

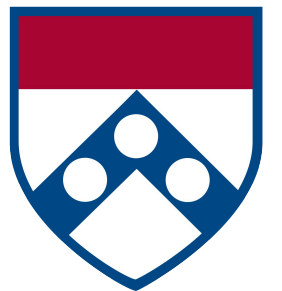
31 July 2017

Wharton PhD Tech Camp

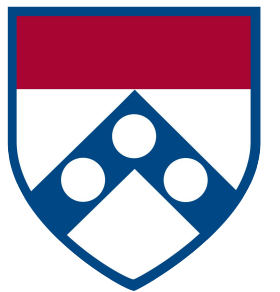
Session 1

Alex Miller

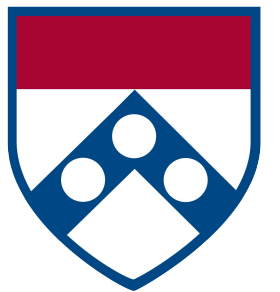
Ph.D. Student, Information Systems
Wharton, OID



What are we doing here?



What are YOU doing here?



Goals for the course

- Have a solid understanding of how to conduct empirical research
- Become familiar with tools available for empirical research
- Be able to collect raw data
- Have some idea of how more advanced methods can fit into your research process

Content of the Course

- First few sessions: lots of info
 - Intro, R, Python, HPCC
 - Could potentially be useful, even for people who are familiar with both R and Python
- Remaining sessions:
 - Collecting and working with data
 - API/databases
 - Crawling/Scraping
 - Text processing
 - NLP/Machine learning

Who is this course for?

- Mostly beginners
 - People who have some or little experience working in Python and R
- Kind of intermediate programmers
 - You may get less out of course, but I will try to focus on things you won't learn elsewhere
- Not so much for advanced programmers
 - Probably not much benefit from the course
 - Come to Session 3 (Friday) if you want a demonstration of how to use the Grid (High Performance Computing Cluster)

End goal



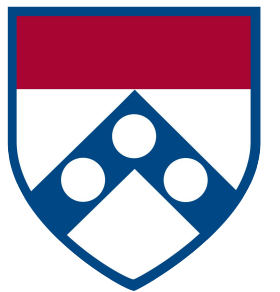
Format of the course

- 45-60 minutes of slides/lecture/demonstration
- 30-45 minutes of “lab” time
- This should minimize the amount of time you need to watch me typing into a terminal
- I will provide some brief exercises that will allow you to work on some skills covered during lecture

Teaching/Learning Philosophy

- I am going to throw a lot at you
 - My main goal is for you to know what is possible, i.e., what tools/functions exist and how they can make your life easier
 - Ultimately, you should focus on big picture things and use Google for everything else
- You **must** learn by doing
 - Use lab time to sit down and actually learn
 - I will probably tell you to do something that you don't know how to do: Use me, Google, Stack Overflow
- I will be experimenting with lab format. Bear with me.

Alex's Soapbox Time



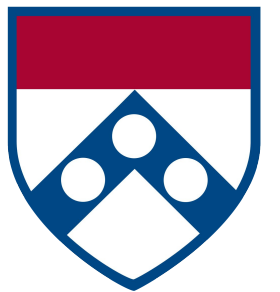
Unsolicited Advice for 1st Years

- Your PhD program does not have to be a horrible experience
 - You should (generally) enjoy what you're doing
- But things WILL be difficult at times
 - Academia is hard work, lots of rejection, very little feedback
- Prioritize your health and well-being!
 - Mental, physical, social
 - Take vacations!
 - Explore Philadelphia!
 - Develop a hobby outside of academics
- Find an advisor that you actually get along with
- When you become rich and famous, be nice to other people

Remainder of class

- Outline:
 - Big Picture
 - Unix/Git
 - R
 - Practice

The Big Picture



What is Empirical Research?

