

Week 1: *Getting Started*

🏛️ EMSE 4572/6572: Exploratory Data Analysis

👤 John Paul Helveston

📅 August 27, 2025

Week 1: *Getting Started*

1. Course Goal
2. Course Introduction
3. Break: Install Stuff
4. Quarto
5. Workflow & Reading In Data
6. Wrangling Data
7. Visualizing Data

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Course 1: Intro to Programming for Analytics

"Computational Literacy"

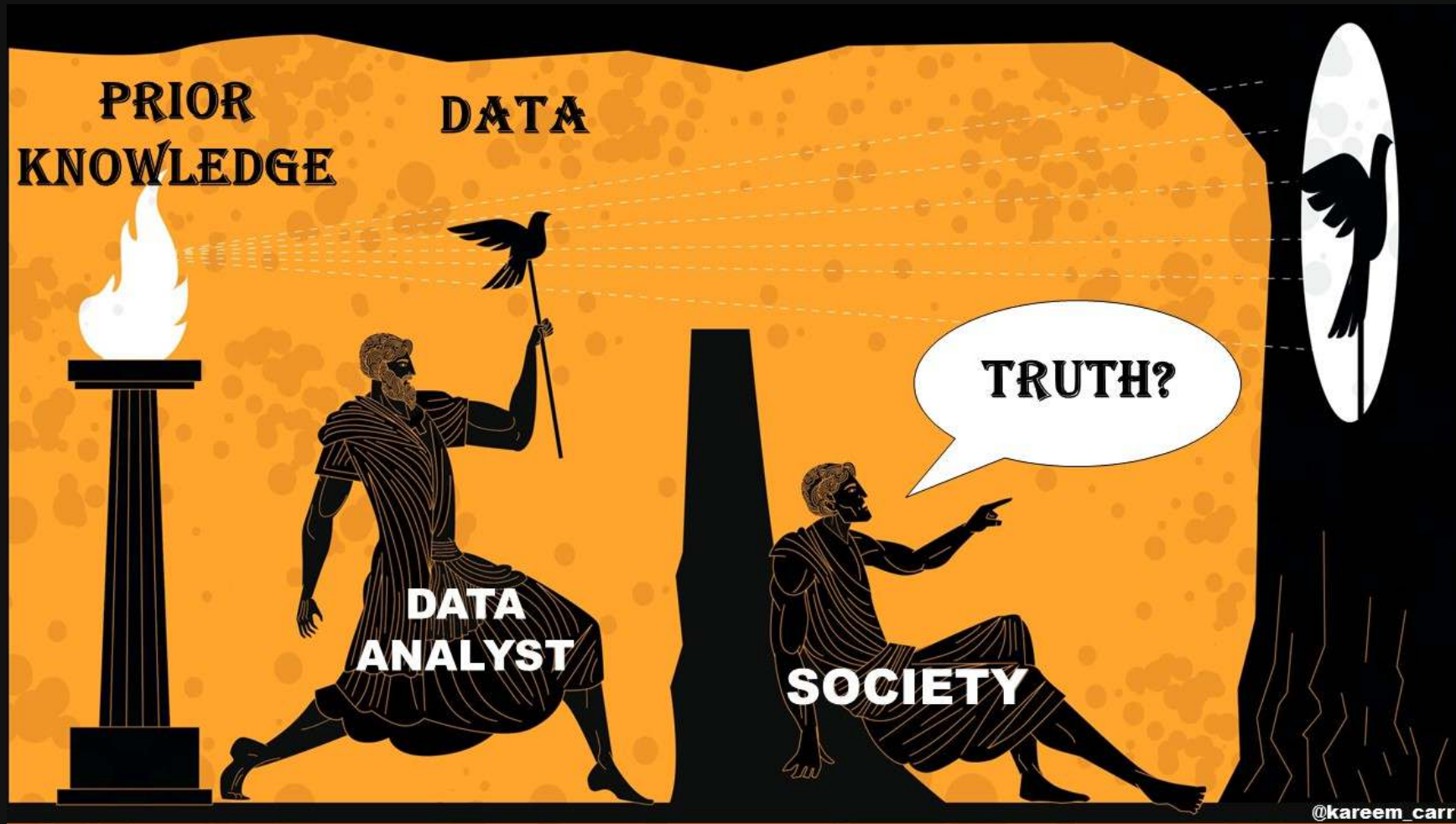
- Programming: Conditionals (if/else), loops, functions, testing, data types.
- Analytics: Data structures, import / export, basic data manipulation & visualization.

Course 2: Exploratory Data Analysis

"Data Literacy"

- Strategies for conducting an exploratory data analysis.
- Design principles for visualizing and communicating *information* extracted from data.
- Reproducibility: Reports that contain code, equations, visualizations, and narrative text.

Class goal: translate *data* into *information*



Class goal: translate *data* into *information*

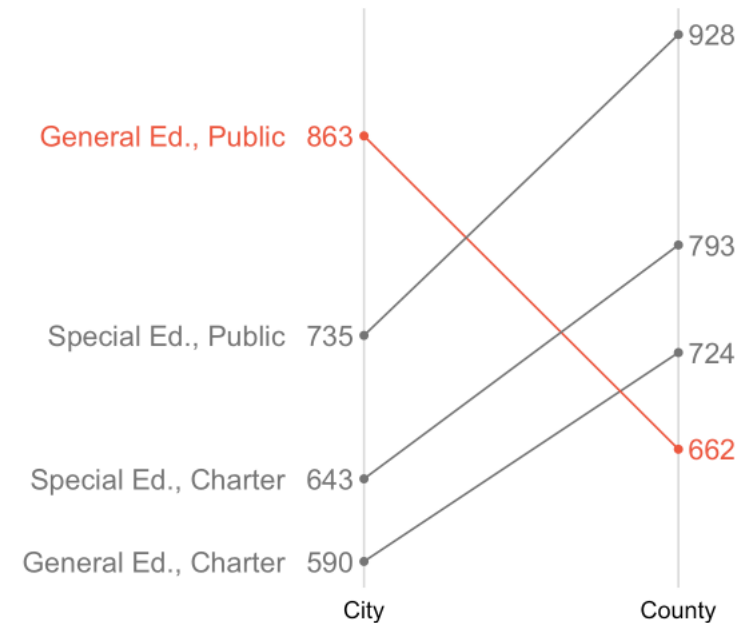
Data

Average student engagement scores

Class	Type	City	County
Special Ed.	Charter	643	793
Special Ed.	Public	735	928
General Ed.	Charter	590	724
General Ed.	Public	863	662

Information

Students in public, general education classes in county schools have surprisingly low engagement



Data exploration: an iterative process

Encode data:

```
engagement_data <- data.frame(  
  City    = c(643, 735, 590, 863),  
  County  = c(793, 928, 724, 662),  
  School  = c('Special Ed., Charter', 'Special Ed., Pu  
              'General Ed., Charter', 'General Ed., Pu  
engagement_data
```

```
#>   City County      School  
#> 1  643   793 Special Ed., Charter  
#> 2  735   928 Special Ed., Public  
#> 3  590   724 General Ed., Charter  
#> 4  863   662 General Ed., Public
```

Re-format data for plotting:

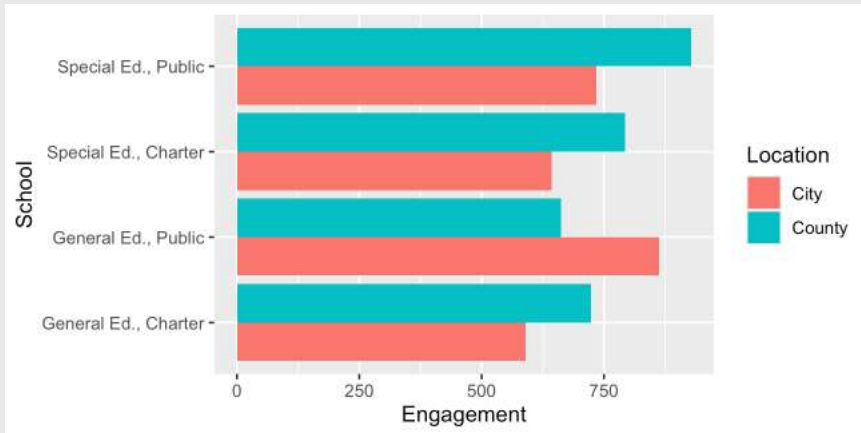
```
engagement_data <- engagement_data %>%  
  gather(Location, Engagement, City:County) %>%  
  mutate(Location = fct_relevel(  
    Location, c('City', 'County')))  
engagement_data
```

```
#>      School Location Engagement  
#> 1 Special Ed., Charter      City      643  
#> 2 Special Ed., Public      City      735  
#> 3 General Ed., Charter      City      590  
#> 4 General Ed., Public      City      863  
#> 5 Special Ed., Charter     County      793  
#> 6 Special Ed., Public     County      928  
#> 7 General Ed., Charter     County      724  
#> 8 General Ed., Public     County      662
```

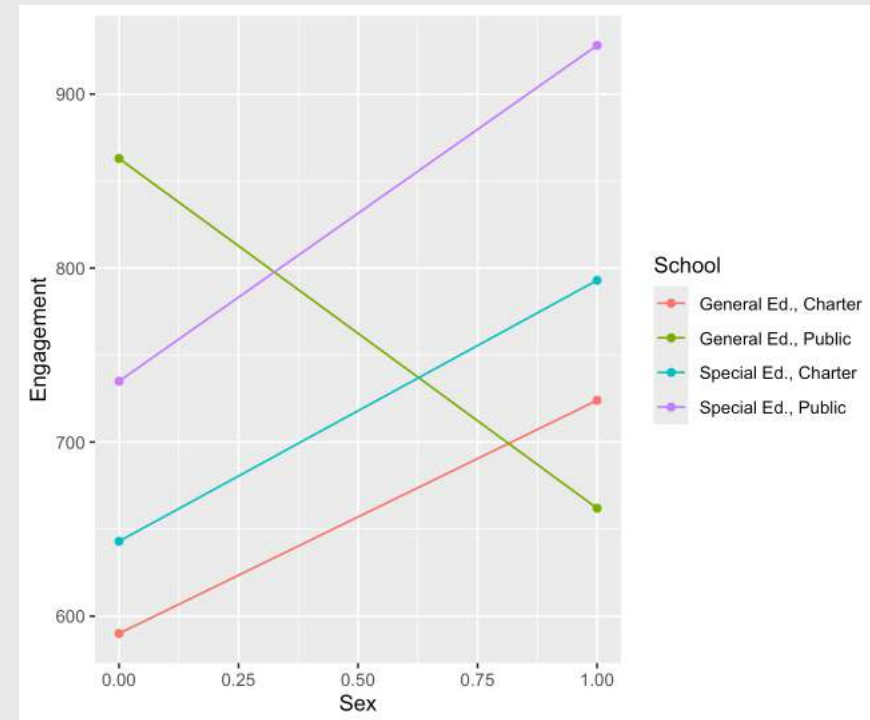
Data exploration: an iterative process

Initial exploratory plotting:

```
engagement_data %>%  
  ggplot() +  
  geom_col(aes(x = Engagement, y = School,  
              fill = Location),  
          position = 'dodge')
```

