

# Intro to R Visualization with Power BI

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# 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 2<sup>nd</sup> place Netflix Prize winner.
  - #1 Data Scientist on YouTube.
- Joined Data Science Dojo to democratize Data Science.

# Motivation

- Power BI is core to Microsoft's analytics strategy:
  - Sustained Engineering investment.
  - Cloud, On-Premises, and Hybrid.
  - Growing community of contributors.
- Support for R Visualizations!

# Why Power BI?

- Highly productive environment for:
  - Data exploration and analysis.
  - Executive Scorecards.
  - Operational Dashboards.
- Excellent solution for descriptive analytics.
- Power BI Desktop is free\*!

# Expectation Setting

- I am assuming the following:
  - You are familiar with Power BI
  - You are familiar with R
  - You prefer R over DAX
- This is not a Power BI tutorial.
- This is not a R programming tutorial.
- We will cover practical aspects of using R visualizations from within Power BI.

# Prerequisites

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

# The Scenario

- You are an Analyst for Wide World Importers and have been asked to conduct an analysis on WWI Customers.
- You have access to the WWI Database.
- You have access to Power BI, but are more comfortable using R than DAX in your work.

# THE DATA



# POWER BI DESKTOP

# USING R WITH POWER BI

# Gotchas

- Power BI limits data to R at 150,000 rows.
- Power BI automatically drops duplicate rows:
  - Leveraging PKs is a good idea!
- Power BI allows for very permissive column names.

# Gotchas

Duplicate Data!

Make sure you have no more than 150,000 rows!

May want to remove space.

Make records unique!

Question Number	Response	User ID
1	3	A
1	4	B
1	3	C
1	4	D
1	5	E
2	3	A
2	3	B
2	4	C
2	5	D
2	5	E

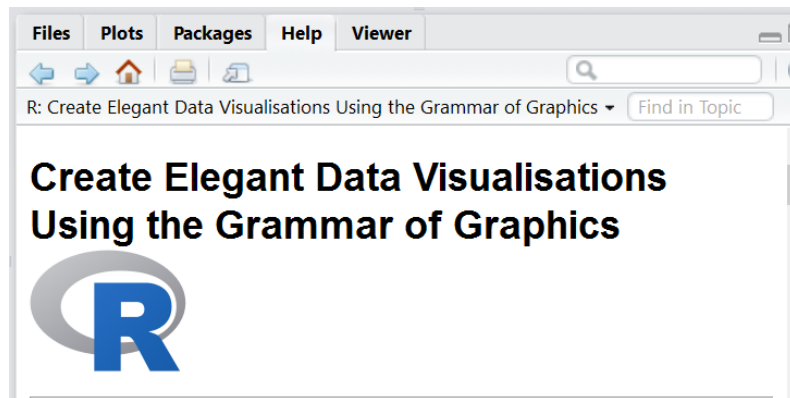
# Some Best Practices

- Prefer native Power BI visuals:
  - <https://app.powerbi.com/visuals/>
- Change columns names to be R-friendly.
- Use R Studio, Visual Studio R Tools, etc. to develop your visuals:
  - Coding & debugging R in Power BI is painful.

# GGPLOT2 & DPLYR

# ggplot2

- De facto standard visualization package in R.
- Designed for print-quality graphics.
- Fine-grained control via an API focusing on layering graphical element to build visualization.



# ggplot2

Main function –  
the starting point.

The collection of  
data that we're  
working with.

The aesthetic –  
how data is  
mapped to visuals.

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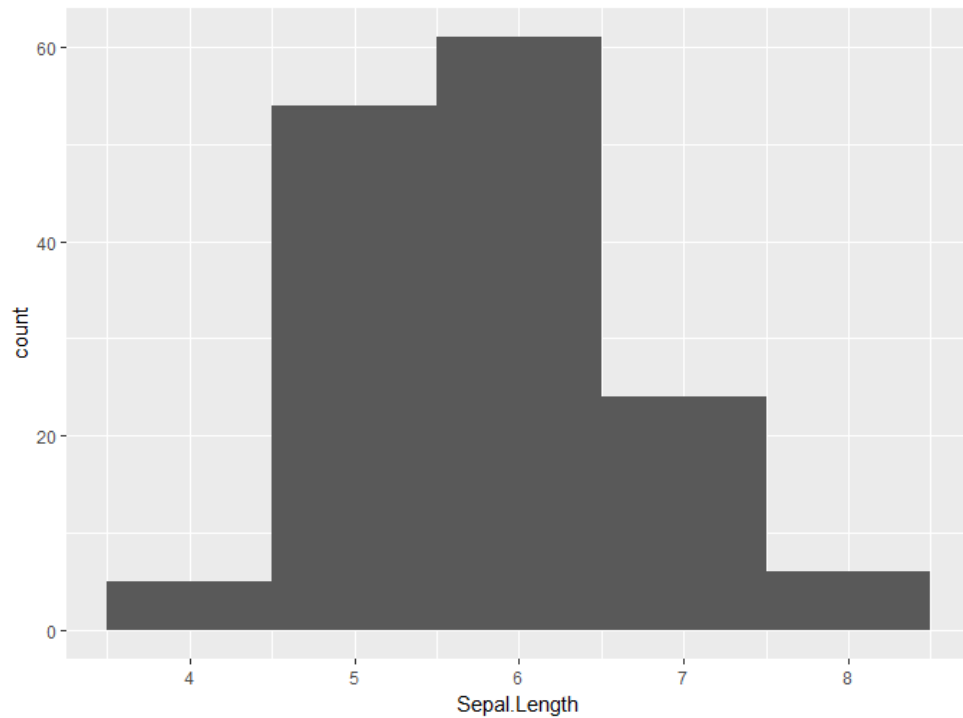
```
library(ggplot2)
data("iris")
ggplot(iris, aes(x = sepal.Length)) +
  geom_histogram(binwidth = 1)
```

A visual to layer on  
to the canvas.

