# Amounts and Proportions

#### **Session 4**

PMAP 8921: Data Visualization with R Andrew Young School of Policy Studies May 2020

#### **Plan for today**

Reproducibility

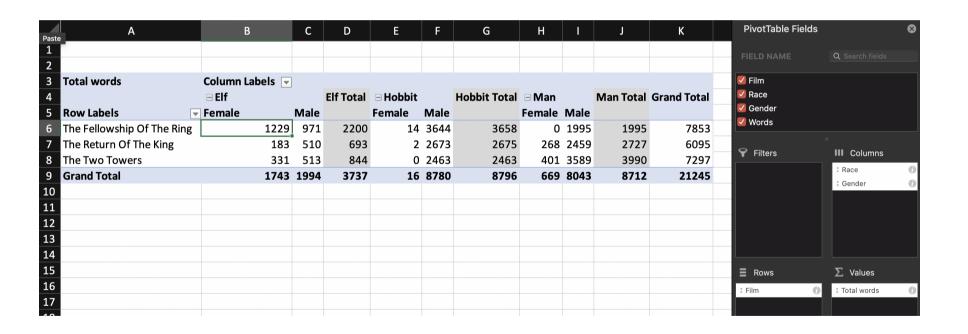
**Amounts** 

Proportions

# Reproducibility

## Why am I making you learn R?

#### Pivot Tables do the same thing!



## Why am I making you learn R?

More powerful

Free and open source

Reproducibility

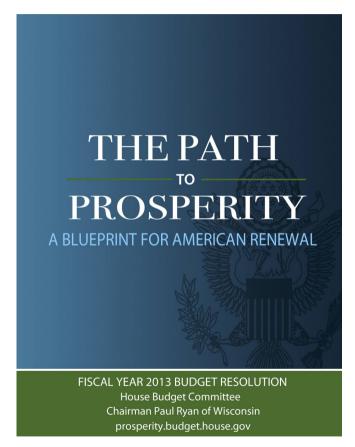
#### **Austerity and Excel**

Growth in a Time of Debt Carmen M. Reinhart and Kenneth S. Rogoff NBER Working Paper No. 15639 January 2010, Revised January 2010 JEL No. E2,E3,E6,F3,F4,N10

#### **ABSTRACT**

We study economic growth and inflation at different levels of government and external debt. Our analysis is based on new data on forty-four countries spanning about two hundred years. The dataset incorporates over 3,700 annual observations covering a wide range of political systems, institutions, exchange rate arrangements, and historic circumstances. Our main findings are: First, the relationship between government debt and real GDP growth is weak for debt/GDP ratios below a threshold of 90 percent of GDP. Above 90 percent, median growth rates fall by one percent, and average growth falls considerably more. We find that the threshold for public debt is similar in advanced and emerging economies. Second, emerging markets face lower thresholds for external debt (public and private)—which is usually denominated in a foreign currency. When external debt reaches 60 percent of GDP, annual growth declines by about two percent; for higher levels, growth rates are roughly cut in half. Third, there is no apparent contemporaneous link between inflation and public debt levels for the advanced countries as a group (some countries, such as the United States, have experienced higher inflation when debt/GDP is high). The story is entirely different for emerging markets, where inflation rises sharply as debt increases.

Debt:GDP ratio 90%+ → -0.1% growth



Paul Ryan's 2013 House budget resolution

#### **Austerity and Excel**



**Thomas Herndon** 

Over time, another problem emerged: Other researchers, using seemingly comparable data on debt and growth, couldn't replicate the Reinhart-Rogoff results. They typically found some correlation between high debt and slow growth — but nothing that looked like a tipping point at 90 percent or, indeed, any particular level of debt.

Finally, Ms. Reinhart and Mr. Rogoff allowed researchers at the University of Massachusetts to look at their original spreadsheet — and the mystery of the irreproducible results was solved. First, they omitted some data; second, they used unusual and highly questionable statistical procedures; and finally, yes, they made an Excel coding error. Correct these oddities and errors, and you get what other researchers have found: some correlation between high debt and slow growth, with no indication of which is causing which, but no sign at all of that 90 percent "threshold."