Empirical

- Working with **data**
 - Experimental
 - Observational/archival
- Almost all psych/behavioral work
- Typically **regression analysis**

Theoretical

Analytical

- Working with **math**
 - Based on economic models
 - Statistical theories

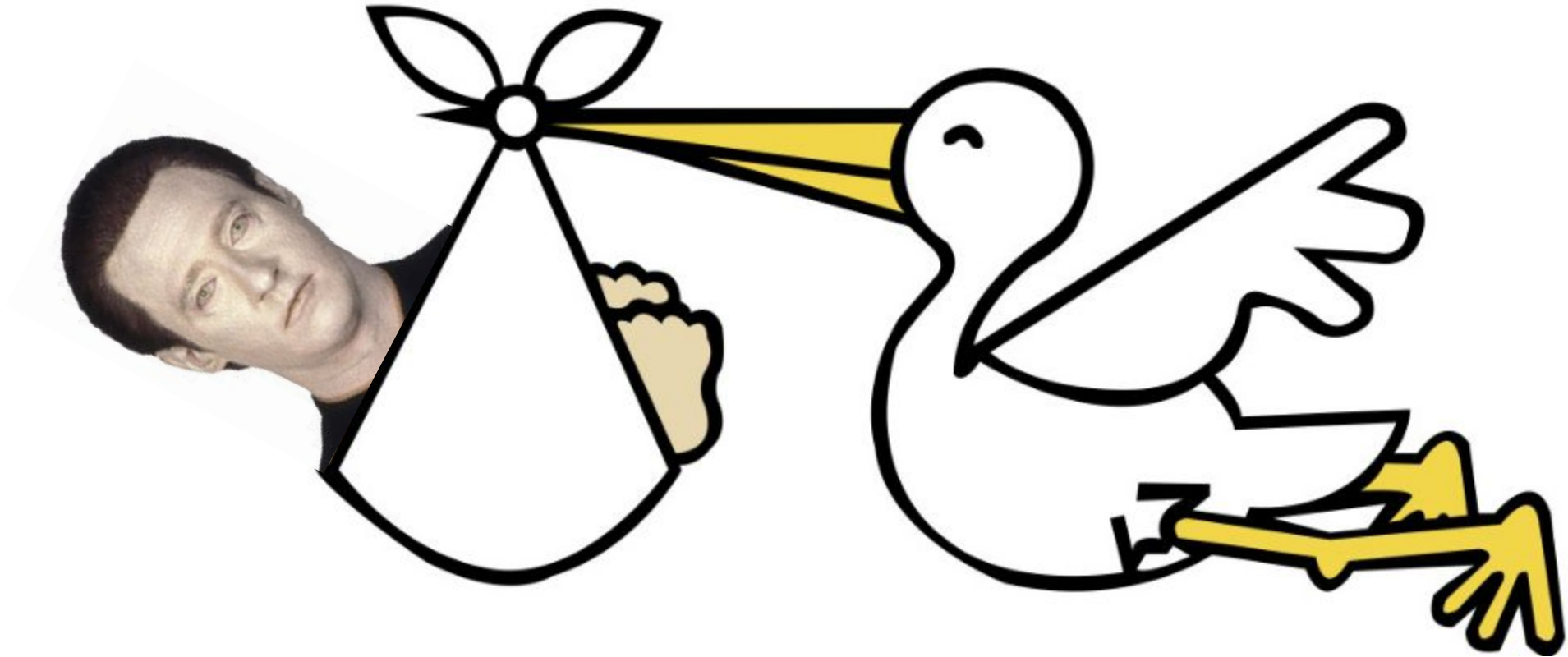
Simulations

Structural Modeling

Regression Analysis

- Trying to show (usually causal) relationship between two (or more variables)
 - Smoking \rightarrow Lung cancer?
 - College \rightarrow Higher income?
 - Chocolate and wine \rightarrow Happiness?
 - $X \rightarrow Y$ (fill in the blank for your field)
- To answer these questions, you need data

Where does data come from?



Where does data come from?

- Experiments
 - Wharton Behavioral Lab
 - Qualtrics/Amazon Mechanical Turk
- Existing Databases
 - Public: Census, Compustat, etc.
 - Private:
 - Paid access
 - Work directly with private company
 - Use Lippincott Library for data requests (seriously!!!)
- Simulations
 - The Grid (HPCC)
- Scraping

What do you do with data?

- Find it (see previous slide)
- Gather/save it
- Pre-process it
 - “cleaning”, “munging”
- Analyze it

What do you do with data?

