# Introduction to Data Science



## **Course Wrap-Up**

Today we are going to discuss a few final closing thoughts about the course:

- 1. data science in the wild
- 2. data science minor
- 3. SQL and Python
- 4. evaluations

I will be brief so that we have plenty of time to work on the final project.



# 1. Data Science in the Wild



## **Data Science in the Wild**

I encourage you to use the techniques in this class in future endeavors. This may be as soon as a course next semester, or it may be a job several years down the road.

I suggest keeping two things in mind:

- 1. We covered the basics in the first 8 weeks, and these will cover most use-cases. Do not get overwhelmed by the more advanced stuff.
- 2. The most important thing to keep in mind are the class notes about creating data. If you make the basic data well, it is easy to use **ggplot2** + **dplyr**.

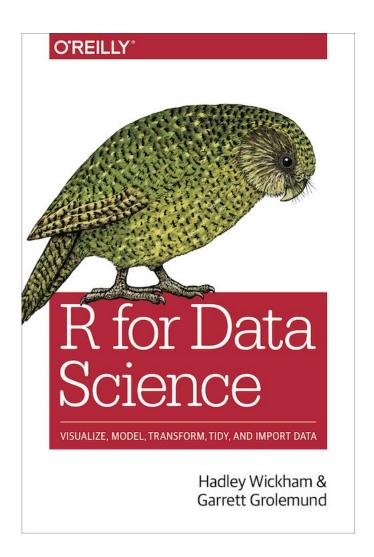




# **Getting Help**

If you are trying to get help with R, or just data science in general, here are a few sources of help:

- 1. our course website (it should still be up)
- 2. The R for Data Science book
- 3. R package vignettes
- 4. GitHub issues
- 5. ROpenSci and the R Journal



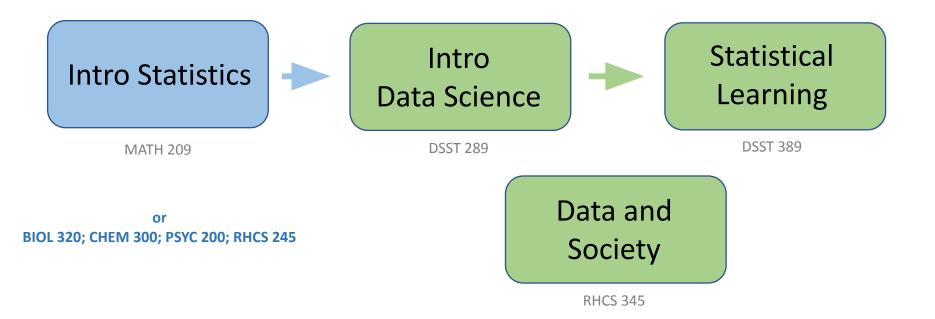


#### 2. DS Minor



#### **Data Science and Statistics Minor: Structure**

A total of six credits. See <u>datascience.richmond.edu</u> for more information



**BIOL 336** 

CHEM 301, 314, 315

CMCS 325, 327

GEOG 260, 360, 365

MATH 329, 330

PSYC 300, 343

INTRODUCTION (1)

**CORE COURSES (3)** 

**ELECTIVES (2)** 

the full list is much longer; see the website for the most up-to-date set of electives

# 3. SQL and Python



## **What We Have Learned**

I warned you all at the start of the semester that this course can often feel like we are just learning the programming language R, while in fact we are learning more general concepts from various fields of data science.

As a way of giving a course recap and to illustrate this, let's look at two other popular languages for data analysis. This is not a complete introduction to them, but a good starting point.







# **Structured Query Language (SQL)**

A language for manipulating data from a data base. Closely linked to the data verbs from **dplyr**.

SELECT calories, sugar FROM food WHERE food\_group = "fruit" ORDER BY sugar;

```
food %>%
filter(food_group == "fruit") %>%
select(calories, sugar) %>%
arrange(sugar)
```

SELECT mu AS avg(calories) FROM food GROUP BY food\_group;

```
food %>%
  group_by(food_group) %>%
  summarize(mu = mean(calories))
```



# **Python**

A language very similar to R, particularly when you use the popular pandas library and Python notebooks.

```
import numpy as np
import pandas as pd
from plotnine import ggplot, geom_point, geom_text, aes
dt = pd.read_csv('notes/data/food.csv')
   dt <- read_csv("notes/data/food.csv")</pre>
(ggplot(dt, aes('calories', 'sugar', color='factor(food_group)')) +
  geom_point())
   dt %>% ggplot(aes(calories, sugar)) + geom point()
```



# Python (cont.)

```
dt[dt['food_group'] == 'vegetable']
   filter(dt, food_group == 'vegetable')
dt['calories2'] = dt['calories'] * 2
    mutate(dt, calories2 = calories * 2)
dt[['fiber', 'food_group']]
   select(dt, fiber, food_group)
dt.sort_values('calories')
   arrange(dt, calories)
```



#### 4. Course Evaluations



#### **Course Evaluations**

Finally, I ask that you finish today by taking a few minutes to fill out the course evaluations. These are helpful as we think about how this class evolves under the new data science minor.