From Paul Krugman, "The Excel Depression"

## **Austerity and Excel**

Table 1. Real GDP Growth as the Level of Government Debt Varies: Selected Advanced Economies, 1790-2009

(	annual	percent	chan	ge`	)
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		C	entral (Federal) go	vernment debt/ C	GDP
Country	Period	Below 30	30 to 60	60 to 90	90 percent and
		percent	percent	percent	above
Australia	1902-2009	3.1	4.1	2.3	4.6
Austria	1880-2009	4.3	3.0	2.3	n.a.
Belgium	1835-2009	3.0	2.6	2.1	3.3
Canada	1925-2009	2.0	4.5	3.0	2.2
Denmark	1880-2009	3.1	1.7	2.4	n.a.
Finland	1913-2009	3.2	3.0	4.3	1.9
France	1880-2009	4.9	2.7	2.8	2.3
Germany	1880-2009	3.6	0.9	n.a.	n.a.
Greece	1884-2009	4.0	0.3	4.8	2.5
Ireland	1949-2009	4.4	4.5	4.0	2.4
Italy	1880-2009	5.4	4.9	1.9	0.7
Japan	1885-2009	4.9	3.7	3.9	0.7
Netherlands	1880-2009	4.0	2.8	2.4	2.0
New Zealand	1932-2009	2.5	2.9	3.9	3.6
Norway	1880-2009	2.9	4.4	n.a.	n.a.
Portugal	1851-2009	4.8	2.5	1.4	n.a.
Spain	1850-2009	1.6	3.3	1.3	2.2
Sweden	1880-2009	2.9	2.9	2.7	n.a.
United Kingdom	1830-2009	2.5	2.2	2.1	1.8
United States	1790-2009	4.0	3.4	3.3	-1.8
Average		3.7	3.0	3.4	1.7
Median		3.9	3.1	2.8	1.9
Number of observa	tions = 2,317	866	654	445	352

Debt:GDP ratio = 90%+  $\rightarrow$  2.2% growth (!!)

#### **Genes and Excel**

**Septin 2** 

Membrane-Associated Ring Finger (C3HC4) 1

2310009E13

	Α	В
1	Actual value	What Excel turns it into
2	SEPT2	2-Sep
3	MARCH1	1-Mar
4	2310009E13	2.31E+19
_		

20% of genetics papers between 2005–2015 (!!!)

## General guidelines

Don't touch the raw data

If you do, explain what you did!

Use self-documenting, reproducible code

R Markdown!

**Use open formats** 

Use .csv, not .xlsx

#### R Markdown in real life

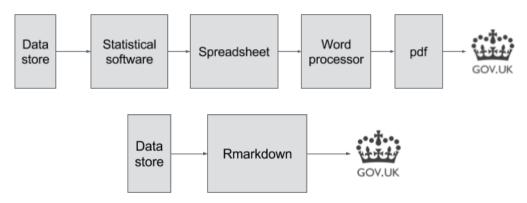
#### 3.1.2 Data Visualization

We use ggplot2 as our main package to create ad-hoc exploratory graphics as well as polished-looking customized visualizations. When combined with tools to clean and transform data, ggplot2 allows analysts to quickly translate insights into high quality, compelling visualizations. In addition to the static graphics of ggplot2, we often make interactive visualizations or dashboards using R packages such as plotly (Sievert et al. 2017), leaflet (Cheng et al. 2017), dygraphs (Vanderkam et al. 2017), DiagrammeR (Sveidqvist et al. 2017), and shiny (Chang et al. 2017).

#### 3.1.3 Reproducible Research

At Airbnb, all R analyses are documented in rmarkdown, where code and visualizations are combined within a single written report. Posts are carefully reviewed by experts in the content area and techniques used, both in terms of methodologies and code style, before publishing and sharing with the business partners. The peer review process is