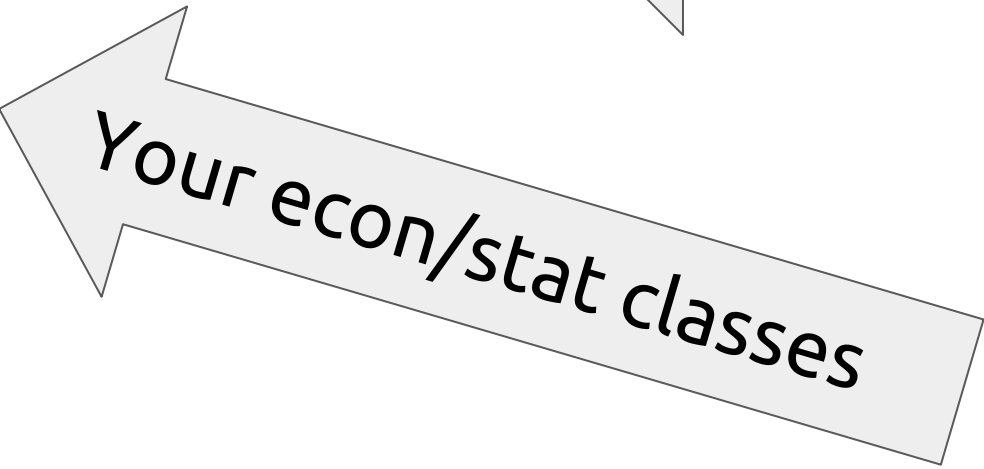
- Find it (see previous slide)
- **Gather/save it**
- **Pre-process it**
 - **“cleaning”, “munging”**
- Analyze it



Your entire PhD



This course



Your econ/stat classes

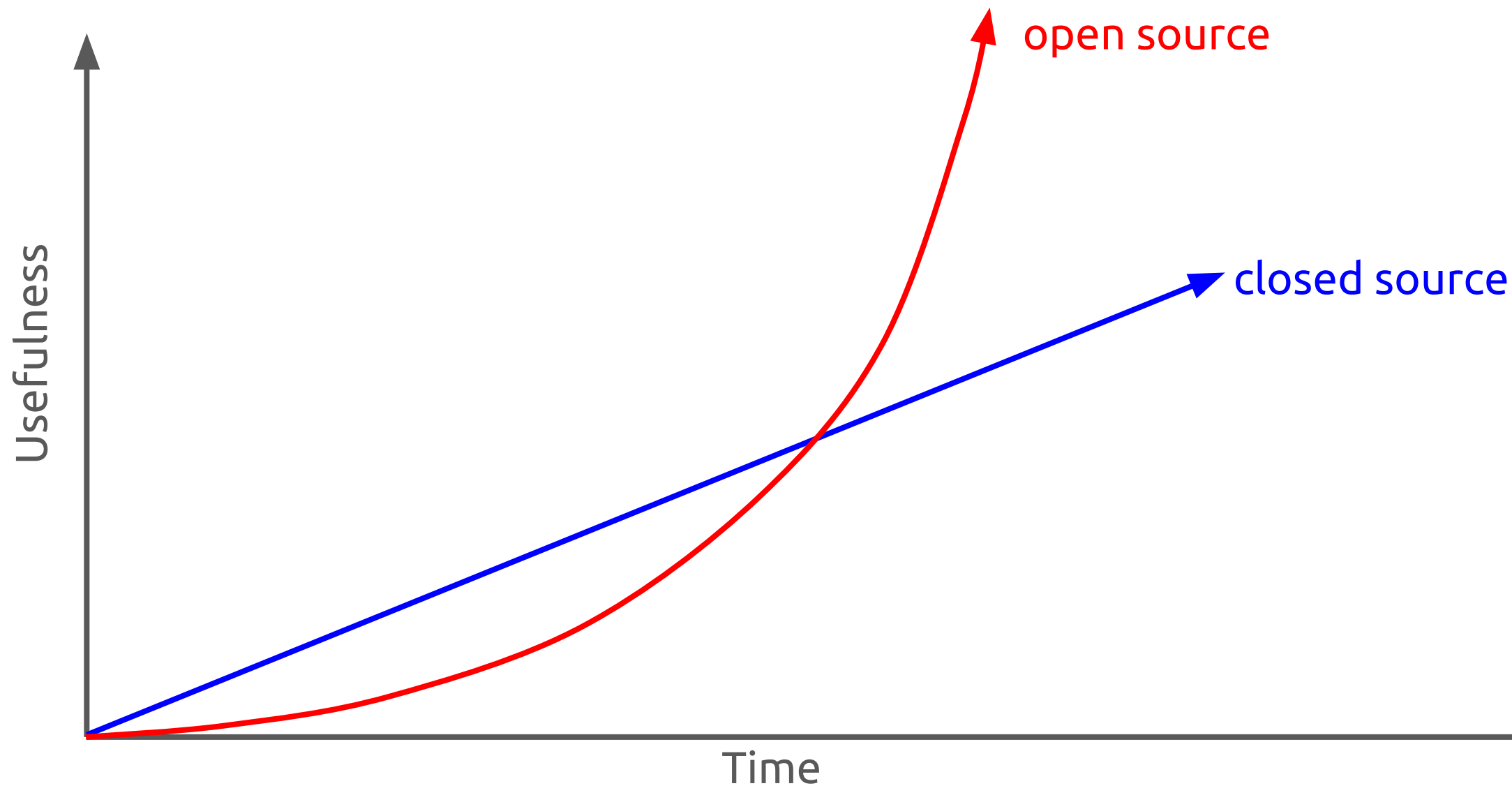
What tools do people use for this?



