# Refining your plots

Daniel Anderson Week 6, Class 1



# Agenda

- Axes and aspect ratios
- Annotations
- Themes

### What we won't get to

Each of the following are pretty fundamental to good data viz, but we won't have time to go over them today. Please make sure to read the corresponding chapters:

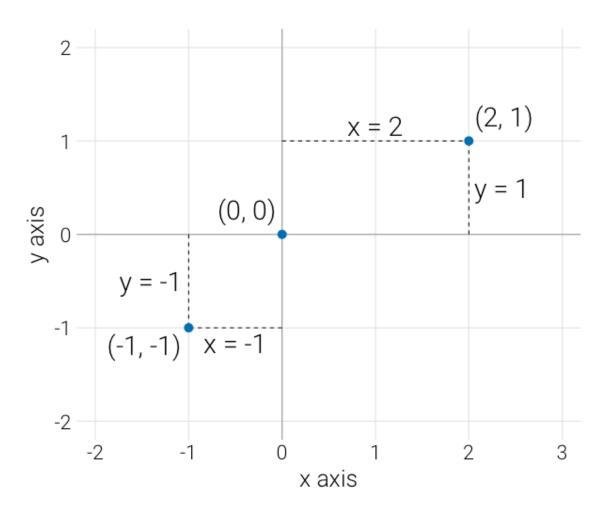
- Handling high data density (lots of overlapping points)
- Compound figures
  - See {patchwork} and {cowplot}
- Exporting figures

# **Learning Objectives**

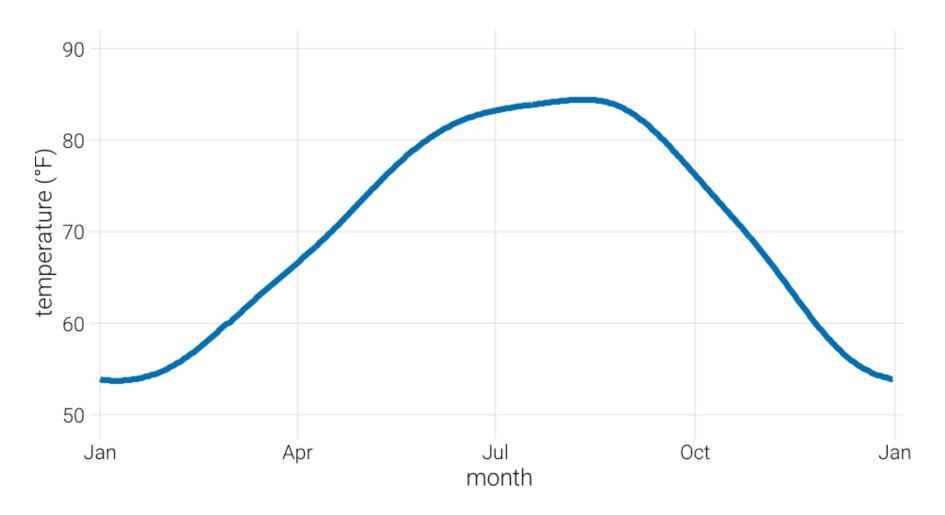
- Understand how to make a wide variety of tweaks to ggplot to essentially make it look however you want it to.
- Understand common modifications to plots to make them more clear and reduce cognitive load

## Axes

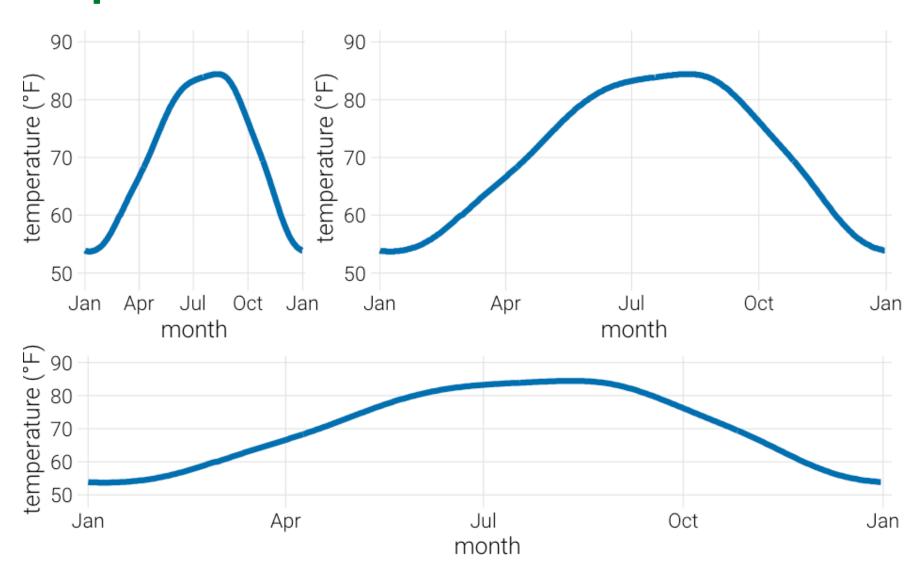
• Cartesian coordinates - what we generally use

