

# Teaching R in DASC 1104

## R Part I

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# Materials

- Available on gitHub at XXXXXX

# Guiding principles

- Make the initial burden to working with real data as minimal as possible
- Teach a **process** of working with data more than the actual programming
  - Focus on actions/principles rather than specific syntax
  - More than in other programming classes, focus on **reproducibility**
- Great resources for learning R/python/git/shell
  - <https://software-carpentry.org/lessons/index.html>
- Scaffold Learning objectives and provide many examples
  - Here is a dataset -- Generate a scatterplot of mpg vs. engine displacement (demo with `mtcars`)
  - Start with simple examples and show how to build
  - Encourage student participation by working examples alongside

# Guiding principles

- Allow for student creativity and spontaneity
  - Different from typical programming classes
  - Here is a dataset -- Create two visualizations that tell me something interesting about it
- Focus on the high-level concepts before getting into the details
  - Some low level knowledge is ok -- variable types
  - Other low level knowledge not needed at this stage
    - Parsing files one line at a time, etc.

# The Rstudio IDE

- I recommend the **RStudio IDE**, but you do not have to.
  - Allows for R and python code development
- Can be installed **locally** or on a **server** (I use a SSO server option)
- **RStudio Cloud** provides a remote-hosting option for teaching

# Project-oriented workflows

- Especially useful for students with less comp-sci experience
- Makes file path management easier
- References [here](#) and [here](#) and [here](#)

## What does this code do?

```
library(tidyverse)
setwd("~/Documents/John/hw-1")
dat <- read_csv("my-secrets.csv")
```

- What might be the issue with this code?

# Hands-on experience project management in Rstudio

- Use of Rstudio projects (5-10 minutes)
- Use of the `here` library and the `here()` function
- Namespaces in R
- Generate "data file" in excel, load and process the data
- For more details, see Lecture 9 in unit 4

# The Two Towers

- The R language has split into two dialects "base R" and the "tidyverse"
- **Much** has **been written** about this difference
- Having learned Base R first and tidyverse second -- I am a strong proponent of the tidyverse -- there is a strong movement to **translate this syntax to python**
  - Most of my students would agree that tidyverse is the way to go
  - **Many python users** also agree



# The Rings of Power

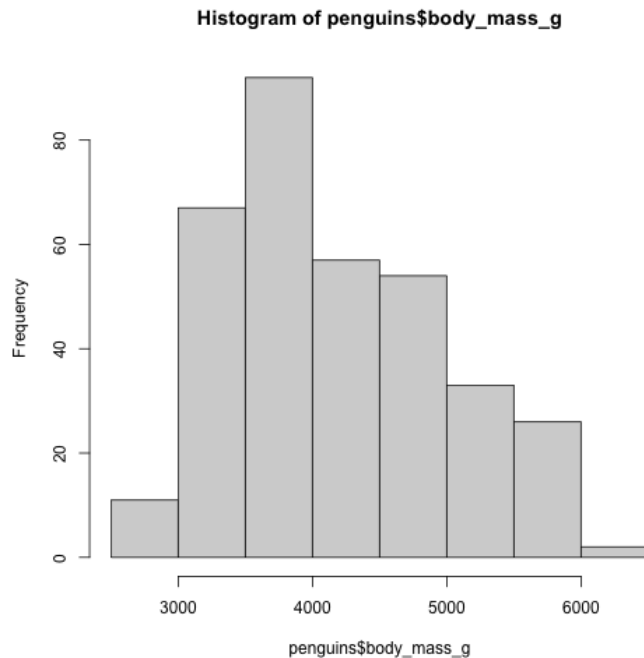
```
library(tidyverse)
library(palmerpenguins)
glimpse(penguins)
```

```
## Rows: 344
## Columns: 8
## $ species      [3m [38;5;246m<fct> [39m [23m Adelie, Adelie, Adelie, Adelie, Adelie, Adelie
## $ island       [3m [38;5;246m<fct> [39m [23m Torgersen, Torgersen, Torgersen, Torgersen, To
## $ bill_length_mm [3m [38;5;246m<dbl> [39m [23m 39.1, 39.5, 40.3, NA, 36.7, 39.3, 38.9, 39.2, 3
## $ bill_depth_mm [3m [38;5;246m<dbl> [39m [23m 18.7, 17.4, 18.0, NA, 19.3, 20.6, 17.8, 19.6, 1
## $ flipper_length_mm [3m [38;5;246m<int> [39m [23m 181, 186, 195, NA, 193, 190, 181, 195, 193, 19
## $ body_mass_g   [3m [38;5;246m<int> [39m [23m 3750, 3800, 3250, NA, 3450, 3650, 3625, 4675, 3
## $ sex           [3m [38;5;246m<fct> [39m [23m male, female, female, NA, female, male, female
## $ year          [3m [38;5;246m<int> [39m [23m 2007, 2007, 2007, 2007, 2007, 2007, 2007, 2007
```