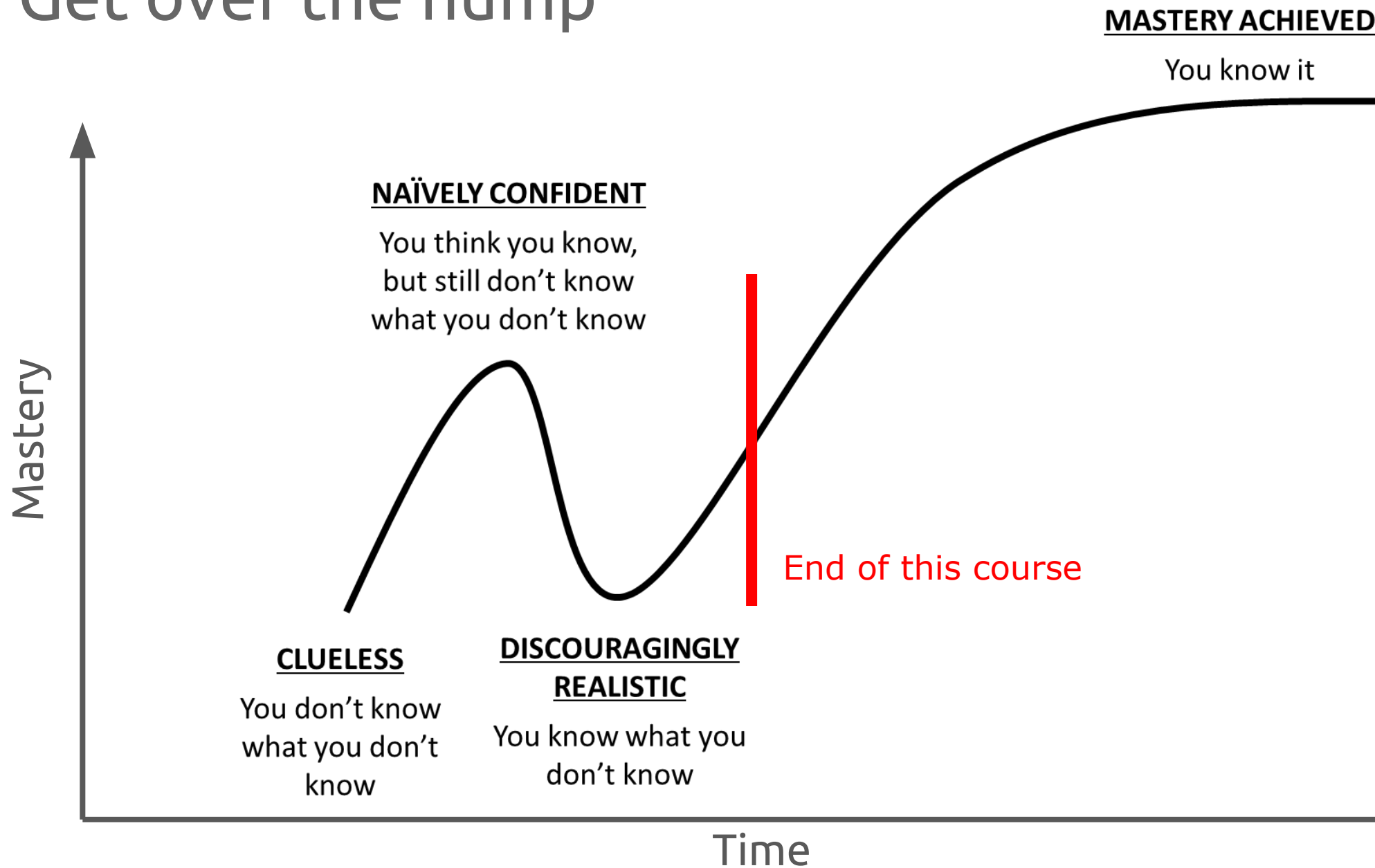
Why focus on Python and R?

