Lecture 12: Exploratory Data Analysis - Part 2

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July 14, 2016

On the Agenda

- 1. Administrative Issues
 - ► Midterm (Part 2)
 - ▶ Group Project Update due on 7/19/16 at 11:59 PM CDT.
- 2. Graphing Systems in R
 - ▶ Base R, lattice, ggplot2
- 3. Exploratory Data Analysis
 - Visual Techniques

Midterm (Part 2)

- ▶ 10 minutes to do Midterm (Part 2)
- No notes or collaborating!
- ▶ 5 Points of E.C. up for grabs!

Group Project Update

- Help me, help you by letting me know about your group's project status
- Please answer:
 - 1. How is the project progressing?
 - 2. What has been accomplished thus far?
 - 3. What have you learned?
 - 4. What issues have arisen?
- Avoid showing me code in the report.
- Score for the progress report is based on how much work has been completed since the project proposal was initially submitted.

Moving along...

▶ We're going to cover **Graphing in** *R* next!

Graphing in R

- Before now, we never really focused on plotting.
- ▶ Instead, we aimed to understand the computing logic behind calculations in *R*.
- ▶ Now, to support visual *EDA*, we really need to start focusing on such features.

R and the Three Graphing Systems

- ▶ Dilemma: There are **3** graphing systems to chose from in *R*.
- ▶ Similar to the Goldilocks and the Three Bears problem.



Figure 1:

Selecting the graphical system is important...

Different Plotting Systems in *R* - Overview

- R's Base plotting system.
 - plot(), hist(), barplot()
- ▶ lattice formulaic graphing in R.
 - xyplot(), dotplot(), histogram(), *plot()
- ggplot2 rapid layered graphing approach
 - graphs start with ggplot() and add layers via + typically denoted by geom_point(), geom_*()

R's Base Plotting System - Example

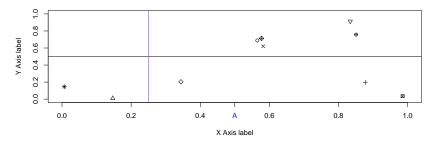
View it as an artists blank canvas

```
plot(NULL, xlim=c(0,1), ylim=c(0,1),
       ylab="Y Axis label", xlab="X Axis label")
Y Axis label
   9.0
   0.0
                     0.2
                                 0.4
        0.0
                                              0.6
                                                           0.8
                                                                        1.0
                                     X Axis label
```

R's Base Plotting System - Example

Each subsequent function calls adds lines, points, axis, et cetera.

```
x = runif(10); y = runif(10)
abline(h = .5)  # Horizontal Line
abline(v = .25, col="purple") # Vertical Line w/ color
points(x, y, pch = 1:10)  # Points w/ shapes
axis(1, .5, LETTERS[1], col.axis = "blue")
```



R's Base Plotting System - Verdict

- Con: No ability to change plot settings (e.g. ?par settings) or draw content added once started.
- ► **Pros:** Easier custom graphs and higher quality graphs (e.g. AVLR using the tikzDevice package)
- Verdict: Academics only



Credit: The Global Warmers

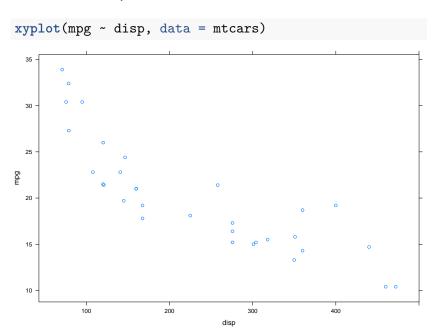
lattice - A formulaic approach to graphs.

- ► The lattice package written by Deepayan Sarkar provides the ability to make graphs in *one* call vs. Base R's multiple calls.
- ▶ The call form is normally:

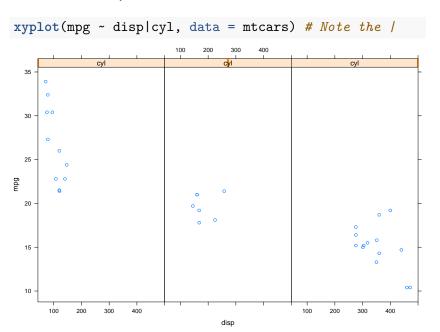
```
type_of_plot(formula, data=list())
```

- Uses the formula object associated with lm to specify: response (y~), explanatory (~x), conditional relationships (y~x|A).
- Great for viewing conditional relationships and multivariate data.

lattice - Example



lattice - Example with Condition



lattice - Verdict

- ▶ **Cons:** Everything in 1 function call is *messy* and awkward.
- Pros: Handle all margin settings of multiple graphs and conditioning.
- Verdict: Casual R users.