Visual parameters  
– control the look.



# ggplot2

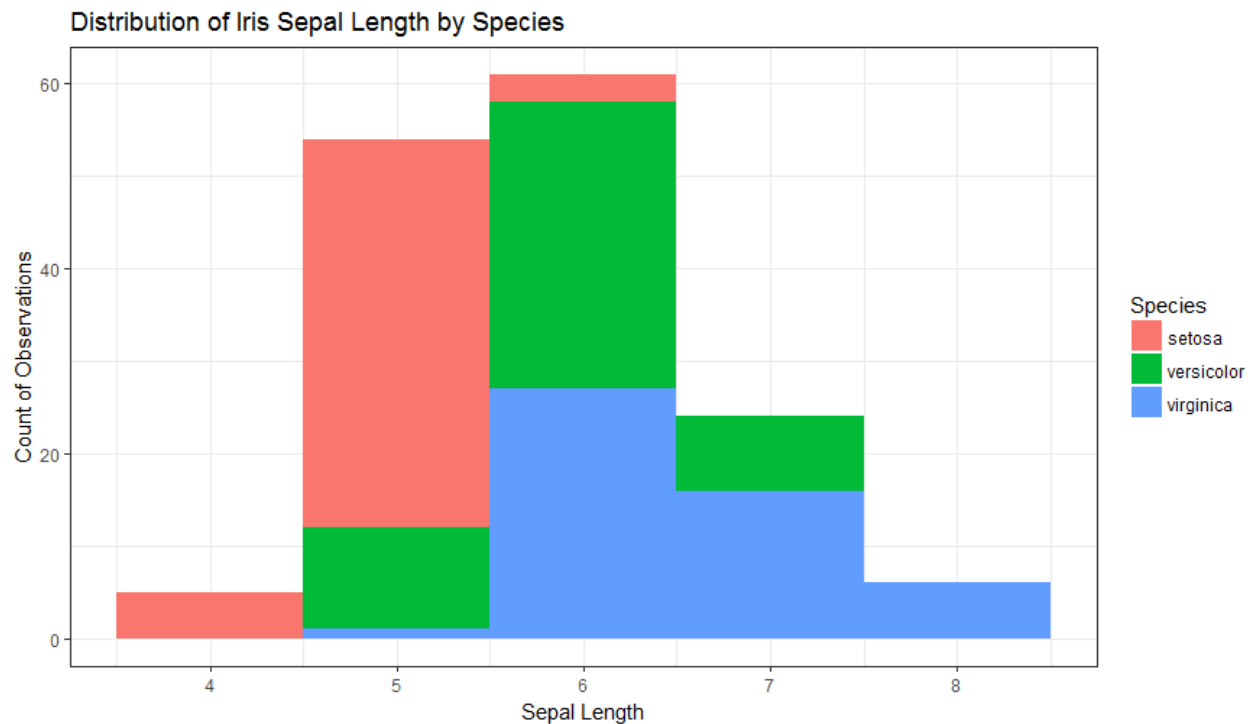


# ggplot2

Incrementally build  
more powerful  
visualizations

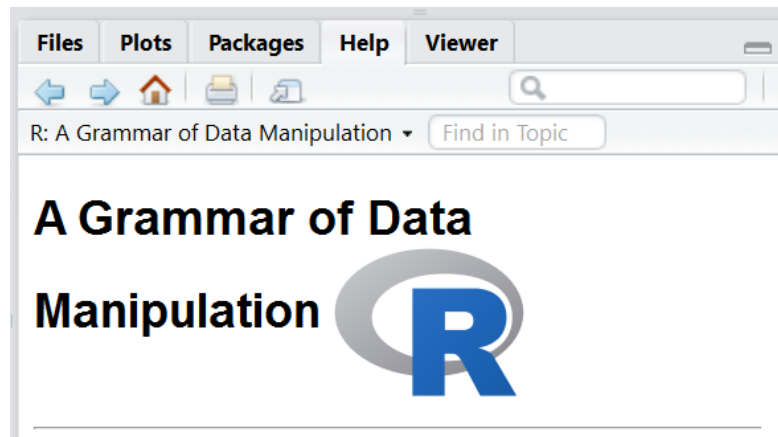
```
1 library(ggplot2)
2
3 data("iris")
4
5 ggplot(iris, aes(x = sepal.Length, fill = species)) +
6   theme_bw() +
7   geom_histogram(binwidth = 1) +
8   labs(x = "sepal Length",
9        y = "Count of observations",
10        title = "Distribution of Iris sepal Length by species")
```

# ggplot2



# dplyr

- Popular package for data wrangling.
- Similar to Pig – data flows and pipelines.
- 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.


Filter data down.

Group into categories.

```
14 library(dplyr)
15
16 iris.stats <- iris %>%
17   filter(species == "setosa" |
18          species == "virginica") %>%
19   group_by(species) %>%
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))
25 view(iris.stats)
```

Summarize groups.

# dplyr

 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

# R CODE!

# QUESTIONS



# APPENDIX

# Get the Files!

- GitHub Repo:

- [https://github.com/datasciencedojo/meetup/tree/master/r\\_visualization\\_with\\_power\\_bi](https://github.com/datasciencedojo/meetup/tree/master/r_visualization_with_power_bi)

- World Wide Importers DB:

- <https://github.com/Microsoft/sql-server-samples/releases/tag/wide-world-importers-v1.0>