- They are powerful, scripting languages
 - Designed to accomplish small tasks quickly
- These languages are open source
 - Leads to much more dynamic, state-of-the-art capabilities
- Most popular among younger and more technical academics
- However, there are tradeoffs
 - Learning curve
 - Documentation can be weaker for newer R/Python packages
 - Things are often in flux; there are many ways to do the same thing; not obvious which one to use

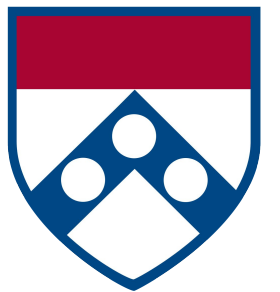
Over time, open source wins



Get over the hump



Command Line Basics



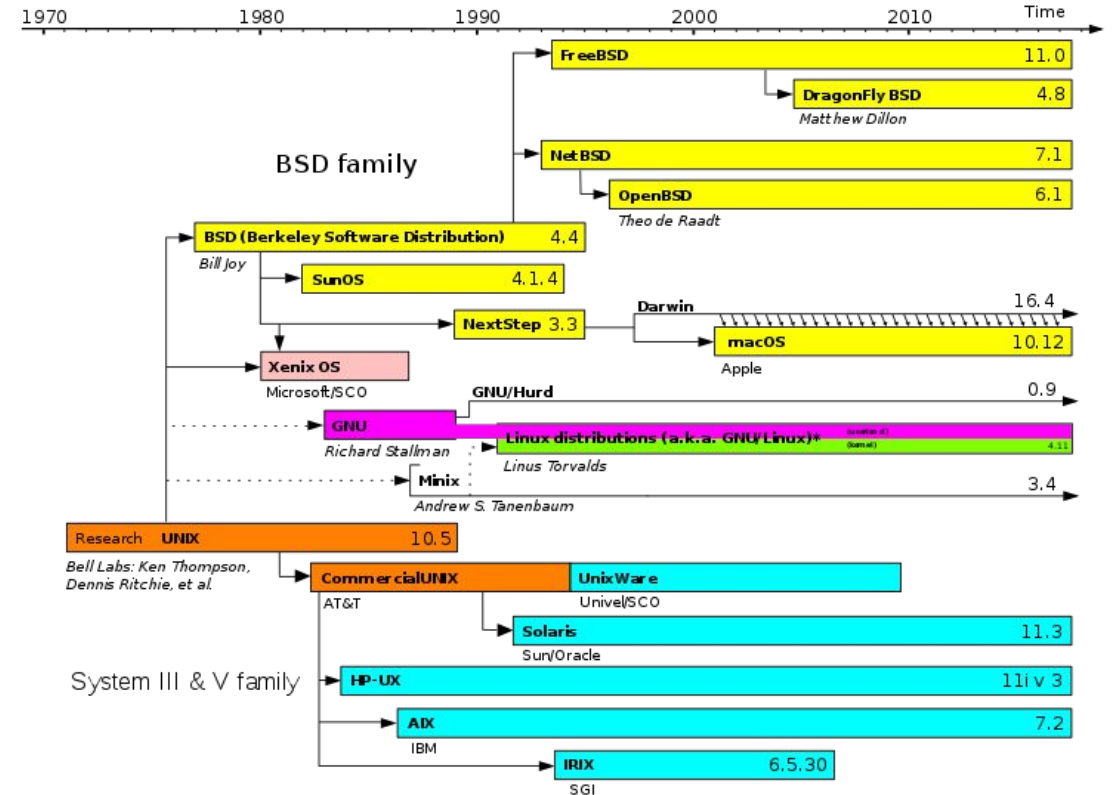
Is this really necessary?