Credit: The Global Warmers

ggplot2 - Grammar of Graphics

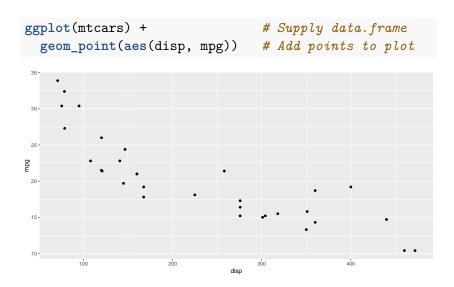
- ggplot2 is the implementation of the pivotal 1999 Book Grammar of Graphics by Leland Wilkinson.
 - Each Graph shares a common structure.
 - The difference between graphs is different component layers and rules.

ggplot2 - Grammar of Graphics

- Historical Information
 - ggplot1 written by Hadley Wickham as part of his PhD thesis.
 - ggplot2 released for ease of use alongside A Layered Grammar of Graphics
 - UseR 2016 Keynote: ggplot1 is better than ggplot2 API wise due to the piping operator

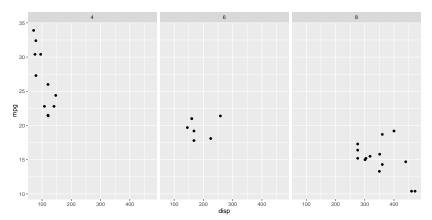
► Time: 36:38 to 38:32

ggplot2 - Scatterplot



ggplot2 - Scatterplot Conditioned

```
ggplot(mtcars) +  # Supply data.frame
geom_point(aes(disp, mpg)) + # Add points to plot
facet_wrap(~cyl)  # Write conditioning
```



ggplot2 vs. Base R

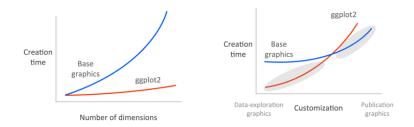


Figure 2: Credit: Sean C. Anderson

ggplot2 - Verdict

- Cons: Data must be in a data.frame, global scoping of variables, data copies, and simple things might be complex.
- Pros: Rapidly iterate visualizations, grammatical structure, and extendable graphing system.
- ▶ Verdict: Data Scientists, Researchers, and Causal R users.



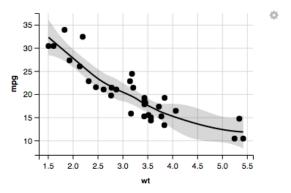
I lied...

- ► As is the case with technology, there always a new graphing system around the corner.
- ► Coming Soon a 4th Graphical System for *R* using the parts of ggplot2....

ggvis - Coming Soon (TM)

▶ Introducing ggvis, the successor to ggplot2...

Scatterplot with smooth curve and interactive control:



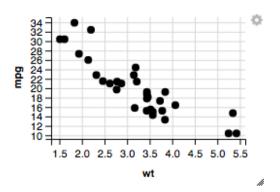


ggvis - Coming Soon (TM)

- ggvis will usher in a new way of interactive graphics (think identify()). Keep an eye on this project.
- Alas, as the API is currently in flux, we will not dedicate any time to it. However, note that ggvis replaces ggplot2's concatenation with the %>% operator.

ggvis - Example

```
# install.packages("ggvis")
library("ggvis")
mtcars %>% ggvis(~wt, ~mpg) %>% layer_points()
```



Note: This code can only be run in *HTML* rich environments. No LaTeX environments need apply.

Moving Along . . .

- Any questions on R's Plotting Systems?
- ► Next up... Visual EDA!

Visual EDA

"Use a picture. It's worth a thousand words"

— Tess Flanders in Speakers Give Sound Advice

Visual EDA

Data Wrangling birth from msos package.

