce to?

```
c(1.25, TRUE, 4L)  
c(1L, FALSE)  
c(7L, 6.23, "eight")  
c(TRUE, 1L, 0L, "False")
```

# Answers

```
typeof(c(1.25, TRUE, 4L))
```

```
## [1] "double"
```

```
typeof(c(1L, FALSE))
```

```
## [1] "integer"
```

```
typeof(c(7L, 6.23, "eight"))
```

```
## [1] "character"
```

```
typeof(c(TRUE, 1L, 0L, "False"))
```

```
## [1] "character"
```

# Missing values

- Missing values breed missing values

```
NA > 5
```

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

```
NA * 7
```

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

# Missing values

- Missing values breed missing values

```
NA > 5
```

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

```
NA * 7
```

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

- But this one may surprise you

```
NA == NA
```

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

# Missing values

- Missing values breed missing values

```
NA > 5
```

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

```
NA * 7
```

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

- But this one may surprise you

```
NA == NA
```

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

It is correct, however, because there's no reason to presume that one missing value is equal to another missing value.

# When missing values don't propagate

```
NA & FALSE
```

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

```
NA | TRUE
```

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

```
x <- c(NA, 3, NA, 5)  
any(x > 4)
```

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

# How to test missingness?

- We've already seen the following doesn't work

```
x == NA
```

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

# How to test missingness?

- We've already seen the following doesn't work

```
x == NA
```

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

- Instead, use `is.na`

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

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

# Next time

- More on data types
  - Subsetting
  - Attributes
  - More on coercion (good to be fluent)
- Lists

*Any lingering questions?*