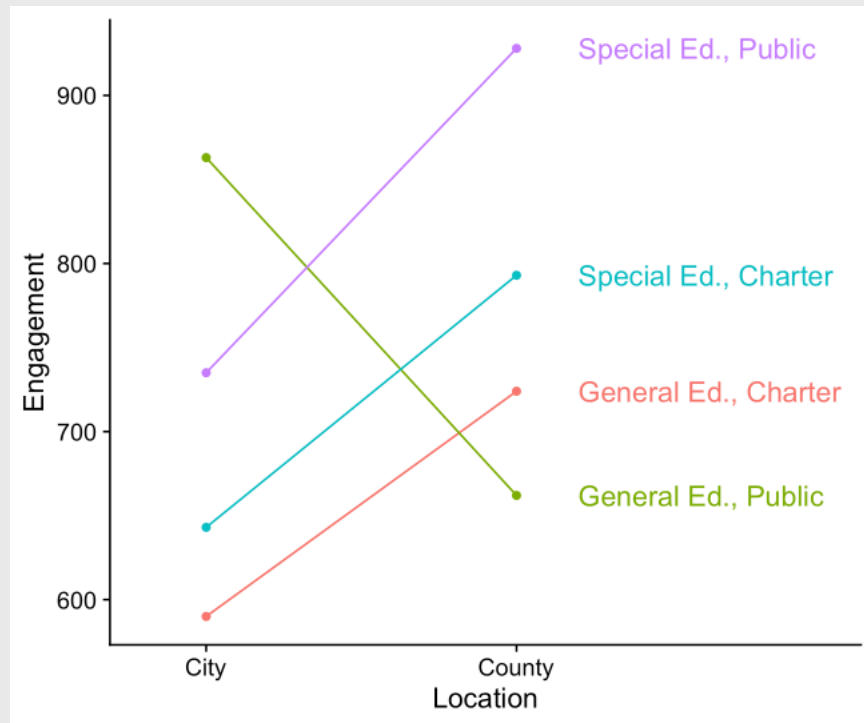


More exploratory plotting:
highlight difference

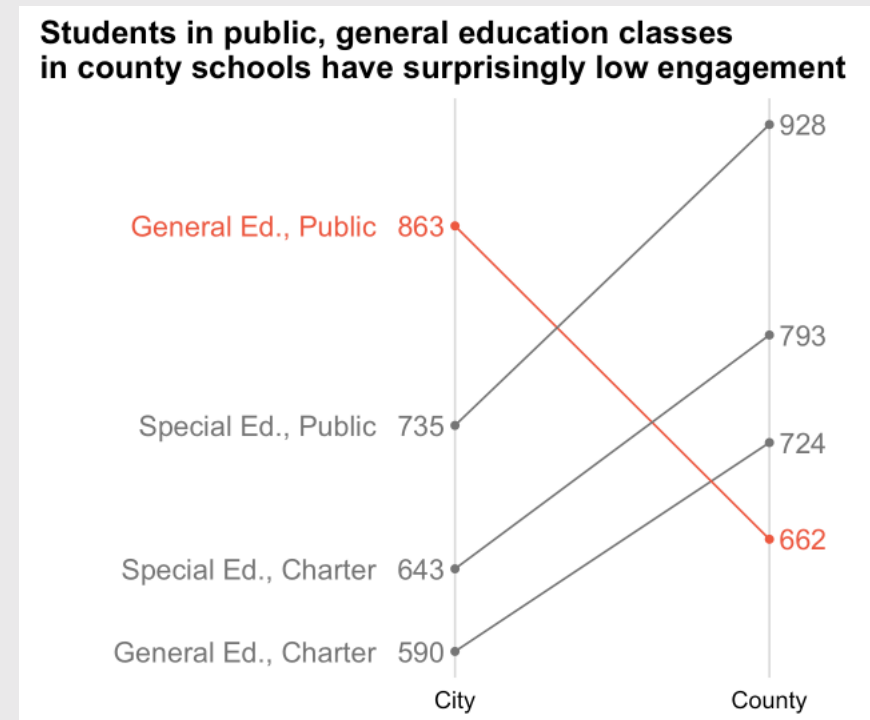


Data exploration: an iterative process

Directly label figure:



Remove unnecessary axes, change colors, fix labels:



A fully reproducible analysis

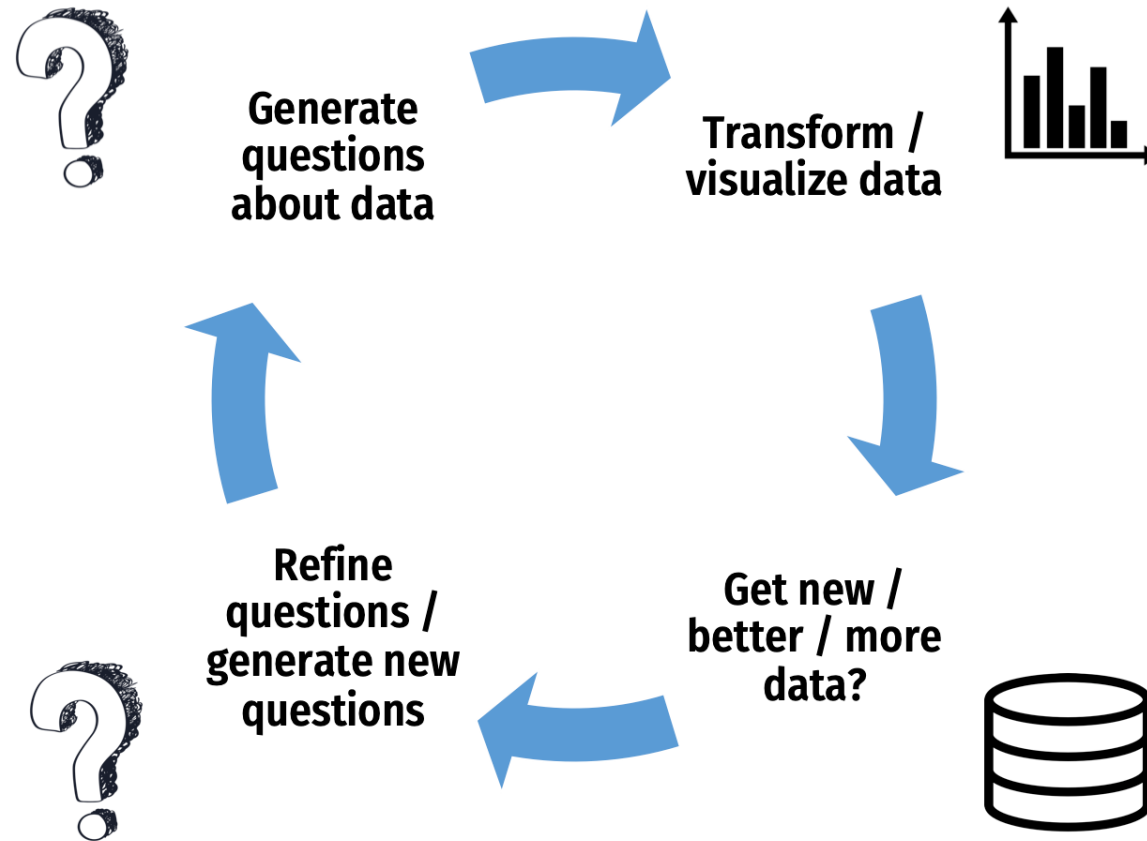
Code

Plot

```
data <- data.frame(
  City = c(643, 735, 590, 863),
  County = c(793, 928, 724, 662),
  School = c('Special Ed., Charter', 'Special Ed., Public',
             'General Ed., Charter', 'General Ed., Public'),
  Highlight = c(0, 0, 0, 1)) %>%
gather(Location, Engagement, City:County) %>%
mutate(
  Location = fct_relevel(Location, c('City', 'County')),
  Highlight = as.factor(Highlight),
  x = ifelse(Location == 'County', 1, 0))
```

```
plot <- ggplot(data, aes(x = x, y = Engagement, group = School, color = Highlight))
  geom_point() +
  geom_line() +
  scale_color_manual(values = c('#757575', '#ed573e')) +
  labs(x = 'Sex', y = 'Engagement',
       title = paste0('Students in public, general education classes\n',
                      'in county schools have surprisingly low engagement')) +
  scale_x_continuous(limits = c(-1.2, 1.2), labels = c('City', 'County'),
                    breaks = c(0, 1)) +
  geom_text_repel(aes(label = Engagement, color = as.factor(Highlight)),
                 data = subset(engagement, Location == 'County'),
                 size = 5,
                 nudge_x = 0.1,
                 segment.color = NA) +
  geom_text_repel(aes(label = Engagement, color = as.factor(Highlight)),
                 data = subset(engagement, Location == 'City'),
                 size = 5,
                 nudge_x = -0.1,
                 segment.color = NA) +
  geom_text_repel(aes(label = School, color = as.factor(Highlight)),
                 data = subset(engagement, Location == 'City'),
                 size = 5,
                 nudge_x = -0.25,
                 hjust = 1,
                 segment.color = NA) +
  theme_cowplot() +
  background_grid(major = 'x') +
  theme(axis.line = element_blank(),
        axis.title.x = element_blank(),
        axis.title.y = element_blank(),
        axis.text.y = element_blank(),
        axis.ticks = element_blank(),
        legend.position = 'none')
```

Data exploration: an iterative process



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Meet your instructor!



John Helveston, Ph.D.

- 2025 - Present: Associate Professor, EMSE
- 2018 - 2025: Assistant Professor, EMSE
- 2016-2018: Postdoc at [Institute for Sustainable Energy](#), Boston University
- 2016: PhD in Engineering & Public Policy at Carnegie Mellon University
- 2015: MS in Engineering & Public Policy at Carnegie Mellon University
- 2010: BS in Engineering Science & Mechanics at Virginia Tech
- Website: www.jhelvy.com

Meet your tutors!



Pingfan Hu

- Graduate Teaching Assistant (GTA)
- PhD student in EMSE
- Website: www.pingfanhu.com

Meet your tutors!



Bogdan Bunea

- Learning Assistant (LA)
- EMSE Junior & P4A / EDA alumni
- Check out his team's [project](#) from 2023

Prerequisites

EMSE 4574 / 6574: Intro to Programming for Analytics


You should be able to:

- Use RStudio to write basic R commands.
- Know the distinctions between different R operators and data types, including numeric, string, and logical data.
- Use **tidyverse** functions to wrangle and manipulate data in R.
- Use the **ggplot2** library to create plots in R.



Check out R for Analytics Primer

Course website

 Everything you need will be on the course website:
<https://eda.seas.gwu.edu/2025-Fall/>

 The [schedule](#) is the best starting point

Quizzes (10% of grade)




📅 At the start of class every other week-ish. Make ups only for excused absences (i.e. don't be late).