For the next section, I'll aim to use the birth data from msos. Note, the data is in **wide** form in a matrix. The below script sets up the data for graphing by converting it to **long** form and class data.frame.

```
# Extract hospital birth dates
data(births, package="msos")

library("tidyr")
df_births = as.data.frame(births)
df_births$time = seq_len(nrow(df_births))
long_births = gather(df_births, hospital, value, -time)
```

Looking into long_births

Let's peek at what the data in long_births looks like.

head(long_births)

```
## time hospital value
## 1 1 Hospital1 13.56
## 2 2 Hospital1 14.39
## 3 3 Hospital1 14.63
## 4 4 Hospital1 14.97
## 5 5 Hospital1 15.13
## 6 6 Hospital1 14.25
```

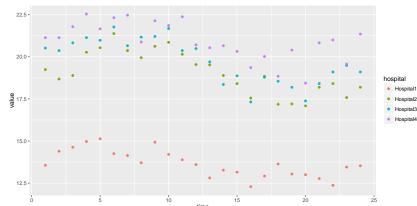
Note: long_births is of class data.frame!!

Your first ggplot!

```
ggplot(long_births) +
                                # Initialize qqplot w/ data
  geom_point(
                                  # Add a point layer
    aes(x = time, y = value)# Add an aesthetic mapping
 22.5 -
 20.0 -
9 17.5 -
 15.0 -
 12.5 -
                                                    20
                                        15
                                 time
```

Your second ggplot!

```
ggplot(long_births) + # Initialize ggplot w/ data
geom_point(  # Add a point layer
aes(x = time, # Add an aesthetic mapping
    y = value,
    color = hospital)# Added color
    )
```



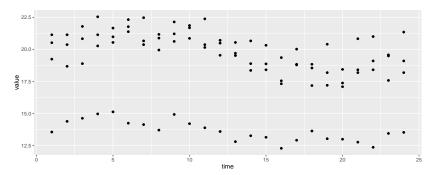
Key Terms and Ideas with ggplot2

- ggplot: Initialization function of the graph
- ▶ **geom_**: Geometric (shape) objects
- ▶ aes: Provides the aesthetic options the geom should take.
 - Examples: color, fill, transparency (alpha), linetype, and point shape.
- scales: Axis kind
 - **Examples:** Continuous, Discrete, log, $\sqrt{\ }$, and so on.
- facet: Panel layout
 - **Examples:** Grid $(x \times y)$ or Wrapped

Reusing ggplot2 base objects

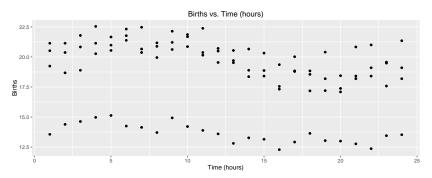
Each ggplot2 object can be saved individually and added to in the future

```
g = ggplot(long_births) +
  geom_point(aes(x = time, y = value))
g
```



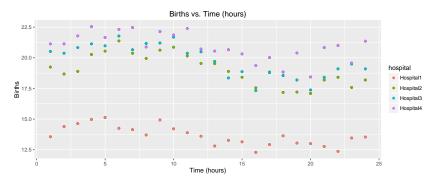
Adding Label Information to ggplot2

```
(g = g + xlab("Time (hours)") + ylab("Births") +
  ggtitle("Births vs. Time (hours)"))
```

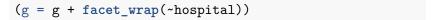


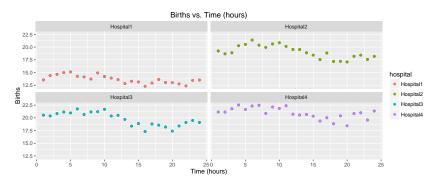
Changing aes for geom_point

(g = g + geom_point(aes(x = time, y = value, color = hospi



Adding a facet_wrap to distinguish variables





Graphing with ggplot2

- ggplot2 makes available various geometric objects via geom_.
- ▶ These objects determine how the data is rendered on the plot.
- ► Some of the geoms_*() typically used:

| geoms_*() | Description |
|-----------------------------|----------------------------------|
| <pre>geom_point()</pre> | Adds data points to plot |
| <pre>geom_line()</pre> | Adds connected lines to the plot |
| <pre>geom_histogram()</pre> | Makes a histogram |
| <pre>geom_bar()</pre> | Creates a bar chart |
| <pre>geom_text()</pre> | Adds text annotations |
| <pre>geom_violin()</pre> | Makes a violin plot |
| | |

► Many more geoms_*() exists and can be found at docs.ggplot2.org with graphing examples!