Airbnb, ggplot, and rmarkdown

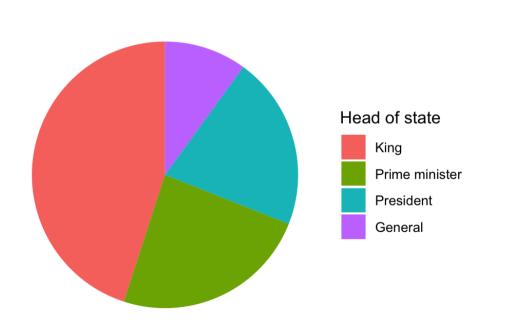


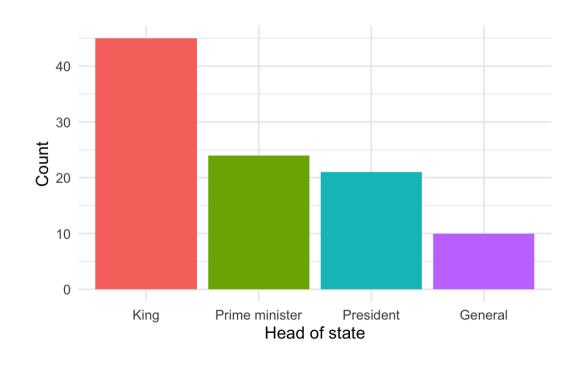
The UK's reproducible analysis pipeline

## Amounts

## Yay bar plots!

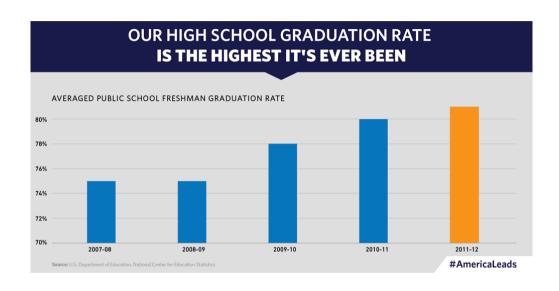
# We are a lot better at visualizing line lengths than angles and areas





#### Oh no bar plots!





#### Start at zero

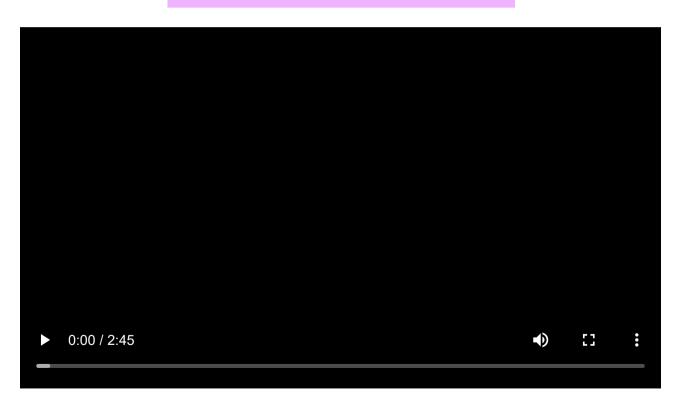
The entire line length matters, so don't truncate it!

Always start at 0

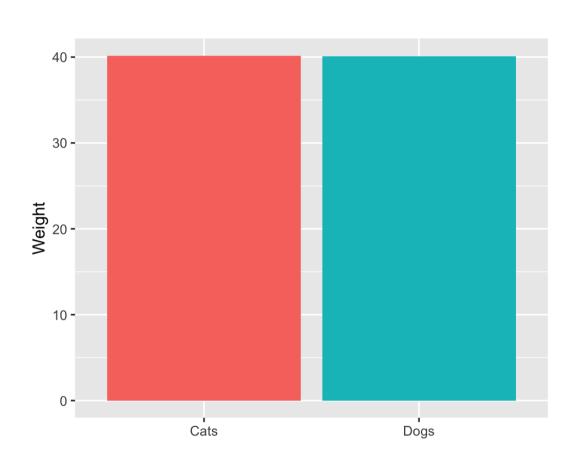
(Or don't use bars)