📅 5 total, lowest dropped

🕒 10 minutes

Why quiz at all? The "retrieval effect" - basically, you have to *practice* remembering things, otherwise your brain won't remember them (see the book ["Make It Stick: The Science of Successful Learning"](#))

Assignments

- 1)  Weekly Homework / Readings: [HW1](#)
- 2)  3 Mini Projects (due 2 weeks from date assigned)
- 3)  [Final Project](#)

Undergrads: Teams of 3 - 4 students

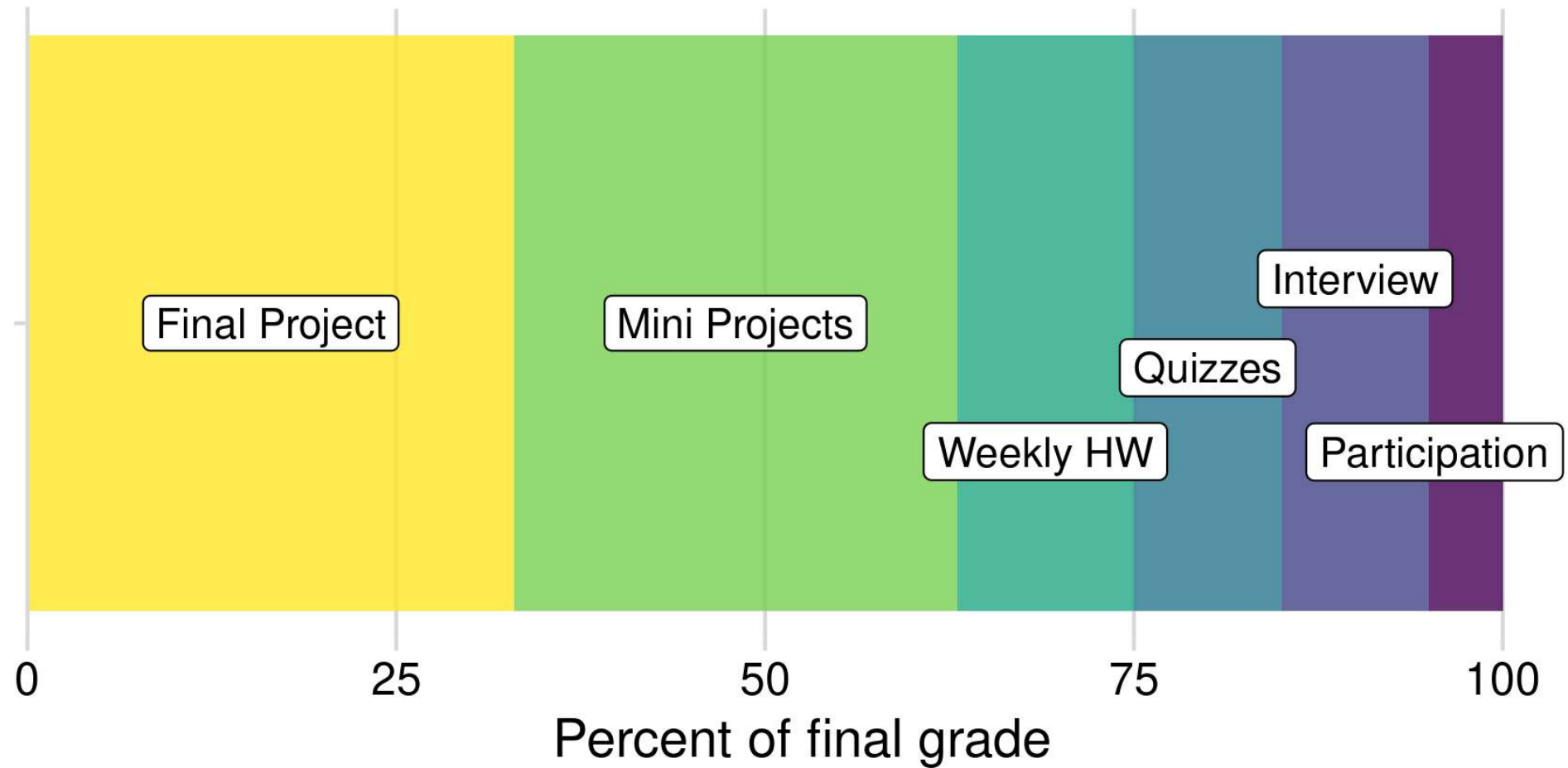
Grads: Teams of 2 students

Item	Due Date
Proposal	Sep 21
Progress Report	Oct 26
Final Report	Dec 07
Presentation	Dec 09

Grades

Item	Weight	Notes
Participation / Attendance	5%	(Yes, I take attendance)
Reflections	12 %	Weekly assignment, lowest dropped)
Quizzes	10 %	5 quizzes, lowest dropped
Mini Project 1	10 %	Individual assignments
Mini Project 2	10 %	
Mini Project 3	10 %	
Final Project: Proposal	6 %	
Final Project: Progress Report	6 %	
Final Project: Report	15 %	
Final Project: Presentation	6 %	
Final Interview	10 %	Individual interview

Grades



Course policies

- BE NICE
- BE HONEST
- DON'T CHEAT

Copying is good, stealing is bad

"Plagiarism is trying to pass someone else's work off as your own. Copying is about reverse-engineering."

-- Austin Kleon, from [Steal Like An Artist](#)

Use of chatGPT and other AI tools

- Large language models (LLMs) are pretty good
- Sometimes they suck.
- I will grade your work. **Your work should not suck**

Ways to not have your work suck:

- Don't submit code that doesn't run (actually run it before submitting it).
- Actually read what the AI generates.
- There are dozens of ways to do things - **you should use the approach I teach.**

Use AI as an assistant, not a solutions manual

Late submissions

- **5** late days - use them anytime, no questions asked
- No more than **2** late days on any one assignment
- Contact me for special cases

How to succeed in this class

 Participate during class!

 Start assignments early and **read carefully!**

 Actually read (before class)!

 Get sleep and take breaks often!

 Ask for help!

Getting Help

✳️ Use [Slack](#) to ask questions.

👤💻 Meet with your tutors

👤🕒 [Schedule a meeting](#) w/ Prof. Helveston:

- Mondays from 8:00-4:30pm
- Tuesdays from 8:00-4:30pm
- Fridays from 8:00-4:00pm

</> [GW Coders](#)

Course Software

 **Slack:** **turn notifications on!**

 **R & RStudio** (Install both)

 **Posit Cloud** (Register for free!)

Break

1. If you haven't already, install everything on the [software page](#)
2. Stand up, meet each other, (maybe form teams?...use [this sheet](#))

05 : 00

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Quick demo

1. Open `quarto_demo.qmd`
2. Click "Render"



Anatomy of a .qmd file

Header

Markdown text

R code

Define overall document options in header

Basic html page

```
---  
title: Your title  
author: Author name  
format: html  
---
```

Add table of contents, change theme

```
---  
title: Your title  
author: Author name  
toc: true  
format:  
  html:  
    theme: united  
---
```

More on themes at

<https://quarto.org/docs/output-formats/html-themes.html>

Render to multiple outputs

PDF uses LaTeX

```
---  
title: Your title  
author: Author name  
format: pdf  
---
```

Microsoft Word

```
---  
title: Your title  
author: Author name  
format: docx  
---
```

If you don't have LaTeX on your computer,
install tinytex in R:

```
tinytex::install_tinytex()
```

Anatomy of a .qmd file

~~Header~~

Markdown text

R code

Right now, bookmark this! 📌

<https://commonmark.org/help/>

(When you have 10 minutes, do this! 📌)

<https://commonmark.org/help/tutorial/>

Headers

HEADER 1

HEADER 2

HEADER 3

HEADER 4

HEADER 5

HEADER 6

HEADER 1

HEADER 2

HEADER 3

HEADER 4

HEADER 5

HEADER 6

Basic Text Formatting

Type this...

- normal text
- *_italic text_*
- **italic text**
- ****bold text****
- *****bold italic text*****
- ~~~~strikethrough~~~~
- ``code text``

..to get this

- normal text
- *italic text*
- *italic text*
- **bold text**
- ***bold italic text***
- ~~strikethrough~~
- `code text`

Lists

Bullet list:

- first item
- second item
- third item

- first item
- second item
- third item

Numbered list:

1. first item
2. second item
3. third item

1. first item
2. second item
3. third item

Links

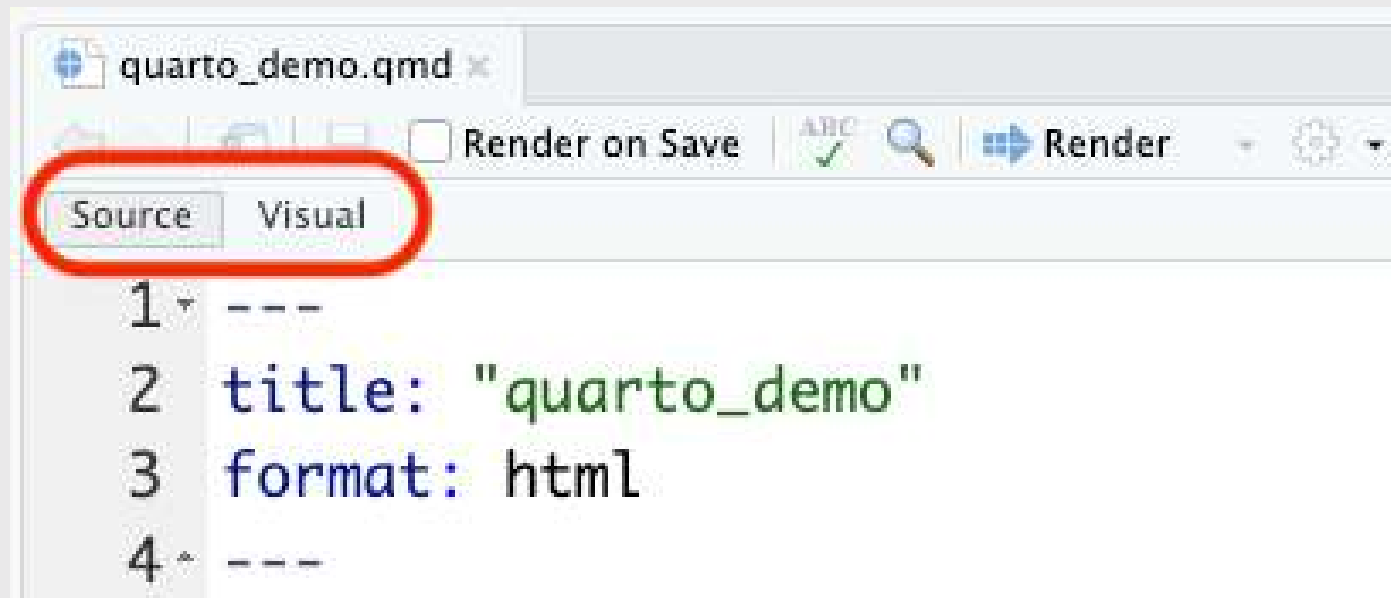
Simple **url link** to another site:

[Download R] (<http://www.r-project.org/>)

[Download R](http://www.r-project.org/)

Don't want to use Markdown?

Use Visual Mode!



Anatomy of a .qmd file

~~Header (think of this as the "settings")~~

~~Markdown text~~

R code

R Code

Inline code

```
`r insert code here`
```

Code chunks

```
```${r}  
insert code here
insert more code here
```
```

Inline R code

```
The sum of 3 and 4 is `r 3 + 4`
```

Produces this:

The sum of 3 and 4 is 7

R Code chunks

This code chunk...