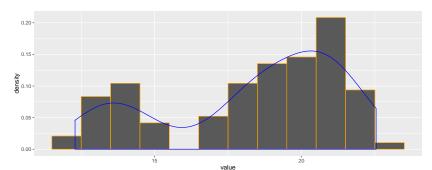
Making a histogram

Histograms typically provide count or frequency values on the y-axis

```
ggplot(long_births) +
  geom_histogram(aes(value), binwidth = 1)
 20 -
 15 -
10 -
 5 -
                                 value
```

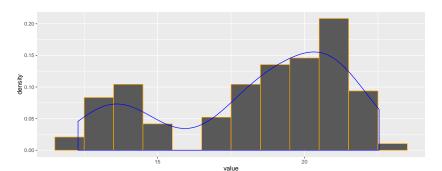
Making a histogram with a density plot

Density plots alongside a histogram require density (bounded between 0 and 1) to be on the y-axis. If count is on the y-axis, then results are not valid.



Storing aes in construction

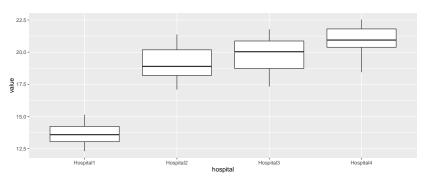
Specify aes() does not necessarily have to be done in the geom_*() call. Some users prefer to specify the relationship in the ggplot() creation.



Boxplot

Boxplot are a helpful way to visualize Q1, Q2, Q3, and outlier information. They are may be referred to as a *box and whiskers* plot.

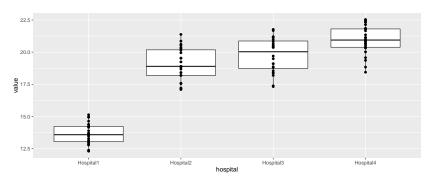
```
ggplot(long_births, aes(x = hospital, y = value)) +
  geom_boxplot()
```



Boxplot with Points

What is nice, is instead of only seeing outliers, you can also see where all the points lie just by adding geom_point()

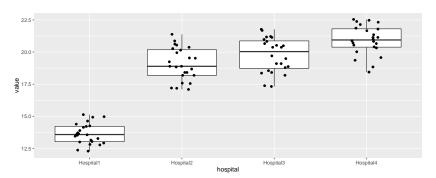
```
ggplot(long_births, aes(x = hospital, y = value)) +
  geom_boxplot() + geom_point()
```



Boxplot with Points Redux

However, adding points without *jittering* them will lead to non-informative clumping. To avoid this, use a jitter: geom_jitter()

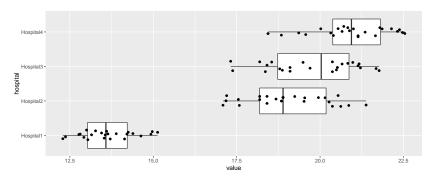
```
ggplot(long_births, aes(x = hospital, y = value)) +
  geom_boxplot() + geom_jitter(height = 0, width = 0.4)
```



Flipping My Box

The coordinate system can also change from being y-based to x-based via coord_flip().

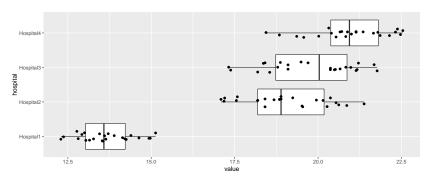
```
ggplot(long_births, aes(x = hospital, y = value)) +
  geom_boxplot() + geom_jitter(height = 0, width = 0.4) +
  coord_flip()
```



Flipping My Box

The coordinate system can also change from being y-based to x-based via coord_flip().