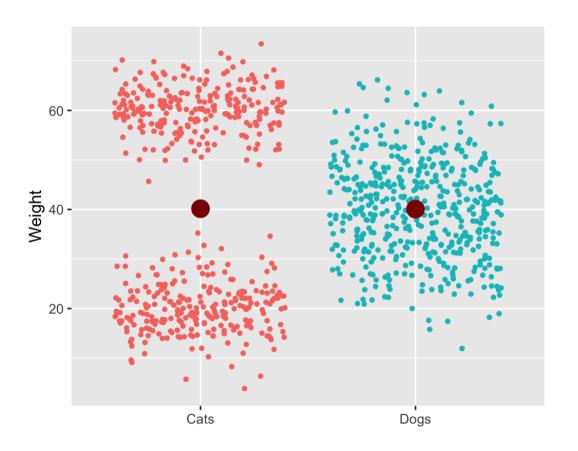
#### Bar plots and summary statistics

#### #barbarplots

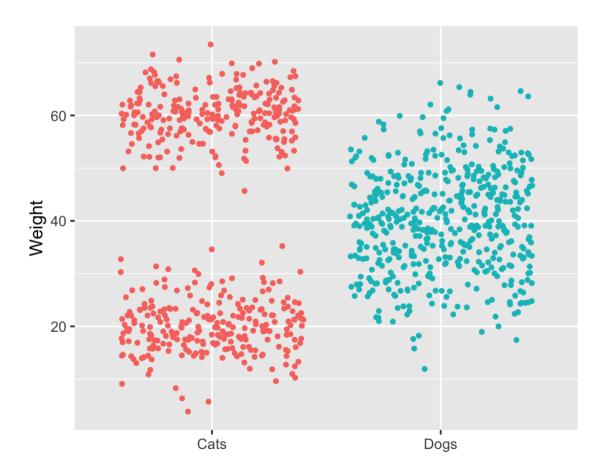


#### Bar plots and summary statistics

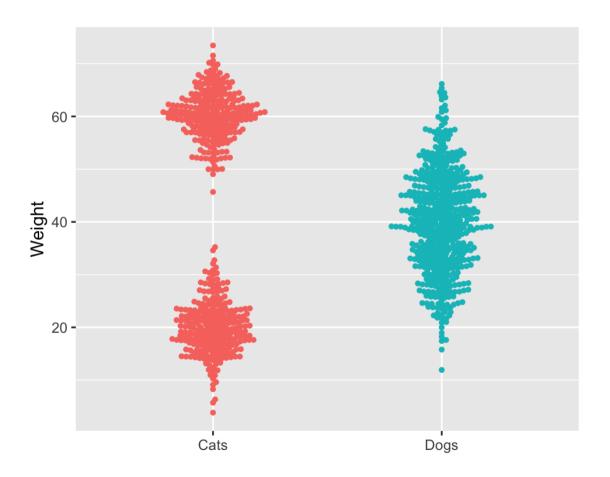




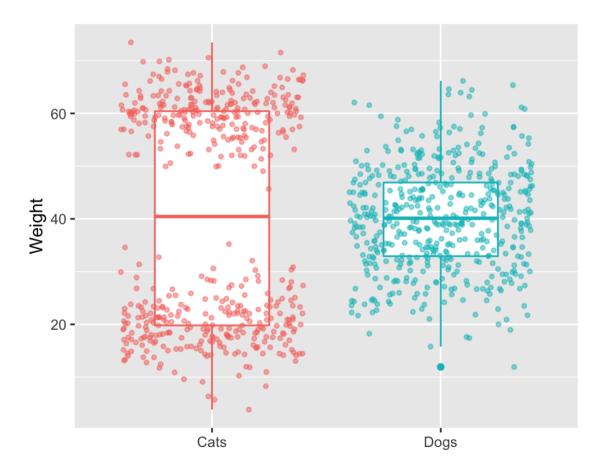
#### Show more data with strip plots



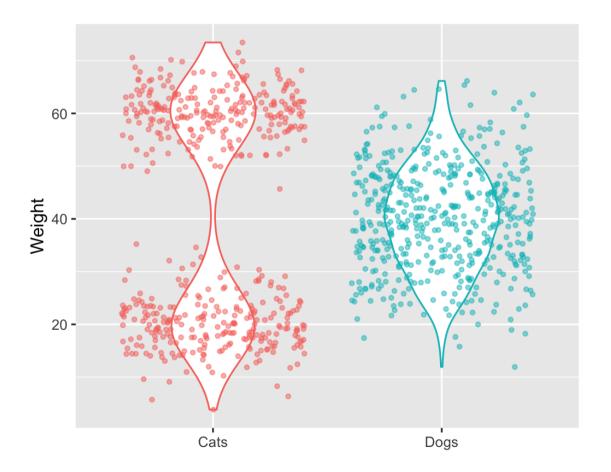
#### Show more data with beeswarm plots



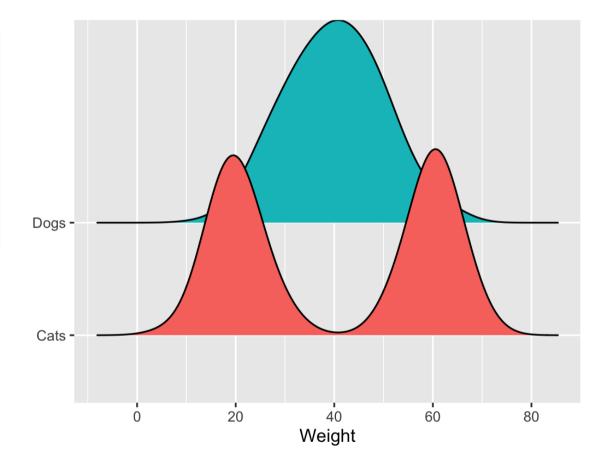
## Combine boxplots with points



#### Combine violins with points



## Overlapping ridgeplots



#### **General rules**

Bar charts always start at zero

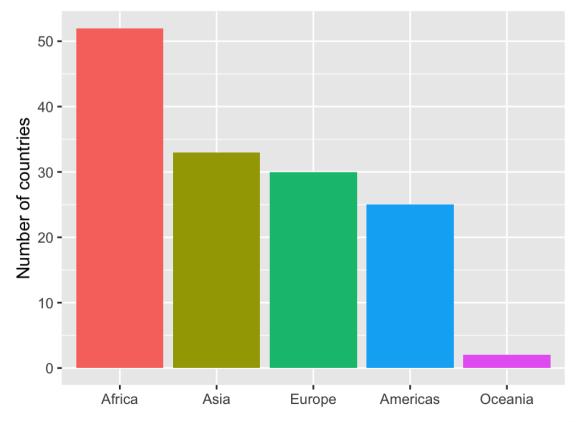
Don't use bars for summary statistics. You throw away too much information.

The end of the bar is often all that matters