```
```{r}  
library(palmerpenguins)

head(penguins)
```
```

...will produce this when compiled:

```
library(palmerpenguins)  
  
head(penguins)
```

```
#> # A tibble: 6 × 8  
#>   species island    bill_length_mm bill_c  
#>   <fct>    <fct>          <dbl>  
#> 1 Adelie  Torgersen         39.1  
#> 2 Adelie  Torgersen         39.5  
#> 3 Adelie  Torgersen         40.3  
#> 4 Adelie  Torgersen          NA  
#> 5 Adelie  Torgersen         36.7  
#> 6 Adelie  Torgersen         39.3
```

Chunk options

Control what chunks output using options

All options [here](#)

| option | default | effect |
|------------|----------|---|
| eval | TRUE | Whether to evaluate the code and include its results |
| echo | TRUE | Whether to display code along with its results |
| warning | TRUE | Whether to display warnings |
| error | FALSE | Whether to display errors |
| message | TRUE | Whether to display messages |
| tidy | FALSE | Whether to reformat code in a tidy way when displaying it |
| results | "markup" | "markup", "asis", "hold", or "hide" |
| cache | FALSE | Whether to cache results for future renders |
| comment | "##" | Comment character to preface results with |
| fig.width | 7 | Width in inches for plots created in chunk |
| fig.height | 7 | Height in inches for plots created in chunk |

Chunk output options

By default, code chunks print **code** + **output**

```
```${r}  
#| echo: false

cat('hello world!')
```
```

Prints only **output**
(doesn't show code)

```
#> hello world!
```

```
```${r}  
#| eval: false

cat('hello world!')
```
```

Prints only **code**
(doesn't run the code)

```
cat('hello world!')
```

```
```${r}  
#| include: false

cat('hello world!')
```
```

Runs, but doesn't print
anything

A global `setup` chunk 🌍

```
```\{r}  
#| label: setup
#| include: false

knitr::opts_chunk$set(
 warning = FALSE,
 message = FALSE,
 fig.path = "figs/",
 fig.width = 7.252,
 fig.height = 4,
 comment = "#>",
 fig.retina = 3
)
```
```

- Typically the first chunk
- All following chunks will use these options (i.e., sets global chunk options)
- You can (and should) use individual chunk options too
- Often where I load libraries, etc.

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Workflow for reading in data

1) Use R Projects (.Rproj files) to organize your analysis - **don't double-click .R files!**



2) Use the [here](#) package to create file paths

```
path <- here::here("folder", "file.csv")
```

3) Import data with these functions:

| File type | Function | Library |
|-----------|--------------|---------------|
| .csv | read_csv() | readr |
| .txt | read_table() | utils |
| .xlsx | read_excel() | readxl |

Importing Comma Separated Values (.csv)

Read in `.csv` files with `read_csv()`:

```
library(tidyverse)
library(here)

csvPath <- here('data', 'milk_production.csv')
milk_production <- read_csv(csvPath)

head(milk_production)
```

```
#> # A tibble: 6 × 4
#>   region    state    year milk_produced
#>   <chr>    <chr>    <dbl>         <dbl>
#> 1 Northeast Maine      1970      619000000
#> 2 Northeast New Hampshire 1970      356000000
#> 3 Northeast Vermont    1970     1970000000
#> 4 Northeast Massachusetts 1970      658000000
#> 5 Northeast Rhode Island 1970       75000000
#> 6 Northeast Connecticut 1970      661000000
```

Importing Text Files (.txt)

Read in `.txt` files with `read.table()`:

```
txtPath <- here('data', 'nasa_global_temps.txt')
global_temps <- read.table(txtPath, skip = 5, header = FALSE)
head(global_temps)
```

```
#>      V1      V2      V3
#> 1 1880 -0.15 -0.08
#> 2 1881 -0.07 -0.12
#> 3 1882 -0.10 -0.15
#> 4 1883 -0.16 -0.19
#> 5 1884 -0.27 -0.23
#> 6 1885 -0.32 -0.25
```

Importing Text Files (.txt)

Read in `.txt` files with `read.table()`:

```
txtPath <- here('data', 'nasa_global_temps.txt')
global_temps <- read.table(txtPath, skip = 5, header = FALSE)
names(global_temps) <- c('year', 'no_smoothing', 'loess') # Add header
head(global_temps)
```

```
#>   year no_smoothing loess
#> 1 1880      -0.15 -0.08
#> 2 1881      -0.07 -0.12
#> 3 1882      -0.10 -0.15
#> 4 1883      -0.16 -0.19
#> 5 1884      -0.27 -0.23
#> 6 1885      -0.32 -0.25
```

Importing Excel Files (.xlsx)

Read in `.xlsx` files with `read_excel()`:

```
library(readxl)
```

```
xlsxPath <- here('data', 'pv_cell_production.xlsx')
```

```
pv_cells <- read_excel(xlsxPath, sheet = 'Cell Prod by Country', skip = 2)
```

```
glimpse(pv_cells)
```

```
#> Rows: 25
#> Columns: 10
#> $ Year      <chr> NA, NA, "1995", "1996", "1997", "1998", "1999", "2000", "2001", "2002"
#> $ China     <chr> "Megawatts", NA, "NA", "NA", "NA", "NA", "NA", "NA", "2.5", "3", "10", "13"
#> $ Taiwan    <chr> NA, NA, "NA", "NA", "NA", "NA", "NA", "NA", "3.5", "8", "17", "39.299"
#> $ Japan     <dbl> NA, NA, 16.4, 21.2, 35.0, 49.0, 80.0, 128.6, 171.2, 251.1, 363.9, 601
#> $ Malaysia  <chr> NA, NA, "NA", "NA", "NA", "NA", "NA", "NA", "0", "0", "0", "0", "0",
#> $ Germany   <chr> NA, NA, "NA", "NA", "NA", "NA", "NA", "NA", "22.5", "23.5", "55", "121.5",
#> $ `South Korea` <chr> NA, NA, "NA", "NA", "NA", "NA", "NA", "NA", "0", "0", "0", "0", "5.3"
#> $ `United States` <dbl> NA, NA, 34.7500, 38.8500, 51.0000, 53.7000, 60.8000, 75.0000, 100.300
#> $ Others    <chr> NA, NA, "NA", "NA", "NA", "NA", "NA", "NA", "48.200000000000017", "69.80000
#> $ World     <dbl> NA, NA, 77.600, 88.600, 125.800, 154.900, 201.300, 276.800, 371.300,
```

Importing Excel Files (.xlsx)

Read in `.xlsx` files with `read_excel()`:

```
library(readxl)

xlsxPath <- here('data', 'pv_cell_production.xlsx')
pv_cells <- read_excel(xlsxPath, sheet = 'Cell Prod by Country', skip = 2) %>%
  mutate(Year = as.numeric(Year)) %>% # Convert "non-years" to NA
  filter(!is.na(Year)) # Drop NA rows in Year
```

```
glimpse(pv_cells)
```

```
#> Rows: 19
#> Columns: 10
#> $ Year      <dbl> 1995, 1996, 1997, 1998, 1999, 2000, 2001, 2002, 2003, 2004, 2005, 2006, 2007, 2008, 20
#> $ China     <chr> "NA", "NA", "NA", "NA", "NA", "2.5", "3", "10", "13", "40", "128.30000000000001", "34
#> $ Taiwan    <chr> "NA", "NA", "NA", "NA", "NA", "NA", "3.5", "8", "17", "39.299999999999997", "88", "16
#> $ Japan     <dbl> 16.4, 21.2, 35.0, 49.0, 80.0, 128.6, 171.2, 251.1, 363.9, 601.5, 833.0, 926.4, 937.5,
#> $ Malaysia  <chr> "NA", "NA", "NA", "NA", "NA", "NA", "0", "0", "0", "0", "0", "0", "100.1", "397.9", "
#> $ Germany   <chr> "NA", "NA", "NA", "NA", "NA", "22.5", "23.5", "55", "121.5", "193", "339", "469.1", "
#> $ `South Korea` <chr> "NA", "NA", "NA", "NA", "NA", "NA", "0", "0", "0", "0", "5.3", "13", "31.883935905674
#> $ `United States` <dbl> 34.7500, 38.8500, 51.0000, 53.7000, 60.8000, 75.0000, 100.3000, 120.6000, 103.0000, 1
#> $ Others    <chr> "NA", "NA", "NA", "NA", "NA", "48.200000000000017", "69.800000000000011", "97.2999999
#> $ World     <dbl> 77.600, 88.600, 125.800, 154.900, 201.300, 276.800, 371.300, 542.000, 749.400, 1198.80
```

Your turn

10:00

Open the `practice.qmd` file.

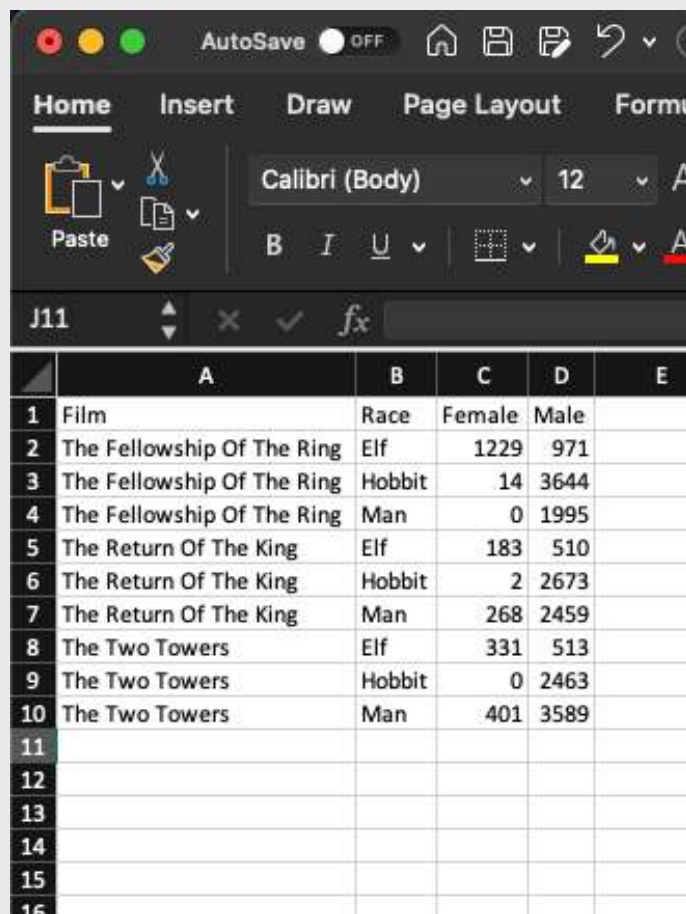
Write code to import the following data files from the "data" folder:

- For `lotr_words.csv`, call the data frame `lotr`
- For `north_america_bear_killings.txt`, call the data frame `bears`
- For `uspto_clean_energy_patents.xlsx`, call the data frame `patents`

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The data frame... in Excel



| | A | B | C | D | E |
|----|----------------------------|--------|--------|------|---|
| 1 | Film | Race | Female | Male | |
| 2 | The Fellowship Of The Ring | Elf | 1229 | 971 | |
| 3 | The Fellowship Of The Ring | Hobbit | 14 | 3644 | |
| 4 | The Fellowship Of The Ring | Man | 0 | 1995 | |
| 5 | The Return Of The King | Elf | 183 | 510 | |
| 6 | The Return Of The King | Hobbit | 2 | 2673 | |
| 7 | The Return Of The King | Man | 268 | 2459 | |
| 8 | The Two Towers | Elf | 331 | 513 | |
| 9 | The Two Towers | Hobbit | 0 | 2463 | |
| 10 | The Two Towers | Man | 401 | 3589 | |
| 11 | | | | | |
| 12 | | | | | |
| 13 | | | | | |
| 14 | | | | | |
| 15 | | | | | |
| 16 | | | | | |

The data frame... in R

lotr