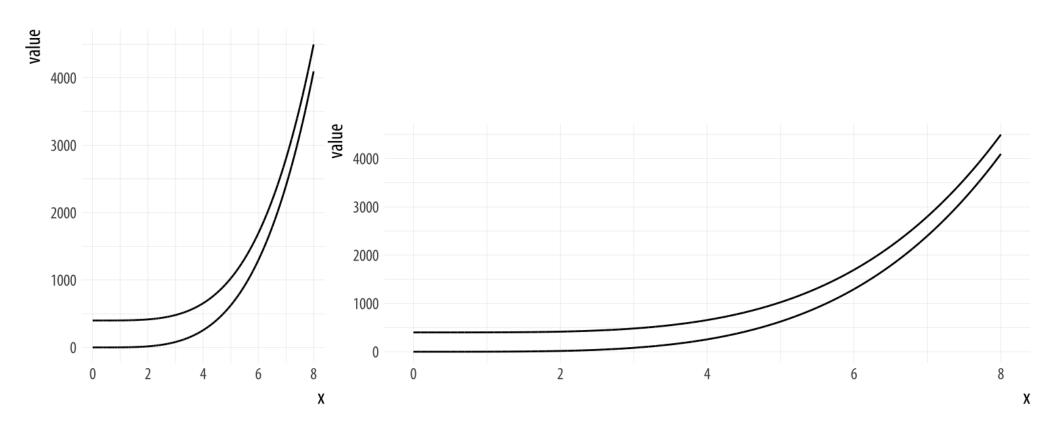


### Different units



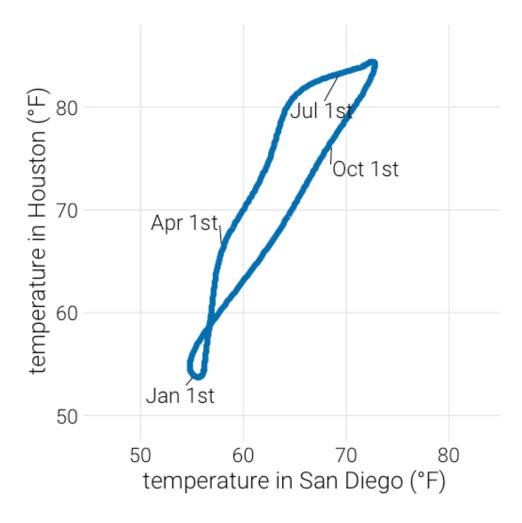
# **Aspect ratio**





### Same scales

Use coord\_fixed()



# Changing aspect ratio

- Explore how your plot will look in its final size
- No hard/fast rules (if on different scales)
- Not even really rules of thumb
- Keep visual perception in mind
- Try your best to be truthful show the trend/relation, but don't exaggerate/hide
   it

# Handy function

(from an apparently deleted tweet)

```
here's my favorite helper #rstats function. preview ggsave() output

ggpreview <- function (..., device = "png") {
  fname <- tempfile(fileext = pasteO(".", device))
  ggplot2::ggsave(filename = fname, device = device, ...)
  system2("open", fname)
  invisible(NULL)
}

— tj mahr  (atjmahr) January 9, 2019
```

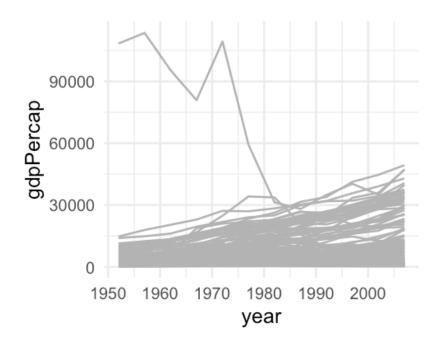
#### Gist

(side note: gists are a good way to share things)

