

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 2nd 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

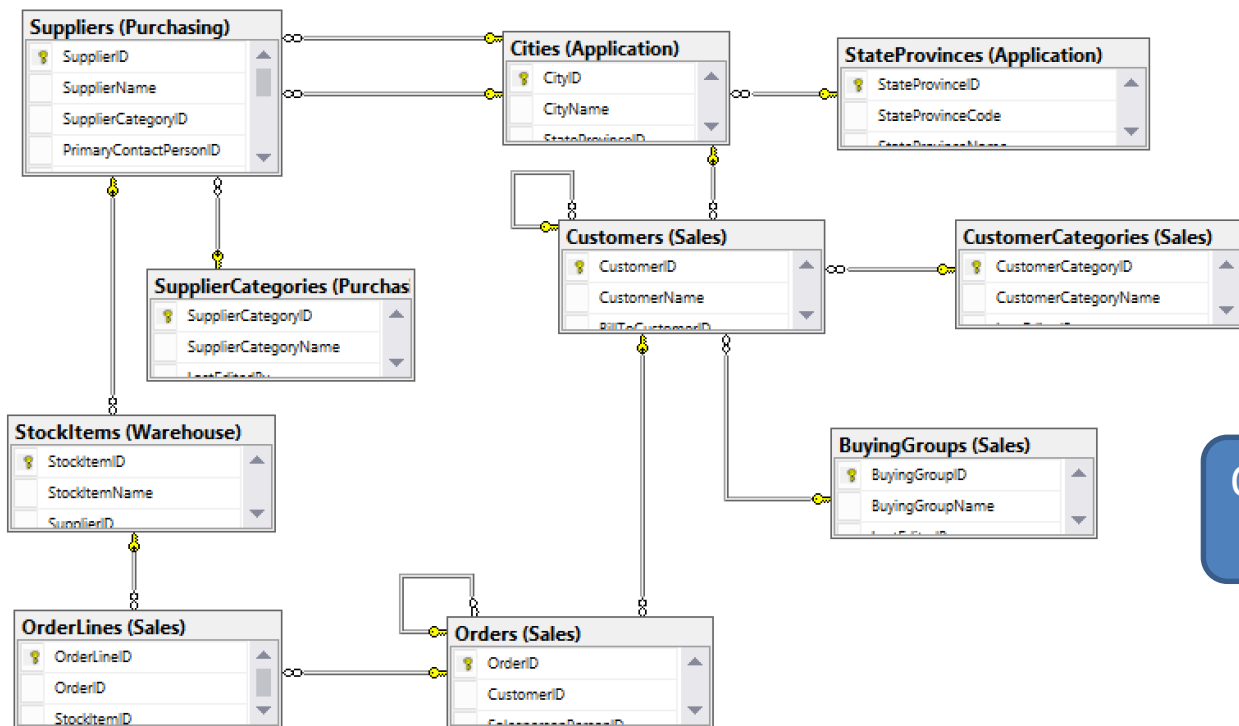
- To follow along you will need the following:
 - Power BI Desktop
 - R
 - RStudio
- The following R packages are required to follow along:
 - dplyr, lubridate, ggplot2, scales, qcc
- The GitHub repo has source, data, and slide files.

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

The Database



Normalized schema,
common for
transactional systems.

Option #1 – Pull all
tables into Power BI.

Option #2– Use SQL query
to denormalize 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.