```
#> # A tibble: 18 × 4
#>   film                                race  gender
#>   <chr>                                <chr>  <chr>
#> 1 The Fellowship Of The Ring  Elf    Female
#> 2 The Fellowship Of The Ring  Elf    Male
#> 3 The Fellowship Of The Ring  Hobbit Female
#> 4 The Fellowship Of The Ring  Hobbit Male
#> 5 The Fellowship Of The Ring  Man    Female
#> 6 The Fellowship Of The Ring  Man    Male
#> 7 The Return Of The King      Elf    Female
#> 8 The Return Of The King      Elf    Male
#> 9 The Return Of The King      Hobbit Female
#> 10 The Return Of The King      Hobbit Male
#> 11 The Return Of The King      Man    Female
#> 12 The Return Of The King      Man    Male
#> 13 The Two Towers             Elf    Female
#> 14 The Two Towers             Elf    Male
```

Columns: *Vectors* of values (must be same data type)

Extract a column using `$`

```
lotr$race
```

```
#> [1] "Elf"      "Elf"      "Hobbit"   "Hobbit"   "Man"      "Man"      "Elf"      "Elf"      "Hobbit"   "
```

Columns: *Vectors* of values (must be same data type)

Can also use brackets:

```
lotr$race
```

```
#> [1] "Elf" "Elf" "Hobbit" "Hobbit" "Man" "Man" "Elf" "Elf" "Hobbit" "
```

```
lotr[,2]
```

```
#> # A tibble: 18 × 1  
#>   race  
#>   <chr>  
#> 1 Elf  
#> 2 Elf  
#> 3 Hobbit  
#> 4 Hobbit  
#> 5 Man  
#> 6 Man  
#> 7 Elf  
#> 8 Elf  
#> 9 Hobbit
```

Rows: Information about individual observations

Information about the first row:

```
lotr[1,]
```

```
#> # A tibble: 1 × 4  
#>   film          race gender word_count  
#>   <chr>      <chr> <chr>      <dbl>  
#> 1 The Fellowship Of The Ring Elf      Female      1229
```

Information about rows 1 & 2:

```
lotr[1:2,]
```

```
#> # A tibble: 2 × 4  
#>   film          race gender word_count  
#>   <chr>      <chr> <chr>      <dbl>  
#> 1 The Fellowship Of The Ring Elf      Female      1229  
#> 2 The Fellowship Of The Ring Elf      Male        971
```

Quick Practice

Read in the `data.csv` file in the "data" folder:

```
data <- read_csv(here('data', 'data.csv'))
```

Now answer these questions:

- How many rows and columns are in the data frame?
- What type of data is each column?
- Preview the different columns - what do you think this data is about? What might one row represent?
- How many unique airlines are in the data frame?
- What is the shortest and longest air time for any one flight in the data frame?

The tidyverse: `stringr` + `dplyr` + `readr` + `ggplot2` + ...



Art by [Allison Horst](#)

The main `dplyr` "verbs"

| "Verb" | What it does |
|--------------------------|-------------------------------|
| <code>select()</code> | Select columns by name |
| <code>filter()</code> | Keep rows that match criteria |
| <code>arrange()</code> | Sort rows based on column(s) |
| <code>mutate()</code> | Create new columns |
| <code>summarize()</code> | Create summary values |

Core `tidyverse` concept: **Chain functions together with "pipes"**

`%>%`

Think of the words "...and then..."

```
data %>%  
  do_something() %>%  
  do_something_else()
```

Select columns with `select()`

Subset Variables (Columns)



Select columns with `select()`

Select the columns `film` & `race`

```
lotr %>%  
  select(film, race)
```

```
#> # A tibble: 18 × 2  
#>   film                                race  
#>   <chr>                             <chr>  
#> 1 The Fellowship Of The Ring      Elf  
#> 2 The Fellowship Of The Ring      Elf  
#> 3 The Fellowship Of The Ring      Hobbit  
#> 4 The Fellowship Of The Ring      Hobbit  
#> 5 The Fellowship Of The Ring      Man  
#> 6 The Fellowship Of The Ring      Man  
#> 7 The Return Of The King           Elf  
#> 8 The Return Of The King           Elf  
#> 9 The Return Of The King           Hobbit  
#> 10 The Return Of The King           Hobbit  
#> 11 The Return Of The King           Man  
#> 12 The Return Of The King           Man  
#> 13 The Two Towers                  Elf
```

Select columns with `select()`

Use the `-` sign to drop columns

```
lotr %>%  
  select(-film)
```

```
#> # A tibble: 18 × 3  
#>   race    gender word_count  
#>   <chr>  <chr>      <dbl>  
#> 1 Elf     Female      1229  
#> 2 Elf     Male        971  
#> 3 Hobbit Female        14  
#> 4 Hobbit Male       3644  
#> 5 Man     Female         0  
#> 6 Man     Male      1995  
#> 7 Elf     Female       183  
#> 8 Elf     Male       510  
#> 9 Hobbit Female         2  
#> 10 Hobbit Male     2673  
#> 11 Man     Female       268  
#> 12 Man     Male     2459  
#> 13 Elf     Female       231
```

Filter for rows with `filter()`

Subset Observations (Rows)



Filter for rows with `filter()`

Keep only the rows with Elf characters

```
lotr %>%  
  filter(race == "Elf")
```

```
#> # A tibble: 6 × 4  
#>   film      race gender word_count  
#>   <chr>   <chr> <chr>      <dbl>  
#> 1 The Fellowship Of The Ring Elf    Female    1229  
#> 2 The Fellowship Of The Ring Elf    Male      971  
#> 3 The Return Of The King    Elf    Female    183  
#> 4 The Return Of The King    Elf    Male     510  
#> 5 The Two Towers            Elf    Female    331  
#> 6 The Two Towers            Elf    Male     513
```

Filter for rows with `filter()`

Keep only the rows with Elf or Hobbit characters

```
lotr %>%  
  filter((race == "Elf") | (race == "Hobbit"))
```

```
#> # A tibble: 12 × 4  
#>   film      race gender word_count  
#>   <chr>    <chr>  <chr>      <dbl>  
#> 1 The Fellowship Of The Ring Elf      Female    1229  
#> 2 The Fellowship Of The Ring Elf      Male      971  
#> 3 The Fellowship Of The Ring Hobbit    Female     14  
#> 4 The Fellowship Of The Ring Hobbit    Male    3644  
#> 5 The Return Of The King    Elf      Female    183  
#> 6 The Return Of The King    Elf      Male     510  
#> 7 The Return Of The King    Hobbit    Female      2  
#> 8 The Return Of The King    Hobbit    Male    2673  
#> 9 The Two Towers           Elf      Female    331  
#> 10 The Two Towers           Elf      Male     513  
#> 11 The Two Towers           Hobbit    Female      0  
#> 12 The Two Towers           Hobbit    Male    2463
```

Filter for rows with `filter()`

Keep only the rows with Elf or Hobbit characters

```
lotr %>%  
  filter(race %in% c("Elf", "Hobbit"))
```

```
#> # A tibble: 12 × 4  
#>   film          race gender word_count  
#>   <chr>         <chr> <chr>      <dbl>  
#> 1 The Fellowship Of The Ring Elf      Female    1229  
#> 2 The Fellowship Of The Ring Elf      Male      971  
#> 3 The Fellowship Of The Ring Hobbit    Female     14  
#> 4 The Fellowship Of The Ring Hobbit    Male    3644  
#> 5 The Return Of The King    Elf      Female     183  
#> 6 The Return Of The King    Elf      Male     510  
#> 7 The Return Of The King    Hobbit    Female      2  
#> 8 The Return Of The King    Hobbit    Male    2673  
#> 9 The Two Towers           Elf      Female     331  
#> 10 The Two Towers           Elf      Male     513  
#> 11 The Two Towers           Hobbit    Female      0  
#> 12 The Two Towers           Hobbit    Male    2463
```


Logic operators for `filter()`

| Description | Example |
|--|---------------------------------|
| Values greater than 1 | <code>value > 1</code> |
| Values greater than or equal to 1 | <code>value >= 1</code> |
| Values less than 1 | <code>value < 1</code> |
| Values less than or equal to 1 | <code>value <= 1</code> |
| Values equal to 1 | <code>value == 1</code> |
| Values not equal to 1 | <code>value != 1</code> |
| Values in the set <code>c(1, 4)</code> | <code>value %in% c(1, 4)</code> |

Combine `filter()` and `select()`

Keep only the rows with Elf characters that spoke more than 1000 words, then select everything but the race column

```
lotr %>%  
  filter((race == "Elf") & (word_count > 1000)) %>%  
  select(-race)
```

```
#> # A tibble: 1 × 3  
#>   film                gender word_count  
#>   <chr>              <chr>      <dbl>  
#> 1 The Fellowship Of The Ring Female      1229
```

Create new variables with `mutate()`

Make New Variables



Create new variables with `mutate()`

Create a new variable, `word1000` which is `TRUE` if the character spoke 1,000 or more words

```
lotr %>%  
  mutate(word1000 = word_count >= 1000)
```

```
#> # A tibble: 18 × 5  
#>   film      race gender word_count word1000  
#>   <chr>    <chr> <chr>      <dbl> <lgl>  
#> 1 The Fellowship Of The Ring Elf      Female      1229 TRUE  
#> 2 The Fellowship Of The Ring Elf      Male        971 FALSE  
#> 3 The Fellowship Of The Ring Hobbit    Female        14 FALSE  
#> 4 The Fellowship Of The Ring Hobbit    Male      3644 TRUE  
#> 5 The Fellowship Of The Ring Man       Female        0 FALSE  
#> 6 The Fellowship Of The Ring Man       Male      1995 TRUE  
#> 7 The Return Of The King    Elf      Female      183 FALSE  
#> 8 The Return Of The King    Elf      Male       510 FALSE  
#> 9 The Return Of The King    Hobbit   Female        2 FALSE  
#> 10 The Return Of The King   Hobbit   Male     2673 TRUE  
#> 11 The Return Of The King   Man      Female      268 FALSE
```

Handling if/else conditions

`ifelse(<condition>, <if TRUE>, <else>)`