# The Rings of Power

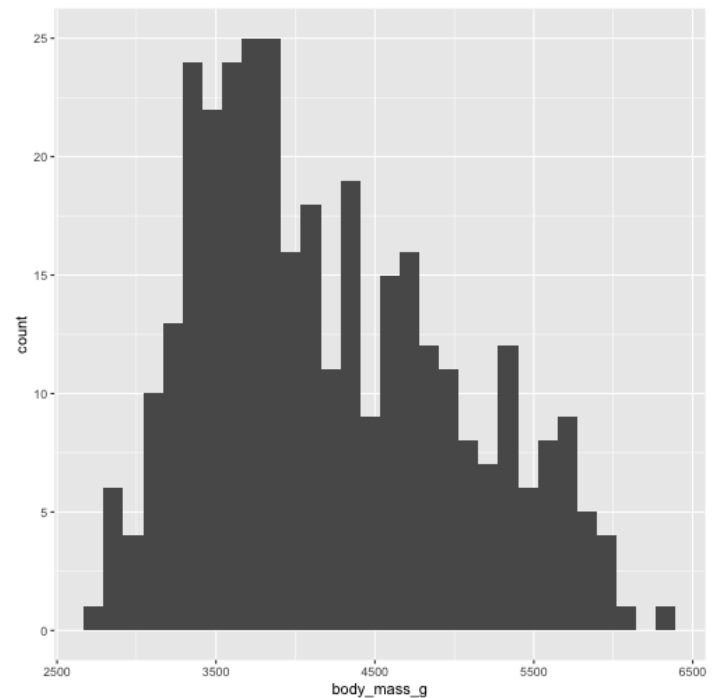
- Create a histogram of `body_mass_g`
- Using base R

```
hist(penguins$body_mass_g)
```



- Using `tidyverse`

```
ggplot(penguins, aes(x = body_mass_g)) +  
  geom_histogram()
```



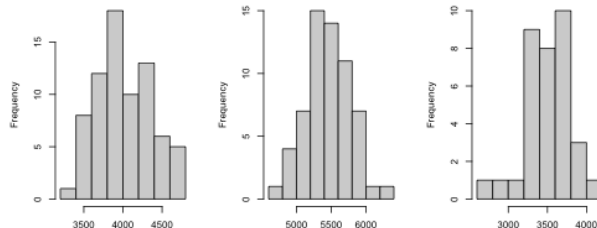
# But they were all of them deceived

- Create a histogram of `body_mass_g` for each `species` and `sex`

- Using base R

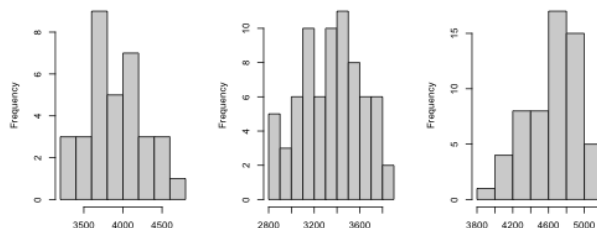
```
layout(matrix(1:6, 2, 3))
hist(penguins[penguins$sex == "male" & penguins$species == "male", "body_mass_g"])
hist(penguins[penguins$sex == "male" & penguins$species == "female", "body_mass_g"])
hist(penguins[penguins$sex == "female" & penguins$species == "male", "body_mass_g"])
hist(penguins[penguins$sex == "female" & penguins$species == "female", "body_mass_g"])
hist(penguins[penguins$sex == "male" & penguins$species == "Adelie", "body_mass_g"])
hist(penguins[penguins$sex == "female" & penguins$species == "Adelie", "body_mass_g"])
```

