Get the Materials!

- GitHub Repo:
 - https://github.com/datasciencedojo/meetup/t ree/master/r programming excel users

- Kaggle Titanic Competition:
 - https://www.kaggle.com/c/titanic



Intro to R Programming for Excel Users

May 3rd, 2017



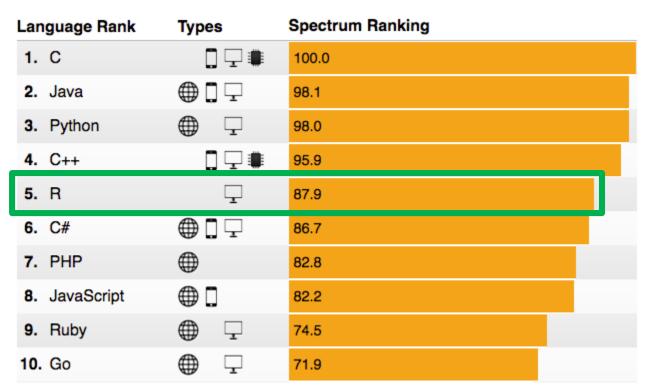
Who Am I?

- Dave Langer, VP of Data Science Data Science Dojo
- 20+ years in technology:
 - Roles in development, architecture, & BI/DW/analytics.
 - Last job Sr. Director, BI & Analytics @ Microsoft.
- Hooked on Data Science 5 years ago:
 - Extensive background in data and analytics.
 - Learned Machine Learning from 2nd place Netflix Prize winner.
 - More tutorials on my YouTube channel!
- Joined Data Science Dojo to democratize Data Science.



Motivation

The IEEE's 2016 Ranking of most Popular Programming Languages



R has experienced rapid YoY increases in popularity.

This is remarkable as R is a specialized language for data and analytics!

http://spectrum.ieee.org/computing/software/the-2016-top-programming-languages



Expectation Setting

- I am assuming the following:
 - You are experienced with Excel tables, formulas, functions & pivot tables.
 - You are not familiar with R.
 - You are interested in learning R.
- This is a quick intro to R using Excel as a framework:
 - I will gloss over a lot of things.
 - I will illustrate some "art of the possible".
 - More in-depth coverage is available on my YouTube channel.
- My goal is to make you confident and excited about learning R!

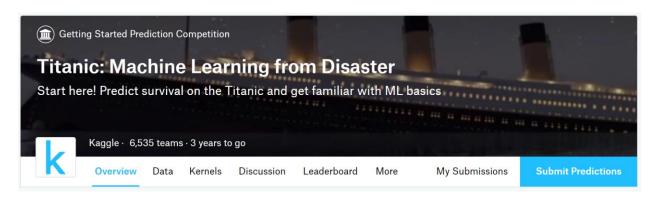


Prerequisites

- To follow along you will need the following:
 - Excel
 - R
 - RStudio
- The following R packages are required to follow along:
 - ggplot2 and dplyr
- The GitHub repo has source, data, and slide files.



The Data



Why use this dataset?

data science for everyone

1. Everyone is familiar with the problem domain.

 It is a good proxy for common business data – for example, customer profile data.

The Data

Data Dictionary

Variable	Definition
survival	Survival
pclass	Ticket class
sex	Sex
Age	Age in years
sibsp	# of siblings / spouses aboard the Titanic
parch	# of parents / children aboard the Titanic
ticket	Ticket number
fare	Passenger fare
cabin	Cabin number
embarked	Port of Embarkation

Key

$$0 = No, 1 = Yes$$

$$1 = 1st, 2 = 2nd, 3 = 3rd$$

C = Cherbourg, Q = Queenstown, S = Southampton



The Scenario

 We've been asked to analyze the Titanic data in terms of finding patterns of passengers that survived vs. those that did not.

- This analysis has common analogies in business for example, customer churn analysis.
- We will use common analytical activities in Excel and then illustrate the same in R.



LET'S LEARN SOME R!



QUESTIONS



APPENDIX



Get the Materials!