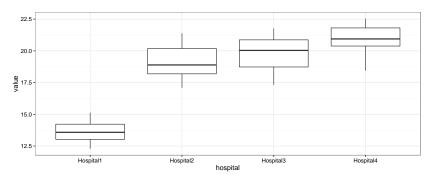
```
ggplot(long_births, aes(x = hospital, y = value)) +
  geom_boxplot() + geom_jitter(height = 0, width = 0.4) +
  coord_flip()
```



Theming - Black & White

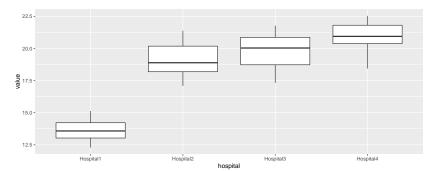
ggplot2 can easily switch to different color themes. By default, the theme_gray() is used. Some prefer the theme_bw().

```
ggplot(long_births, aes(x = hospital, y = value)) +
  geom_boxplot() + theme_bw()
```



Theming - Original

```
ggplot(long_births, aes(x = hospital, y = value)) +
  geom_boxplot()
```



The main question:

Are you team theme_gray() or theme_bw()?

Twitter Question

Exercises

- Load sportsranks from msos and transform it to long form.
 Make sure to add an indicator. Try to create boxplots of the different ratings.
- Open states in the msos data set. Explore the different levels of school enrollment and crime. Try out other dimensions as well!
- 3. Last but not least, try to explore the SAheart data in msos.

Practical Comparison

For the next example, we are going to craft a Q-Q Plot in Base R and ggplot2. The Q-Q plot is normally used to check to see if the residuals of a model follow a normal distribution.

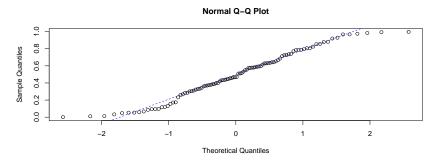
```
# Set seed for reproducibility
set.seed(111)

# Generate data
x = runif(100,0,1)
```

Traditional q-q plot in R

Included in Base R, without any modification is qqnorm(), which provides the Q-Q Plot. Though, it is missing the traditional line connecting the first and third quartiles.

```
qqnorm(x) # normal q-q plot
qqline(x,lty=2,col="blue") # line through the Q1 and Q3 que
```



Crafting the q-q plot

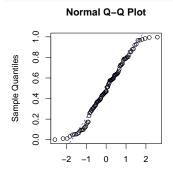
```
qqn = function(w) {
 n = length(w)
  nv = qnorm((1:n)/(n+1)) # Quantiles of Normal Dist.
  plot(nv, sort(w), # X, Y
       xlab = "Theoretical Quantiles",
       ylab = "Sample Quantiles")
 title("Normal Q-Q Plot")
  m = (quantile(w, 0.75) - quantile(w, 0.25))/
      (qnorm(0.75)-qnorm(0.25))
  b = quantile(w, 0.25) - m*qnorm(0.25)
  abline(b, m, lty=2, col="red") # Line through Q1 & Q3
```

Comparing Base implementations

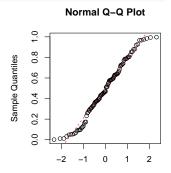
```
par(mfrow=c(1,2))  # Two plots in one window
par(pty="s")  # Square plots

qqnorm(x)  # Base R first
qqline(x,lty=2,col="blue")

qqn(x)  # Our Plot
```



Theoretical Quantiles

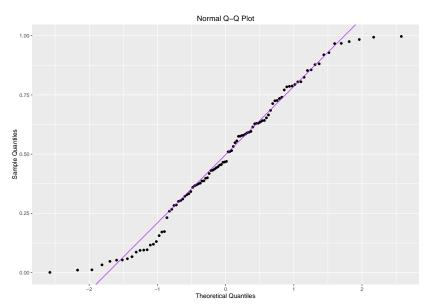


Theoretical Quantiles

ggplot2 implementation

```
df_x = as.data.frame(x)
n = nrow(df x)
m = (quantile(x, 0.75) - quantile(x, 0.25))/
      (qnorm(0.75)-qnorm(0.25))
b = quantile(x, 0.25) - m*qnorm(0.25)
g = ggplot(df_x, aes(sample=x)) +
      stat qq() +
      geom abline(intercept = b,
                  slope = m,
                  color = "purple") +
      xlab("Theoretical Quantiles") +
      ylab("Sample Quantiles") +
      ggtitle("Normal Q-Q Plot")
```

ggplot2 implementation



ggplot2 implementation

g + theme_dark() # Welcome to the dark side!