`penguins$sex == "male" & penguins$species == "male" & penguins$species == "female" & penguins$species == "Adelie"`



`penguins$sex == "male" & penguins$species == "Adelie" & penguins$species == "Gentoo" & penguins$species == "Chinstrap"`

`penguins$sex == "male" & penguins$species == "female" & penguins$species == "female" & penguins$species == "Adelie"`



`penguins$sex == "male" & penguins$species == "Chinstrap" & penguins$species == "Adelie" & penguins$species == "Gentoo"`

- Using `tidyverse`

```
ggplot(penguins, aes(x = body_mass_g)) +
  geom_histogram() +
  facet_grid(sex ~ species)
```

# For in secret the Dark Lord Sauron forged in secret a Master Ring, to control all others.

- How many penguins are in the dataset of each sex and species?

```
with(penguins, table(sex, species))
```

```
##           species
## sex      Adelie Chinstrap Gentoo
## female      73      34      58
## male        73      34      61
```

```
penguins %>%
  count(sex, species)
```

```
## # A tibble: 8 x 3
##   sex      species     n
##   <fct>   <fct>   <int>
## 1 female Adelie      73
## 2 female Chinstrap  34
## 3 female Gentoo     58
## 4 male   Adelie      73
## 5 male   Chinstrap  34
## 6 male   Gentoo     61
## 7 <NA>   Adelie       6
## 8 <NA>   Gentoo       5
```

- Where did the NA values go?

# One ring to rule them all!

- What does this code do?

```
aggregate(airquality[, "Ozone"],  
          list(Month = airquality[, "Month"]),  
          mean, na.rm = TRUE)
```

```
##      Month      x  
## 1      5 23.61538  
## 2      6 29.44444  
## 3      7 59.11538  
## 4      8 59.96154  
## 5      9 31.44828
```

```
airquality %>%  
  group_by(Month) %>%  
  summarize(mean_o3 = mean(Ozone, na.rm = TRUE))
```

```
## # A tibble: 5 x 2  
##   Month mean_o3  
##   <int>   <dbl>  
## 1     5    23.6  
## 2     6    29.4  
## 3     7    59.1  
## 4     8    60.0  
## 5     9    31.4
```

- Which example is more **human readable**?

# Working with RMarkdown

- Demo with RMarkdown examples (15-20 minutes)
- Code chunks
- Running code (ctrl-enter, run all above)
- installing vs. loading packages
- Restarting the R environment and clearing variables
  - Do this often
- **Settings**--don't save your workspace
- Plain text, math equations, and communication
- Code chunk options
- File paths and loading data
- Managing the project environment
- Formatting and troubleshooting
- The tinytex/latex problem
- python code in RMarkdown
- sourcing in scripts
- live preview (I'm learning this in real time)
- Getting help with ?



# What to teach

- Programming concepts are important
    - loops, data structures, functions
  - But, I **don't** teach R as if it were a programming language
    - R is a programming language for **data science**
    - Many "real" programmers don't like R because it is not a "programming first" language
      - R is a **data-science**-first language, don't get mad when it isn't what you want it to be
    - Show how to do data science using R as a tool
  - Lead with data visualization, follow up with data manipulation
- 

## Explore for meaningful datasets to use



# Focus on basic syntax and datatypes (demo)

Typically in class, I include code in the slides, but not here

- vectors
- matrices
- arrays
- lists
- factors
- data.frames

# Atomic types (demo)

- logical/boolean
- numeric
- integer
- complex (not covered)
- character/string

# Working with data

- what are data.frames?
- accessing rows/columns with `[`
- accessing variables with `[`
- accessing variables with `$`