```
lotr %>%  
  mutate(word1000 = ifelse(word_count >= 1000, TRUE, FALSE))
```

```
#> # A tibble: 18 × 5  
#>   film          race gender word_count word1000  
#>   <chr>         <chr> <chr>      <dbl> <lgl>  
#> 1 The Fellowship Of The Ring Elf      Female    1229 TRUE  
#> 2 The Fellowship Of The Ring Elf      Male      971 FALSE  
#> 3 The Fellowship Of The Ring Hobbit    Female     14 FALSE  
#> 4 The Fellowship Of The Ring Hobbit    Male    3644 TRUE  
#> 5 The Fellowship Of The Ring Man      Female     0 FALSE  
#> 6 The Fellowship Of The Ring Man      Male    1995 TRUE  
#> 7 The Return Of The King    Elf      Female    183 FALSE  
#> 8 The Return Of The King    Elf      Male     510 FALSE  
#> 9 The Return Of The King    Hobbit    Female     2 FALSE  
#> 10 The Return Of The King    Hobbit    Male    2673 TRUE  
#> 11 The Return Of The King    Man      Female    268 FALSE  
#> 12 The Return Of The King    Man      Male    2459 TRUE
```

Sort data frame with `arrange()`

Sort the `lotr` data frame by `word_count`

```
lotr %>%  
  arrange(word_count)
```

```
#> # A tibble: 18 × 4  
#>   film          race gender word_count  
#>   <chr>         <chr> <chr>      <dbl>  
#> 1 The Fellowship Of The Ring Man    Female      0  
#> 2 The Two Towers Hobbit Female      0  
#> 3 The Return Of The King Hobbit Female      2  
#> 4 The Fellowship Of The Ring Hobbit Female     14  
#> 5 The Return Of The King Elf     Female    183  
#> 6 The Return Of The King Man     Female    268  
#> 7 The Two Towers Elf     Female    331  
#> 8 The Two Towers Man     Female    401  
#> 9 The Return Of The King Elf     Male     510  
#> 10 The Two Towers Elf     Male     513  
#> 11 The Fellowship Of The Ring Elf     Male     971  
#> 12 The Fellowship Of The Ring Elf     Female   1229  
#> 13 The Fellowship Of The Ring Man     Male     1005
```

Sort data frame with `arrange()`

Use the `desc()` function to sort in descending order

```
lotr %>%  
  arrange(desc(word_count))
```

```
#> # A tibble: 18 × 4  
#>   film          race gender word_count  
#>   <chr>         <chr> <chr>      <dbl>  
#> 1 The Fellowship Of The Ring Hobbit Male      3644  
#> 2 The Two Towers      Man   Male      3589  
#> 3 The Return Of The King  Hobbit Male      2673  
#> 4 The Two Towers      Hobbit Male      2463  
#> 5 The Return Of The King  Man   Male      2459  
#> 6 The Fellowship Of The Ring Man   Male      1995  
#> 7 The Fellowship Of The Ring Elf   Female    1229  
#> 8 The Fellowship Of The Ring Elf   Male      971  
#> 9 The Two Towers      Elf   Male      513  
#> 10 The Return Of The King Elf   Male      510  
#> 11 The Two Towers      Man   Female     401  
#> 12 The Two Towers      Elf   Female     331  
#> 13 The Return Of The King Man   Female     268
```

Your turn

10:00

Read in the `data.csv` file in the "data" folder:

```
data <- read_csv(here('data', 'data.csv'))
```

Now answer these questions:

- Create a new data frame, `flights_fall`, that contains only flights that departed in the fall semester.
- Create a new data frame, `flights_dc`, that contains only flights that flew to DC airports (Reagan or Dulles).
- Create a new data frame, `flights_dc_carrier`, that contains only flights that flew to DC airports (Reagan or Dulles) and only the columns about the month and airline.
- How many unique airlines were flying to DC airports in July?
- Create a new variable, `speed`, in miles per hour using the `time` (minutes) and `distance` (miles) variables.
- Which flight flew the fastest?
- Remove rows that have `NA` for `air_time` and re-arrange the resulting data frame based on the longest air time and longest flight distance.

Week 1: *Getting Started*

1. Course Goal
2. Course Introduction
3. Break: Install Stuff
4. Quarto
5. Workflow & Reading In Data
6. Wrangling Data
7. Visualizing Data

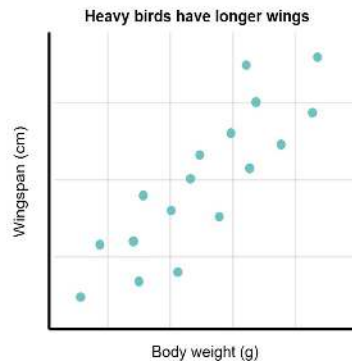
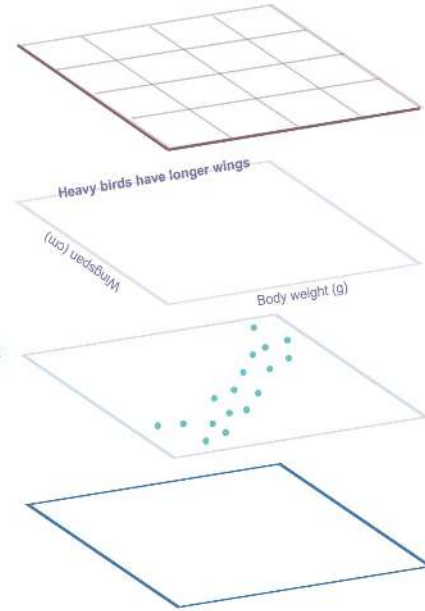
MAKING A GRAPH WITH GGPLOT2

Customise the look of your plot with themes
(pre-made or your own!):
`+ theme_bw()`

Add labels and titles:
`+ labs(x = "Body weight (g)", y = "Wingspan (cm)",
title = "Heavy birds have longer wings")`

Specify the type of graph and the variables to use:
`+ geom_point(aes(x = body.weight, y = wingspan))`

Plot the device containing your data:
`ggplot(data = birds)`



"Grammar of Graphics"

Concept developed by Leland Wilkinson
(1999)

ggplot2 package developed by Hadley
Wickham (2005)

Making plot layers with ggplot2

1. The data
2. The aesthetic mapping (what goes on the axes?)
3. The geometries (points? bars? etc.)
4. The annotations / labels
5. The theme

Layer 1: The data

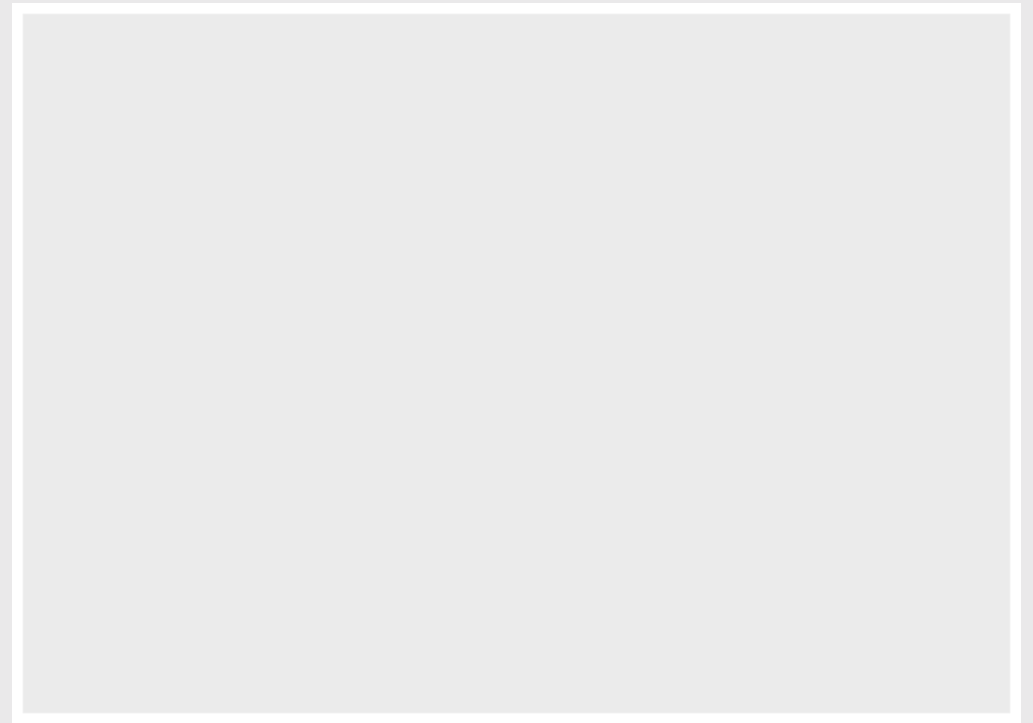
```
head(mpg)
```