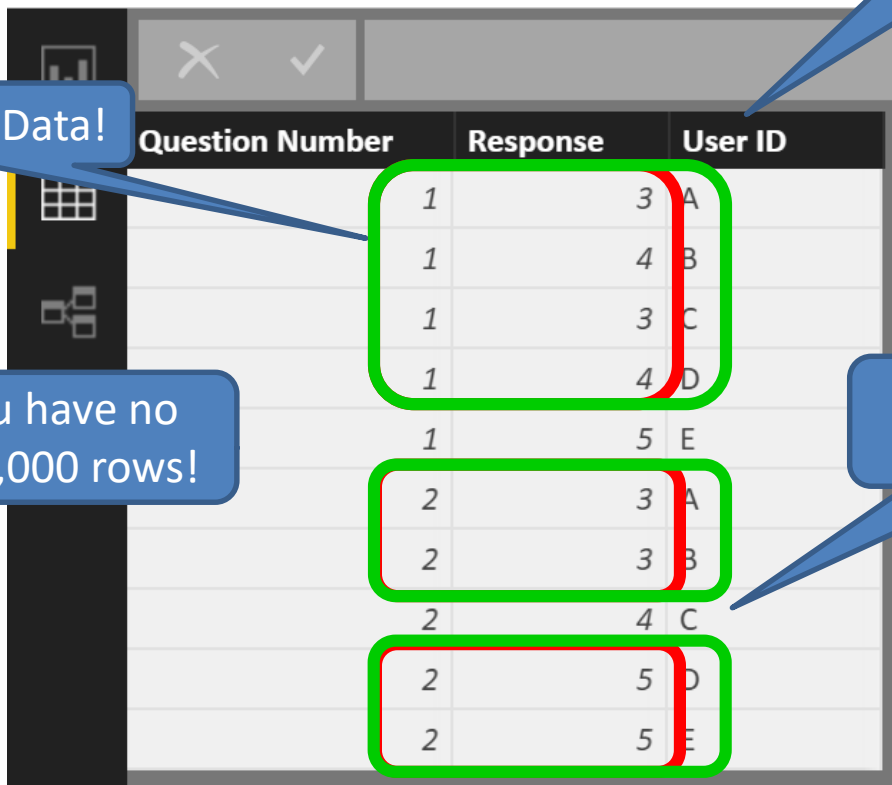
Gotchas

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 Power BI Slicers to reduce data sent to R.
- Use R Studio, Visual Studio R Tools, etc. to develop your visuals:
 - Coding & debugging R in Power BI is painful.
 - Write your R code to work with Power BI.

Power BI's Interface with R

The name of your data frame is "dataset"!

R script editor

⚠ Duplicate rows were removed from the data.

```
# Create dataframe
```

```
# dataset <- data.frame(BuyingGroupName, CustomerCategoryName, LineTotal, Year, Month, OrderLineID, SupplierCategoryName)
```

```
# Remove duplicated rows
```

```
# dataset <- unique(dataset)
```

Paste or type your R-script code here

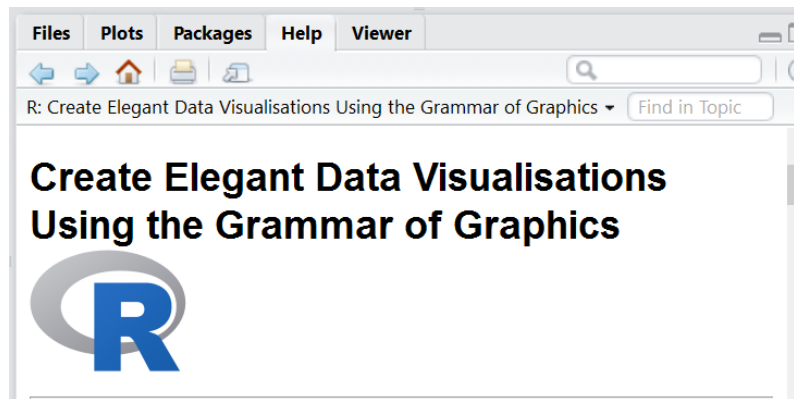
Copy and paste your R code here!

The names of your variables!

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 elements to build visualizations.



ggplot2

Main function –
the starting point.

The collection of
data that we're
working with.

The aesthetic –
how data is
mapped to visuals.