- No, but it's very useful
- Understanding how to use your command line will make you:
 - More technically literate
 - A better programmer
 - Opens up doors to many, many cool things you can do with computers
 - A hacker
- Many modern packages/projects/systems require some familiarity with command line interface (CLI)



Intro: Unix

- Important things for you to know:
 - Unix is an old operating system developed in the 1970s at Bell Labs
 - It quickly spread in Computer Science and academic communities, forking into and inspiring many operating systems



*nix (“Unix-like” environments)

- Important “Unix-like” systems:
 - MacOS
 - Linux. OSes based on Linux kernel:
 - Ubuntu
 - Debian
 - CentOS
 - Android
- Important non-Unix-like systems:
 - Windows

Why should you care?

- Most modern programmers use *nix systems
- Most tutorials, guides, and documentation is for *nix systems
- Windows users should:
 - still be somewhat familiar with the Windows “Command Prompt”, but you can...
 - download a Unix-like system shell (i.e., [git bash](#))
- Mac users should:
 - Feel smug

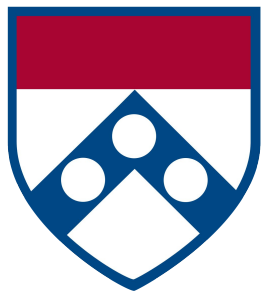
Just for fun

- Mac users:
 - Open “Terminal” and type:
 - “pwd” [enter] (prints working directory)
 - “ls” [enter] (lists files in directory)
- Windows users:
 - Open “Command Prompt” and type:
 - “cd” [enter] (prints current directory)
 - “dir” [enter] (lists files in directory)

Get a *nix-like shell on your machine

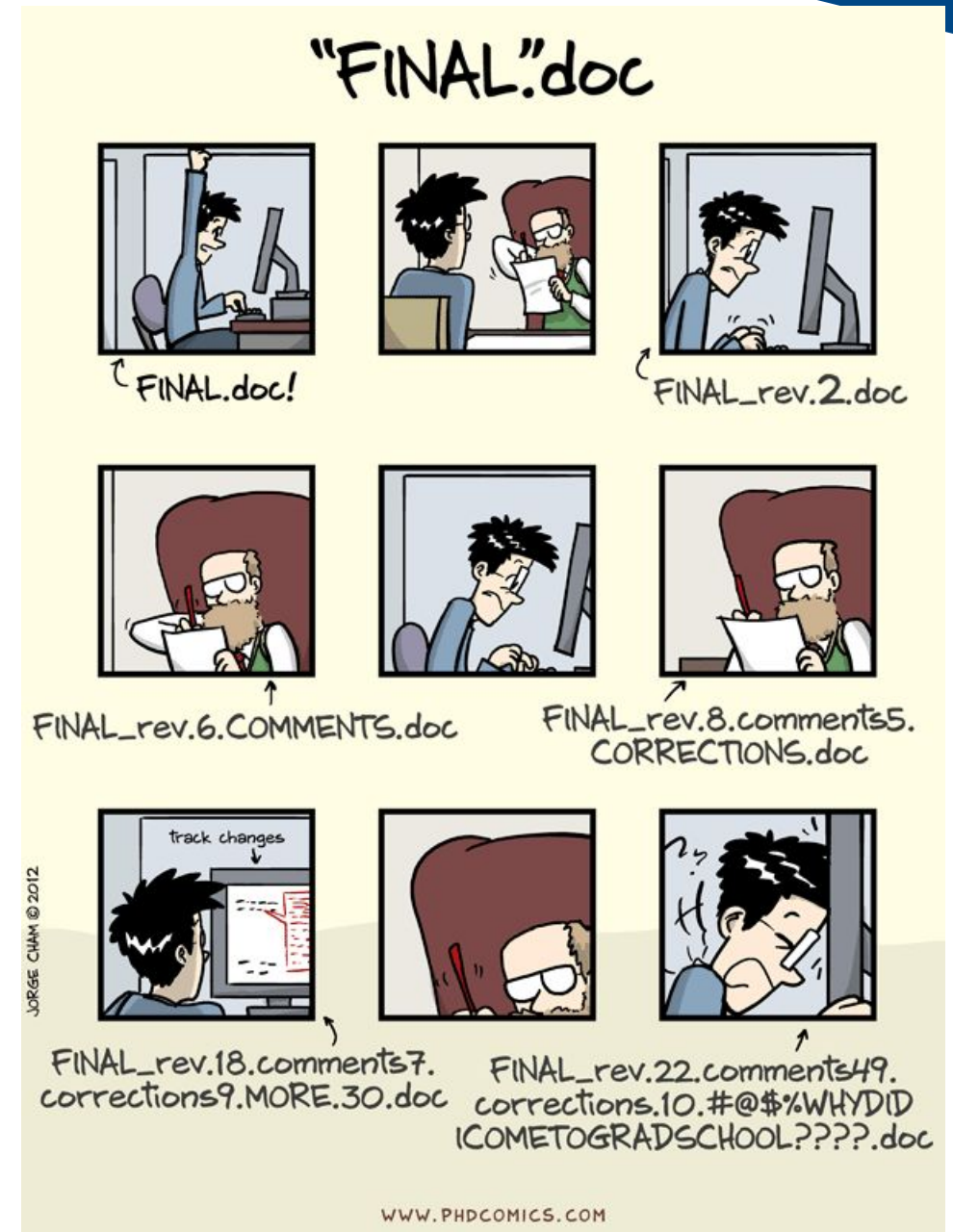
- Windows:
 - Download GitBash: <https://git-for-windows.github.io/>
- Windows 10:
 - You may be able to use “Bash on Ubuntu on Windows”, which is Microsoft’s attempt to join the future
- I can help you during the exercise portion of today’s lecture