#### Lots of alternatives

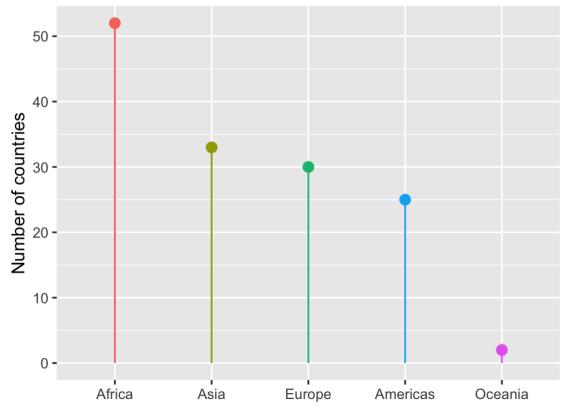
#### We'll use a summarized version of the gapminder dataset as an example

```
library(gapminder)
gapminder_continents <- gapminder %>%
 filter(year == 2007) %>% # Only look at 20
 count(continent) %>% # Get a count of con
  arrange(desc(n)) %>% # Sort descendingly
  # Make continent into an ordered factor
 mutate(continent = fct_inorder(continent))
ggplot(gapminder_continents,
       aes(x = continent, y = n, fill = cont
 geom_col() +
 guides(fill = FALSE) +
  labs(x = NULL, y = "Number of countries")
```



## **Alternatives: Lollipop charts**

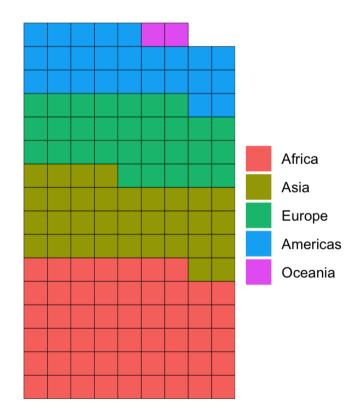
#### Since the end of the bar is important, emphasize it the most