```
#> # A tibble: 6 × 11
#>   manufacturer model displ  year   cyl trans      drv   cty   hwy fl  class
#>   <chr>         <chr> <dbl> <int> <int> <chr>    <chr> <int> <int> <chr> <chr>
#> 1 audi         a4      1.8  1999     4 auto(l5)  f     18    29 p    compact
#> 2 audi         a4      1.8  1999     4 manual(m5) f     21    29 p    compact
#> 3 audi         a4      2    2008     4 manual(m6) f     20    31 p    compact
#> 4 audi         a4      2    2008     4 auto(av)   f     21    30 p    compact
#> 5 audi         a4      2.8  1999     6 auto(l5)  f     16    26 p    compact
#> 6 audi         a4      2.8  1999     6 manual(m5) f     18    26 p    compact
```

Layer 1: The data

The `ggplot()` function initializes the plot with whatever data you're using

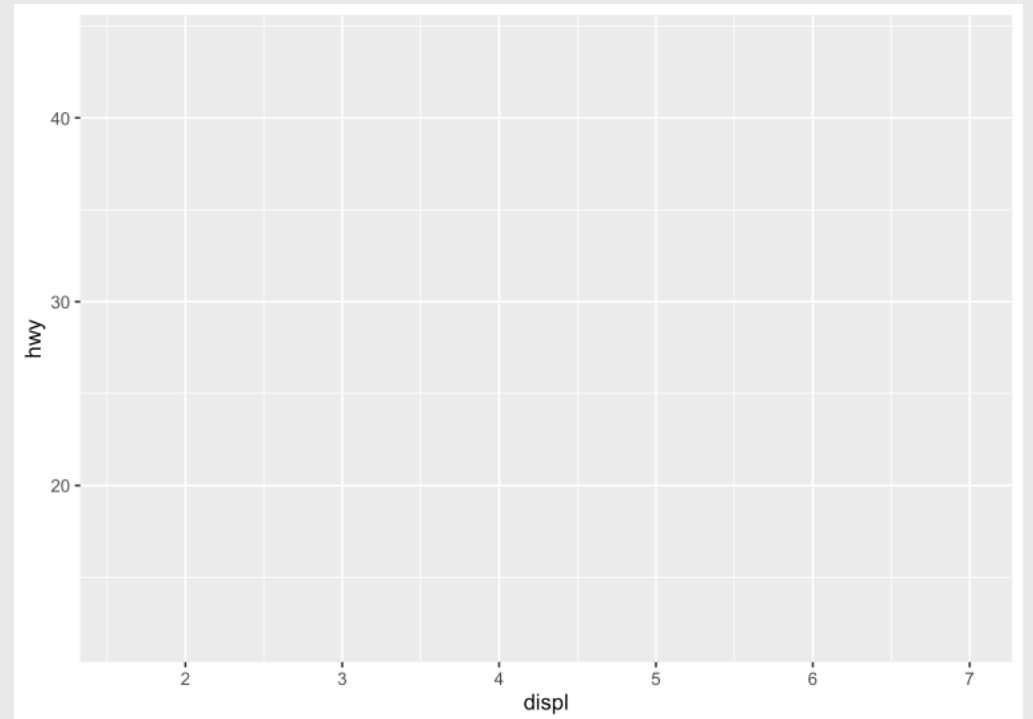
```
mpg %>%  
  ggplot()
```



Layer 2: The aesthetic mapping

The `aes()` function determines which variables will be *mapped* to the geometries (e.g. the axes)

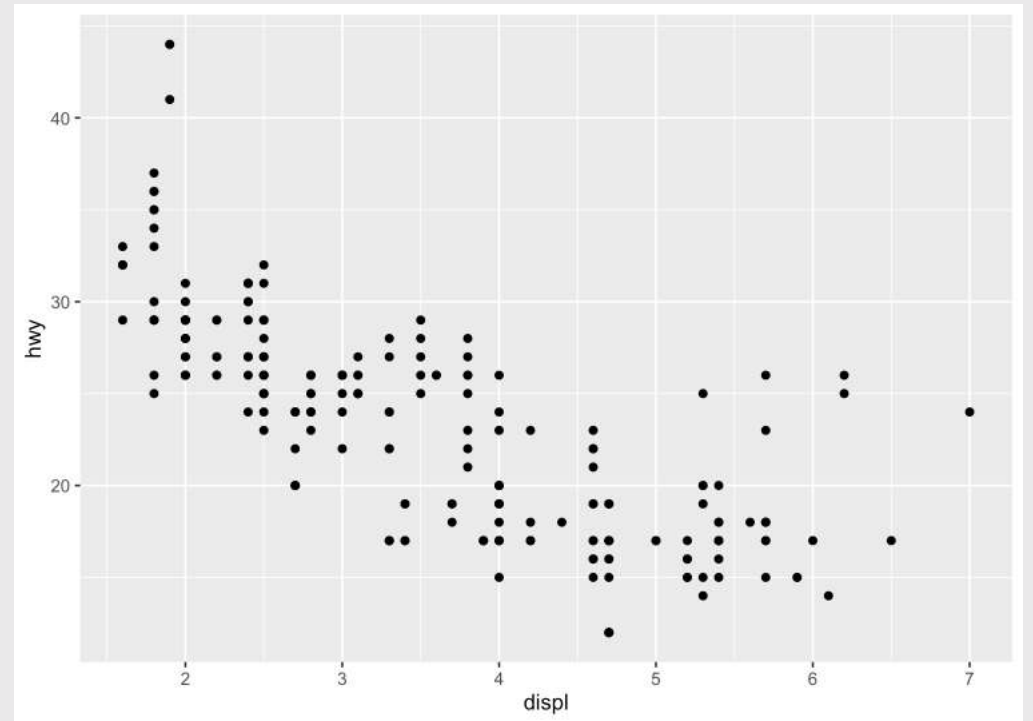
```
mpg %>%  
  ggplot(aes(x = displ, y = hwy))
```



Layer 3: The geometries

Use `+` to add geometries, e.g. `geom_points()` for points

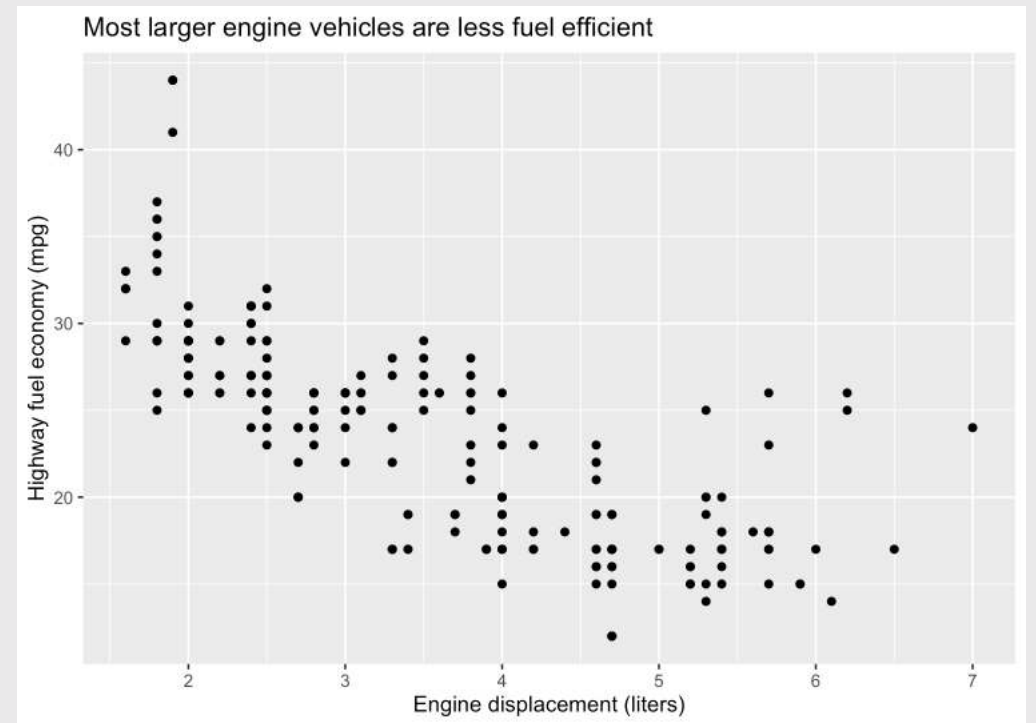
```
mpg %>%  
  ggplot(aes(x = displ, y = hwy)) +  
  geom_point()
```



Layer 4: The annotations / labels

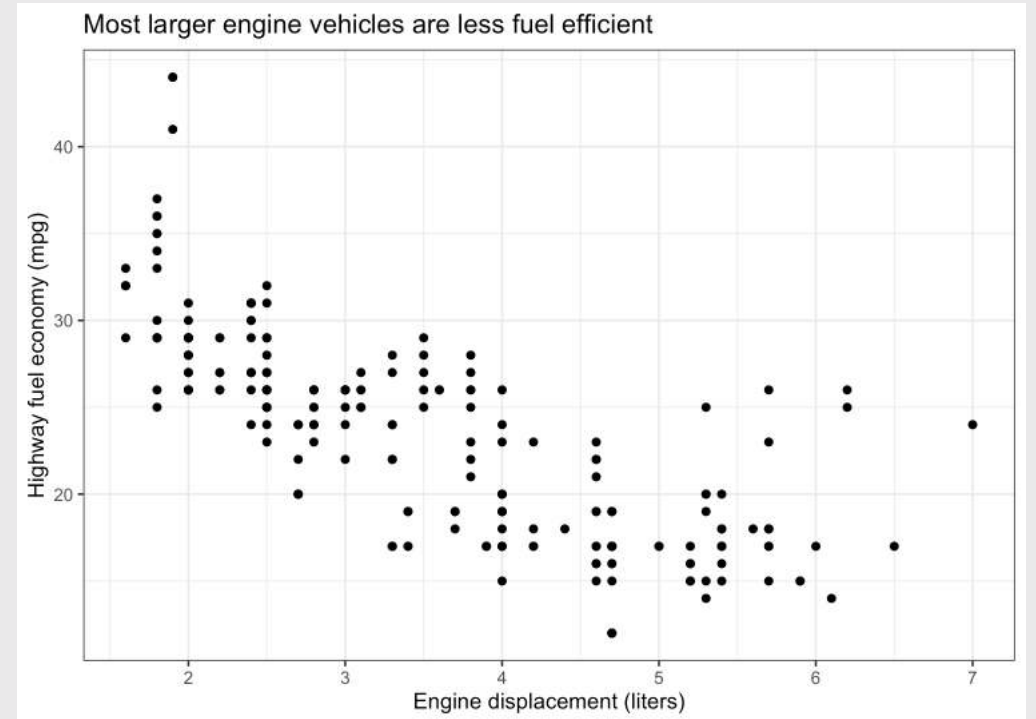
Use `labs()` to modify most labels

```
mpg %>%  
  ggplot(aes(x = displ, y = hwy)) +  
  geom_point() +  
  labs(  
    x = "Engine displacement (liters)",  
    y = "Highway fuel economy (mpg)",  
    title = "Most larger engine vehicles are less fuel efficient"  
  )
```



Layer 5: The theme

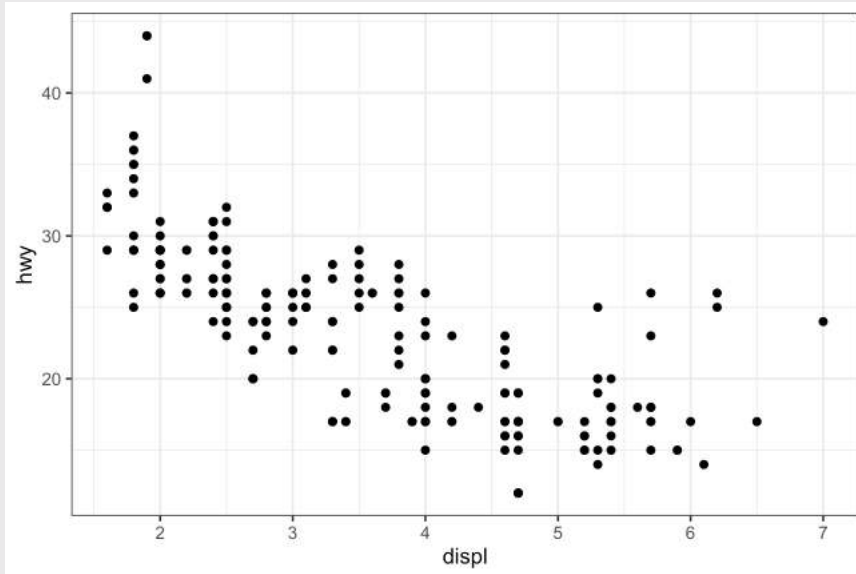
```
mpg %>%  
  ggplot(aes(x = displ, y = hwy)) +  
  geom_point() +  
  labs(  
    x = "Engine displacement (liters)",  
    y = "Highway fuel economy (mpg)",  
    title = "Most larger engine vehicles are  
  ) +  
  theme_bw()
```



Common themes

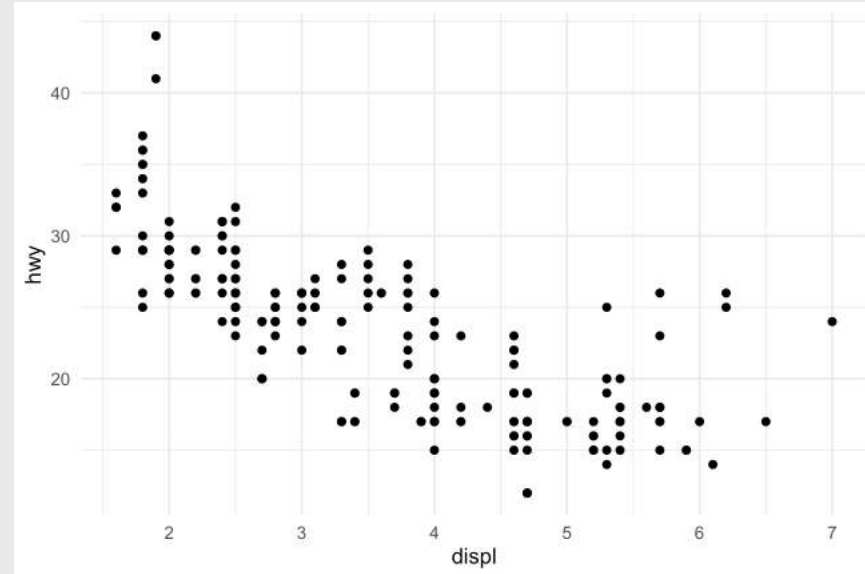
theme_bw()

