MPP-E1180 Lecture 9: Automatic Tables and Static Visualisation

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Objectives for the week

- Assignment 3
- Review
- Static results presentation
 - Automatic table creation
 - Plotting best practices
 - ggplot2 for general graphing
 - Zelig simulations for showing results

Assignment 3

Purpose: Gather, clean, and analyse data

Deadline: 13 November 2015

You will submit a GitHub repo that:

- ► Gathers web-based data from at least **two sources**. Cleans and merges the data so that it is ready for statistical analyses.
- Conducts basic descriptive and inferential statistics with the data to address a relevant research question.
- Briefly describes the results including with dynamically generated tables and figures.
- ► Has a write up of 1,500 words maximum that describes the data gathering and analysis and uses literate programming.

Review

- What is the basic R syntax for a regression model?
- What is a model function? What two parts do GLM model functions have?
- ► How do you find a 95% confidence interval for a parameter point estimate (both mathematically and in R)?
- What is one good way to interpret and present results from a logistic regression to both a statistical and general audience?

Motivation

Today we will learn how to **communicate your research findings** with automatically generated tables and static plots.

Why automatically generate?

- Saves time: don't have to re-enter numbers by hand into a table or restyle a graph each time you change the data/analysis.
- Easier to find and correct errors: all source code that created all tables and figures is linked and output updated when corrections are made.
- ▶ More reproducible: everything is clearly linked together.

General process

In general include the functions to create the tables/figures in a code chunk.

Include in the **code chunk head** echo=FALSE, warning=FALSE, error=FALSE, message=FALSE.

You may need to also include results='asis' for some table functions.

See previous weeks 4 and 5 for figure code chunk options.

Automatic table generation

There are a number of tools for automatically generating tables in R/R Markdown.

- kable in the knitr package
- xtable package
- texreg package
- stargazer package

Today

We will focus on kable and stargazer.

- kable is a good, simple tool for creating tables from data frames (or matrices).
- stargazer is useful for creating more complex tables of regression model output.

Example Docs:

Hertie Data Science / Examples / Paper With Regression Tables

kable example: predicted probabilities

Set up from Lecture 8:

```
# Load data
URL <- 'http://www.ats.ucla.edu/stat/data/binary.csv'</pre>
Admission <- read.csv(URL)
# Estimate model
Logit1 <- glm(admit ~ gre + gpa + as.factor(rank),
              data = Admission, family = 'binomial')
# Create fitted data
fitted <- with(Admission.
               data.frame(gre = mean(gre),
                           gpa = mean(gpa),
                           rank = factor(1:4))
```

kable example: predicted probabilities

| gre | gpa | rank | predicted |
|-------|--------|------|-----------|
| 587.7 | 3.3899 | 1 | 0.5166016 |
| 587.7 | 3.3899 | 2 | 0.3522846 |
| 587.7 | 3.3899 | 3 | 0.2186120 |
| 587.7 | 3.3899 | 4 | 0.1846684 |

kable example: predicted probabilities

You can stylise the table.

Table 2:Predicted Probabilities for Fitted Values

| gre | gpa | rank | predicted |
|-------|------|------|-----------|
| 587.7 | 3.39 | 1 | 0.52 |
| 587.7 | 3.39 | 2 | 0.35 |
| 587.7 | 3.39 | 3 | 0.22 |
| 587.7 | 3.39 | 4 | 0.18 |

Note

Don't showing more digits to the right of the decimal than are statistically and substantively meaningful.

A rule of thumb: more than one or two digits are rarely meaningful.

See also: http://andrewgelman.com/2012/07/02/moving-beyond-hopeless-graphics/

Show regression output with stargazer

kable is limited if we want to create regression output tables, especially for multiple models.

stargazer is good for this.

stargazer example

Estimate models

stargazer example HTML

When you are creating a table for an HTML doc with stargazer use:

stargazer example HTML

stargazer example PDF

When you are creating a PDF use the arguments:

```
type = 'latex'
header = FALSE
```

stargazer output in PDF

Showing Results with Figures

Tables **are important** to include so that readers can **explore details**, but are **usually not the best way** to show your results.

Figures are often more effective.

General principles

(A Selection of) Tufte's Principles for Excellent Statistical Graphics (2001, 13):

- Show the data
- Encourage the eye to compare differences in the data
- Serve a clear purpose
- Avoid distorting the data
- Be closely integrated with the text

Show the data

Show the data, not other things like silly graphics or unnecessary words.