#### Alternatives: Waffle charts

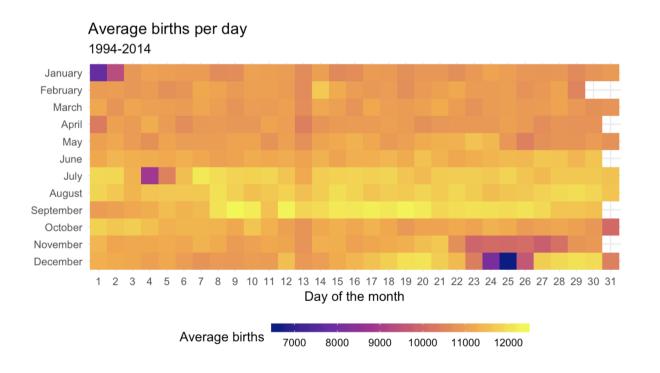
#### Show the individual observations as squares

```
# This has to be installed in a special way-
# Run this in your console:
# devtools::install_github("hrbrmstr/waffle")
library(waffle)
ggplot(gapminder_continents,
       aes(x = continent, y = n,
           fill = continent)) +
  geom_waffle(aes(values = n), # geom_waffle
              n_rows = 9, # It has lots of a
              flip = TRUE) +
  labs(fill = NULL) +
  coord_equal() + # Make all the squares squ
 theme_void() # Use a completely empty them
```



#### **Alternatives: Heatmaps**

If exact counts are less important, try a heatmap with geom\_tile()



# Proportions

#### Why proportions?

Sometimes we want to compare values across a whole population instead of looking at raw counts

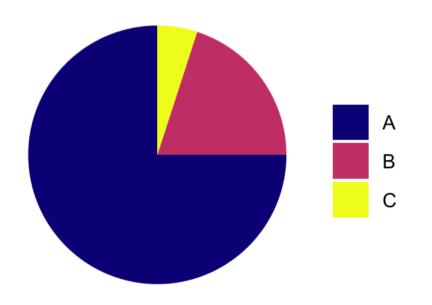
Only do this when it makes analytical sense!

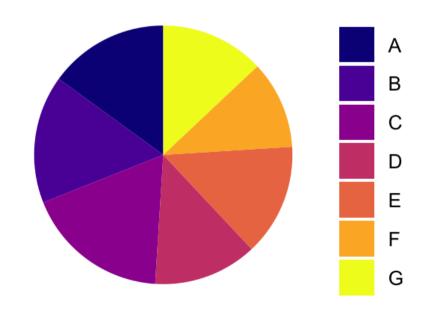
**COVID-19 amounts vs. proportions** 

#### Pie charts

Perceptual issues with angle and fill space

Only okay(ish) if there are a few easily distinguishable categories





#### Alternatives

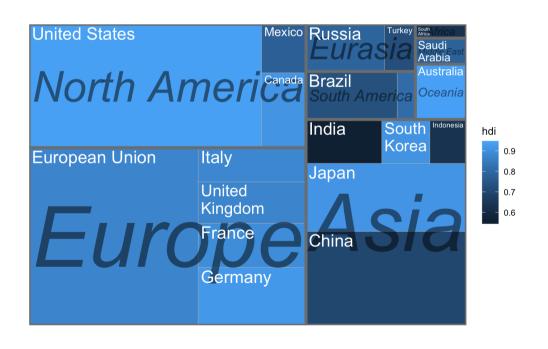
**Bar plots** 

Any of the alternatives to bar plots

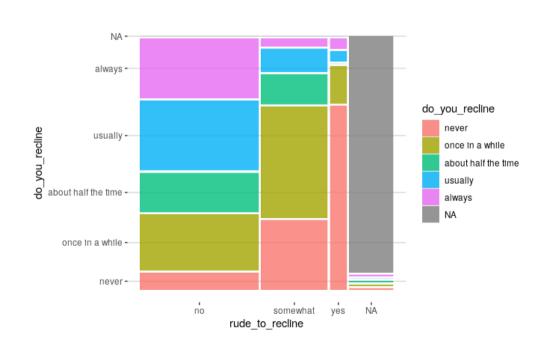
Treemaps and mosaic plots (but these can still be really hard to interpret)

#### Treemaps and mosaic plots

## Treemaps with the treemapify package



# Mosaic plots with the **ggmosaic** package



#### Alternatives

**Bar plots** 

Any of the alternatives to bar plots

Treemaps and mosaic plots (but these can still be really hard to interpret)

**Specialized figures like parliament plots** 

## Parliament plots

#### Parliament plots with the **ggparliament** package