Git, in brief



Git

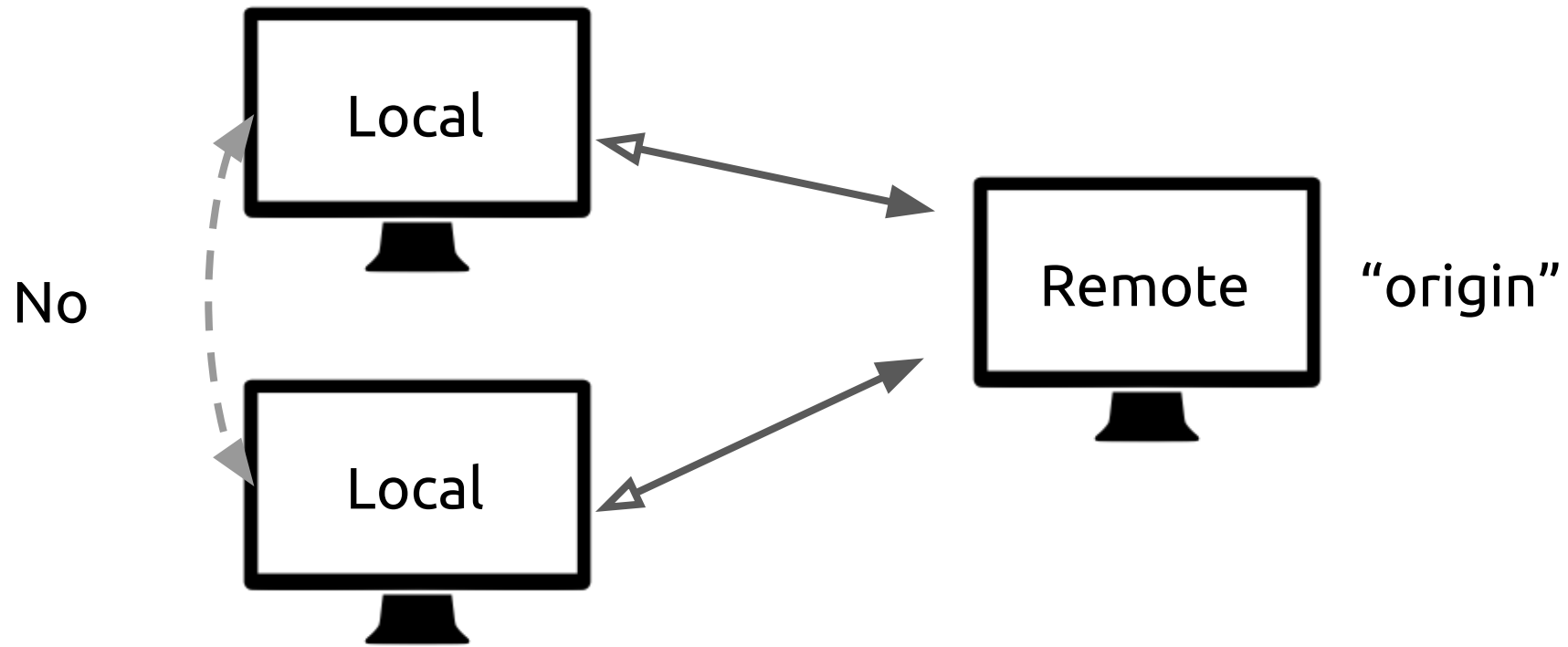
- Allows for efficient collaboration, tracking of changes, and edit history for *any* plaintext-based project
 - Very important tool for software engineering
 - Can be useful for academics, mostly for version history
- Generally good to be somewhat familiar with the basics