Have a high data ink ratio:

$$\mathrm{Data}\,\mathrm{Ink}\,\mathrm{Ratio} = \frac{\mathrm{data} - \mathrm{ink}}{\mathrm{total}\,\mathrm{ink}}$$

Encourage the eye to compare differences

How did the budgets change? (Orange is 2013, Blue is 2012)

A little better

Even better

Avoid distorting the data: special case circles

In general: Avoid using the **size** of a circle to mean something! So, avoid:

- bubble charts
- pie charts

Why avoid circles?

Circles can distort data.

- It is difficult to compare their size.
- ▶ The Ebbinghause Illusion!

Order the circles from smallest to largest.

The circles are on a scale of 0-100, so what are their values?

Order the circles from smallest to largest.

The circles are on a scale of 0-100, so what are their values?

Ebbinghause Illusion

Which circle is bigger?

Colours and Data Distortions

Which square is darkest?

Colour Principles

Only give graphical features (e.g. bars in a bar chart) different colours if it **means something** in the data.

Colour Principles

Colours should be used to:

- highlight particular data,
- group items,
- encode quantitative values
 - Values of continuous variables should be represented using increasing hues of the same colour.
 - Categorical variables should be represented with different colours. (rule of thumb: avoid using more than about 7 colours in a plot)

Bad

Good

Colours and accessibility

Color Blindness

People who are colour blind can have difficulty distinguishing between **red-green** and **blue-yellow**.

About 5-8% of men are colour blind.

We need to choose colour schemes for our graphics that are **colour blind friendly**.

For more information see http://www.usability.gov/get-involved/blog/2010/02/color-blindness.html.

Selecting colors

Color Brewer is a great resource for selecting colours: http://colorbrewer2.org/.

A more systematic introduction to ggplot2

"gg" means "Grammar of Graphics".

"2" just means that it is the second one.

ggplot2 syntax

Each plot is made of **layers**. Layers include the coordinate system (x-y), points, labels, etc.

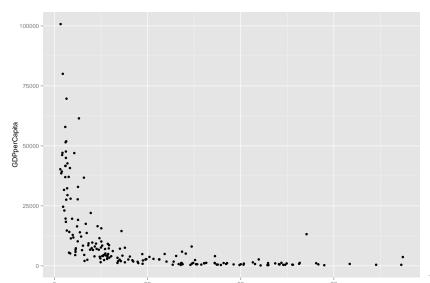
Each layer has **aesthetics** (aes) including the x & y, size, shape, and colour.

The **main layer types** are called **geometrics** (geom). These include lines, points, density plots, bars, and text.

ggplot2 examples setup

```
library(devtools)
library(ggplot2)
source_url("http://bit.ly/OTWEGS")
# Create data with no missing values of infant mortality
InfantNoMiss <- subset(MortalityGDP,</pre>
                            !is.na(InfantMortality))
# Create High/Low Income Variable
InfantNoMiss$DumMort[InfantNoMiss$InfantMortality
                     >= 15] <- "high"
InfantNoMiss$DumMort[InfantNoMiss$InfantMortality
                      < 15] <- "low"
```

Simple example



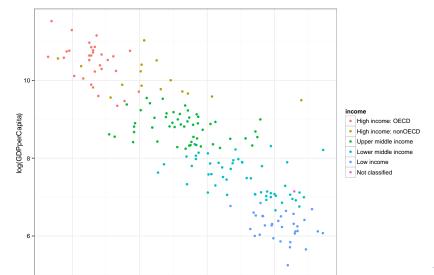
Simple example with BW theme

```
ggplot(data = MortalityGDP, aes(x = InfantMortality,
         y = GDPperCapita)) + geom_point() + theme_bw(base_s
 100000
  75000
3DPperCapita
 25000
    0 -
```

Colours

There are a number of ways to specify colours in ggplot2.

The simplest way is to let ggplot choose the colours for you.



Selecting colours

There are many ways to pick specific colors.

In this class we will mainly use hexadecimal colours.

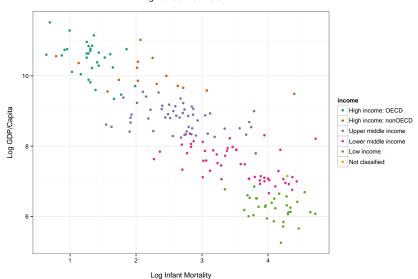
This is probably the most commonly used system for choosing colours on the web.