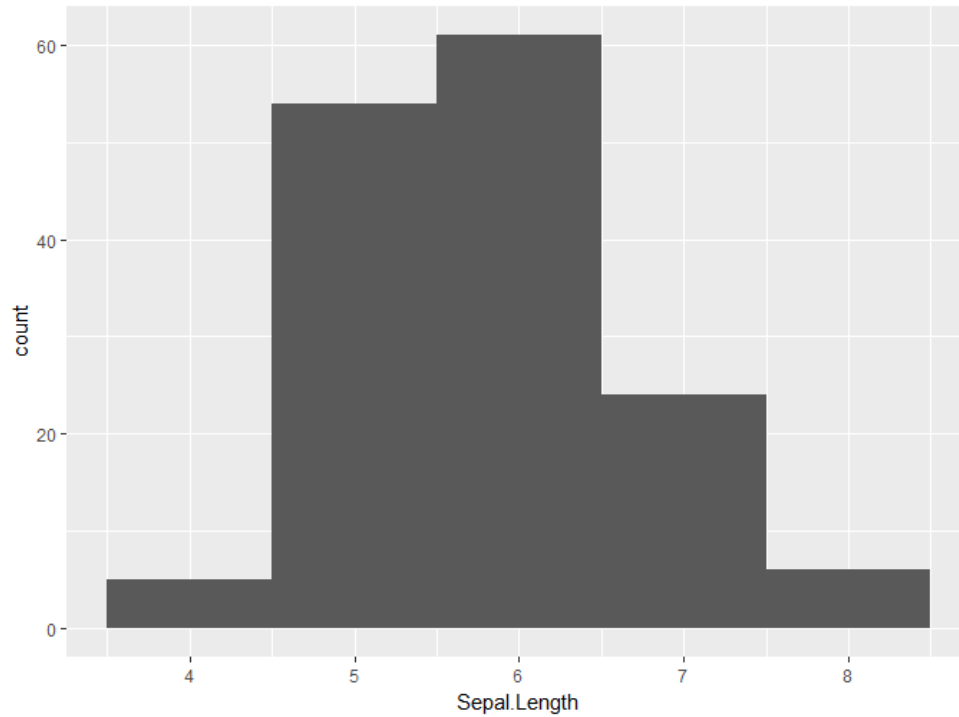
The aesthetic –
how data is
mapped to visuals.

```
library(ggplot2)
2
3 data("iris")
4
5 ggplot(iris, aes(x = sepal.Length)) +
6   geom_histogram(binwidth = 1)
```

A visual to layer on
to the canvas.

Visual parameters
– control the look.

ggplot2



ggplot2

Incrementally build
more powerful
visualizations

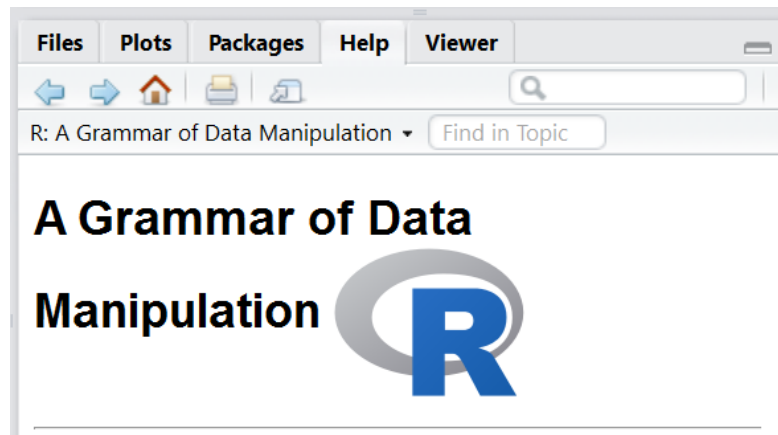
```
1 library(ggplot2)
2
3 data("iris")
4
5 ggplot(iris, aes(x = sepal.Length, fill = species)) +
6   theme_bw() +
7   geom_histogram(binwidth = 1) +
8   labs(x = "sepal Length",
9        y = "Count of observations",
10       title = "Distribution of Iris sepal Length by species")
```

ggplot2



dplyr

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

```
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)
```

Filter data down.

Group into categories.

Summarize groups.

dplyr

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