Git basics

- Projects in git are called “repositories”
 - Consists of set of files/subfolders that are all part of the same code-base
- You typically have a “local” repo and a “remote” repo
 - But you don’t have to!
- Remote destinations are useful for backing up
- The most popular repository of repositories is GitHub
 - Essentially a storage service, with some neat features on top of git
 - Bitbucket is also free and has unlimited private repos

How to use git



- This system ensures there is a “master” version of your project
- Also makes it easy to share code, even with non-collaborators

Basics

```
$ cd myrepo/
```

```
$ git init      # initializes repository
```

```
$ git add --all  # adds files to "staging area"
```

```
$ git commit -m "my first commit" # saves changes
```

```
$ git log
```

- Will have chance to practice basics during exercise portion

Basics: Working with Remotes

```
$ git remote add origin [remote URL]
```

```
$ git remote -v # list remotes
```

```
$ git push origin master # push local to remote
```

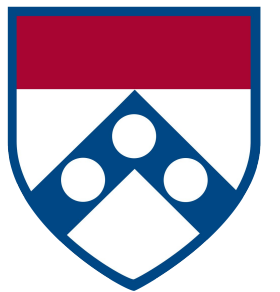
<https://help.github.com/articles/adding-an-existing-project-to-github-using-the-command-line/>

- Will have chance to practice basics during exercise portion

Last notes

- Important!!!!!!!
 - Dealing with merge conflicts is VERY frustrating
 - Do your best not to edit the same file in two different places
 - If you do screw up, use StackOverflow:
 - <https://stackoverflow.com/questions/161813/how-to-resolve-merge-conflicts-in-git>
- RStudio has automatic version control with git built in
 - Will go over this later

R Basics



R

- R is an open-source programming language, based on “S”, a language developed by statisticians at Bell Labs
- RStudio is an IDE based around the R language
 - Importantly, RStudio != R
- Greatest strengths:
 - “Vector-native” calculations
 - Works very naturally with vector/matrix type data
 - R has LOTS of packages for many different stats applications
- Tradeoff/Weakness:
 - Does lots coercing/manipulation without telling you!

Do you need to know R?

- You should be familiar with it
- If other languages are common in your field (e.g., STATA, SAS), you will probably be fine without being an R expert
 - But! Know that R is becoming more popular
- Almost every applied course in the STAT department uses R
- Will only cover basics here so everyone has some experience

This is the boring part

- Unfortunately, R has a lot quirks which you can really only appreciate by seeing them worked through
- More so than other languages, R relies on lots of built-in functions with names you might not be familiar with
- If you are already familiar with R, go ahead and start the exercises (link at Tech Camp website)
 - <http://opim.wharton.upenn.edu/techcamp/2017/>
 - https://github.com/alexmill/techcamp_week1
 - Do the advanced exercise!

R Basics

- Basic Variable Types
 - Double, Integer, Character, Logical
- Compound Variable types
 - Vector, List, Factor, Matrix, DataFrame
 - Indexing/selecting
- Functions
- For Loops
- Braces/spacing: [Reference](#)
- Packages
 - How to discover? Experience + Google!
- Plots

Important things you won't learn elsewhere

- R Markdown and R Notebooks
 - Use them!
 - (Please!) Never copy your code into a word document again!
 - Even better: Use the `stargazer` or `texreg` packages for reproducible tables!

If we have time today

- [tidyverse](#)
 - Important packages: dplyr, ggplot2, readr, stringr
 - Learn it! It will make your life easier
- [data.table](#)
 - Alternative to data.frames
 - Makes it easier to manipulate data; but not easier than dplyr
 - Slightly faster performance on big data than dplyr

Exercises

Follow instructions at:

https://github.com/alexmill/techcamp_week1