Every colour is given six digits.

A good website for getting hexadecimal colour schemes is: http://colorbrewer2.org/.

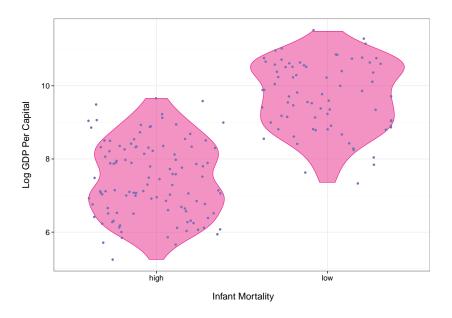
```
# Create colour vector
Colours <- c("#1B9E77", "#D95F02", "#7570B3",
             "#E7298A". "#66A61E". "#E6AB02")
# Create graph
ggplot(data = InfantNoMiss,
                    aes(log(InfantMortality),
                        log(GDPperCapita))) +
        geom_point(aes(colour = income)) +
        scale_color_manual(values = Colours) +
        xlab("\nLog Infant Mortality") +
        ylab("Log GDP/Capita\n") +
        ggtitle("Log Transformed Data\n") +
        theme bw()
```

Log Transformed Data



ggplot2 is very flexible

```
# Create a violin Plot
ggplot(InfantNoMiss, aes(factor(DumMort),
                        log(GDPperCapita))) +
          geom violin(fill = "#E7298A",
                      colour = "#E7298A",
                      alpha = I(0.5)) +
          geom_jitter(color = "#7570B3") +
          xlab("\n Infant Mortality") +
          ylab("Log GDP Per Capital\n") +
          theme bw(base size = 16)
```



Showing results from regression models

King, Tomz, and Wittenberg (2001) argue that **post estimation** simulations can be used to effectively communicate **results from** regression models.

Steps

- 1. Estimate our parameters' point estimates for $\hat{\beta}_{1...k}$.
- 2. Draw n values of the point estimates from multivariate normal distributions with means $\bar{\beta}_{1...k}$ and variances specified by the parameters' estimated co-variance.
- 3. Use the simulated values to calculate quantities of interest (e.g. predicted probabilities).
- 4. Plot the simulated distribution using visual weighting.

Notes

Post-estimation simulations allow us to effectively communicate our estimates and the **uncertainty around them**.

This method is broadly similar to a fully Bayesian approach with Markov-Chain Monte Carlo or bootstrapping. Just differ on **how the parameters are drawn**.

Implementation

It is relatively easy to find the coefficient estimates from an estimated model with coef.

You can find the co-variance matrix with vcov.

Then draw the values from the multivariate normal distribution with myrnorm.

Finally, calculate the quantity of interest with the draws + fitted values using and plot the results.

Easier Implementation

The Zelig package streamlines this process.

Zelig (1)

First estimate your regression model using zelig.

Zelig (2)

Then set the fitted values with setx.

```
setZ1 <- setx(Z1, gre = 220:800)
```

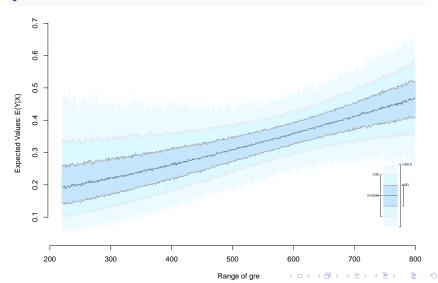
And run the simulations (1,000 by default) with sim.

```
simZ1 \leftarrow sim(Z1, x = setZ1)
```

Zelig (3)

Plot:

plot(simZ1)



Seminar

Create **tables** and **visualisations** of **descriptive** and **inferential** for your Assignment 3 using the techniques covered in class today.

If you don't have your data set fully cleaned yet, use one of the built-in R data sets.

References

King, Gary, Micheal Tomz, and Jason Wittenberg. 2001. "Making the Most of Statistical Analyses: Improving Interpretation and Presentation." *American Journal of Political Science* 22 (4): 341–255.

Tufte, Edward R. 2001. *The Visual Display of Quantitative Information*. 2nd ed. Cheshire, CT: Graphics Press.