```
mpg %>%  
  ggplot(aes(x = displ, y = hwy)) +  
  geom_point() +  
  theme_bw()
```



theme_minimal()

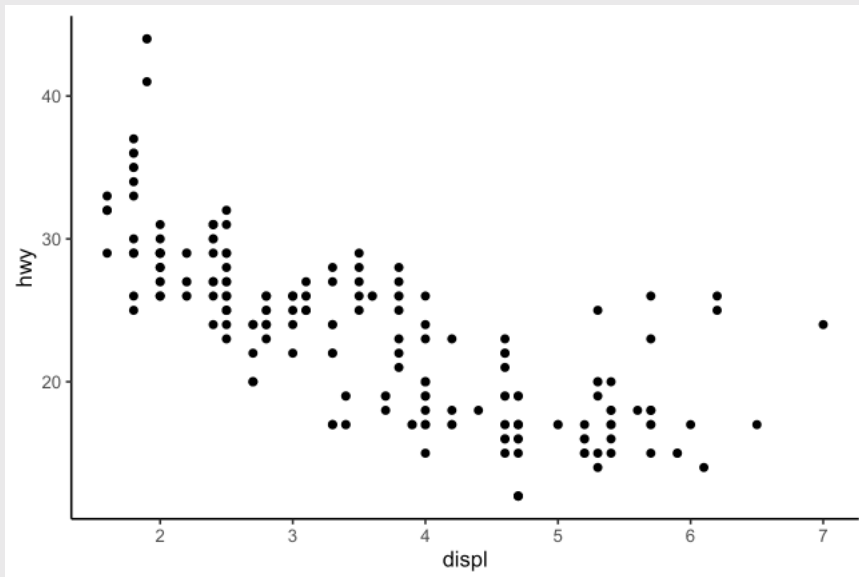
```
mpg %>%  
  ggplot(aes(x = displ, y = hwy)) +  
  geom_point() +  
  theme_minimal()
```



Common themes

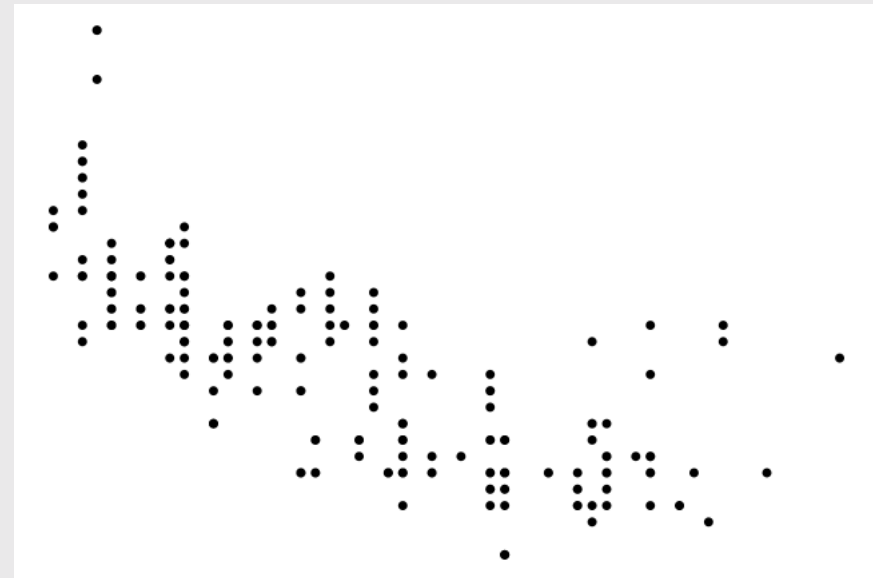
theme_classic()

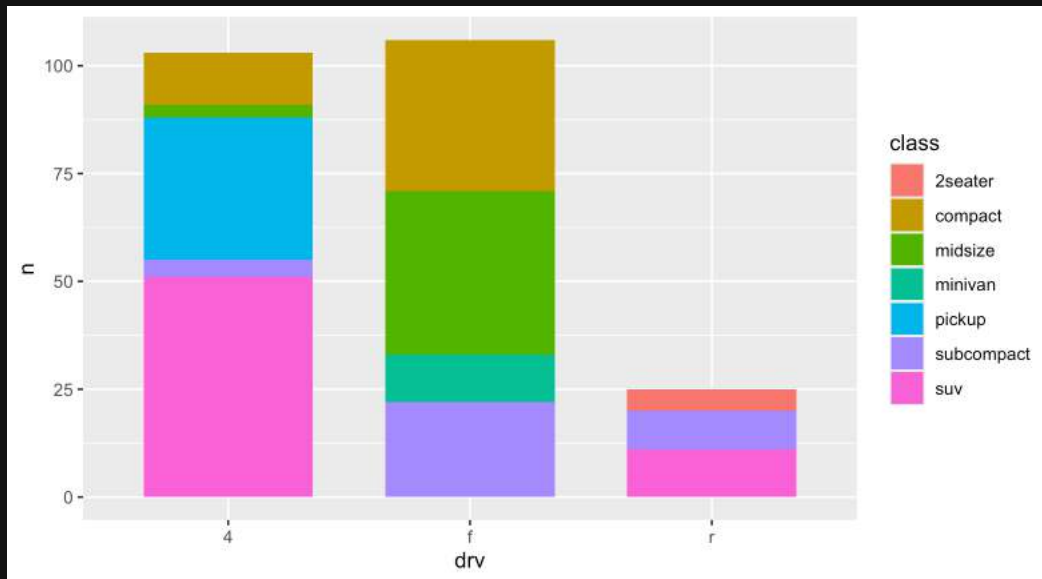
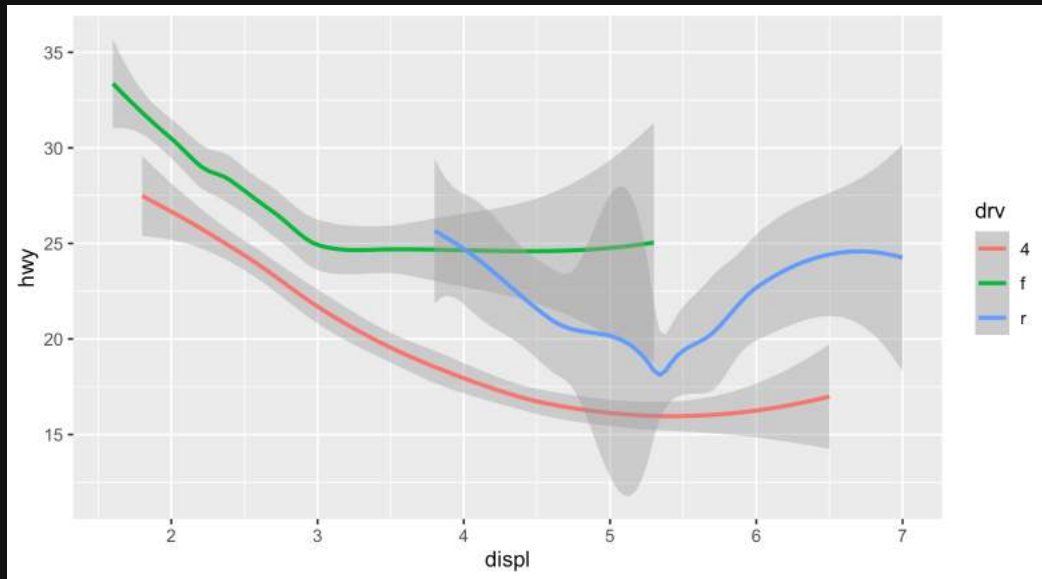
```
mpg %>%  
  ggplot(aes(x = displ, y = hwy)) +  
  geom_point() +  
  theme_classic()
```



theme_void()

```
mpg %>%  
  ggplot(aes(x = displ, y = hwy)) +  
  geom_point() +  
  theme_void()
```



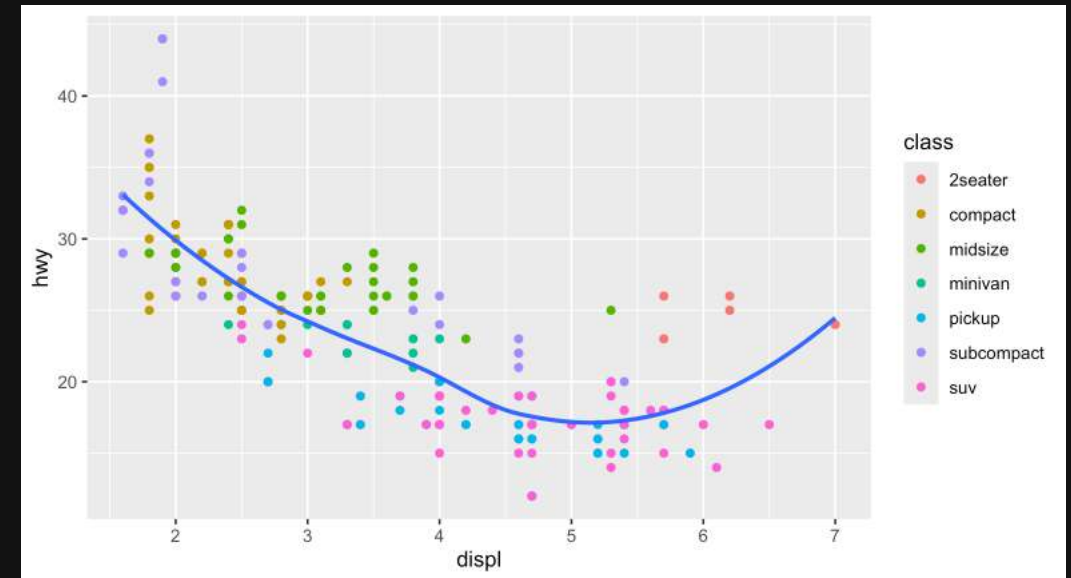


Your turn

15:00

Open `practice.qmd`

Use the `mpg` data frame and ggplot to create these charts



Extra practice