- GitHub Repo:
 - https://github.com/datasciencedojo/meetup/t ree/master/r programming excel users

- Kaggle Titanic Competition:
 - https://www.kaggle.com/c/titanic



- De facto standard visualization package in R.
- Designed for print-quality graphics.



 Fine-grained control via an API focusing on layering graphical elements to build visualizations.



The collection of data that we're working with.

The aesthetic – how data is mapped to visuals.

```
Main function – Sou the starting point.

rary(g) (ot2)
```

The aesthetic – how data is mapped to visuals.

```
3 ta("iri/")
```

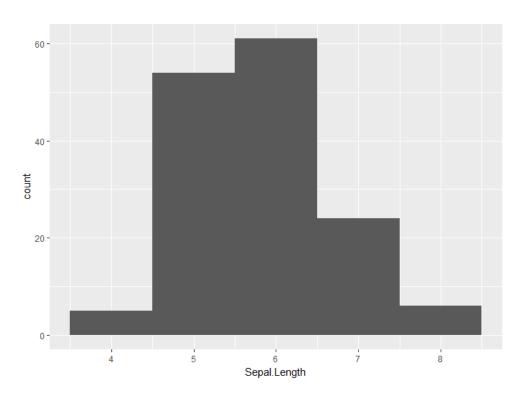
ggplot(iris, aes(x = Sepal.Length)) +
geom_histogram(binwidth = 1)

A visual to layer on to the canvas.

Visual parameters

– control the look.

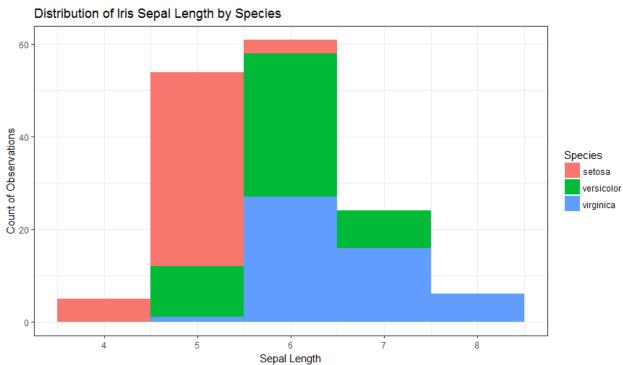




Incrementally build more powerful visualizations

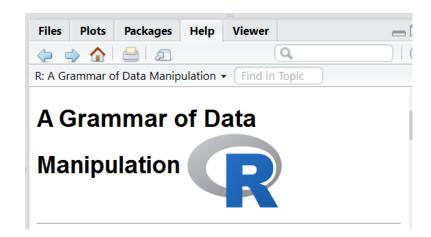
```
library(ggplot2)
  data("iris")
4
  ggplot(iris, aes(x = Sepal.Length, fill = Species)
    theme_bw() +
6
    geom_histogram(binwidth = 1) +
    labs(x = "Sepal Length",
         y = "Count of Observations",
         title = "Distribution of Iris Sepal Length by Species")
10
```





dplyr

- Popular package for data wrangling.
- Intuitive coding style based on data pieplines.



 Very intuitive to those with familiarity with SQL – just about everything you can do with SQL you can do with dplyr.



```
dplyr
```

The data we're working with.

Pipe data to the next operation.

```
library(dplyr)
14
                                      Filter data down.
15
16
   iris.stats <- iris %>%
17
     filter(Species == "setosa"
                                                 Group into categories.
18
             Species == "virginica"
     group_by(Species) %>%
19
20
      summarize(Sepal.Length.Min = min(Sepal.Length),
21
                Sepal.Length.Max = max(Sepal.Length),
22
                Sepal.Length.Mean = mean(Sepal.Length),
23
                Sepal.Length.Median = median(Sepal.Length).
24
                Sepal.Length.SD = sd(Sepal.Length))
   View(iris.stats)
```

Summarize groups.



dplyr

⇔ ⇒ I I T Filter							
	Species [‡]	Sepal.Length.Min	Sepal.Length.Max	Sepal.Length.Mean	Sepal.Length.Median	Sepal.Length.SD	
1	setosa	4.3	5.8	5.006	5.0	0.3524897	
2	virginica	4.9	7.9	6.588	6.5	0.6358796	