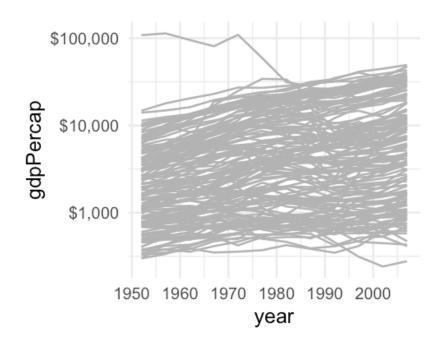
- See the full code/example here
- Let's take 7 minutes to play around:
  - Create a plot (could even be the example in the gist)
  - Try different aspect ratios by changing the width/length

### Scale transformations

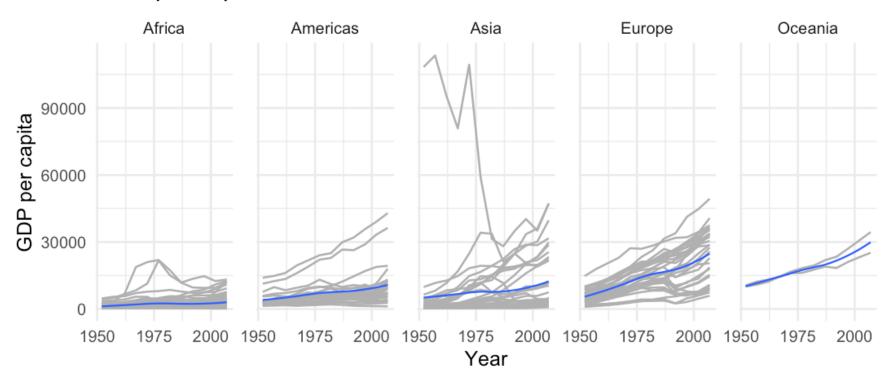
#### Raw scale



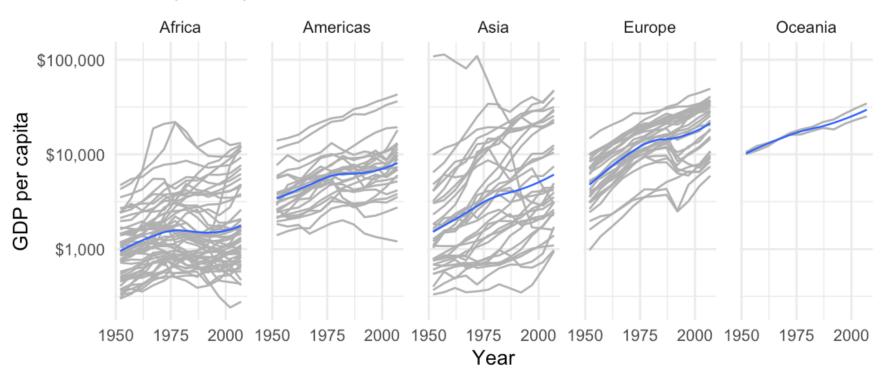
### Log10 scale



#### GDP per capita on Five Continents

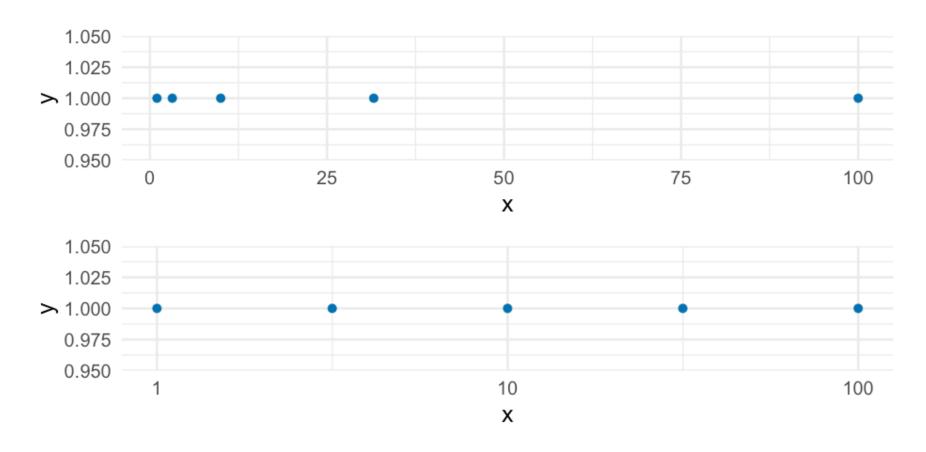


#### GDP per capita on Five Continents



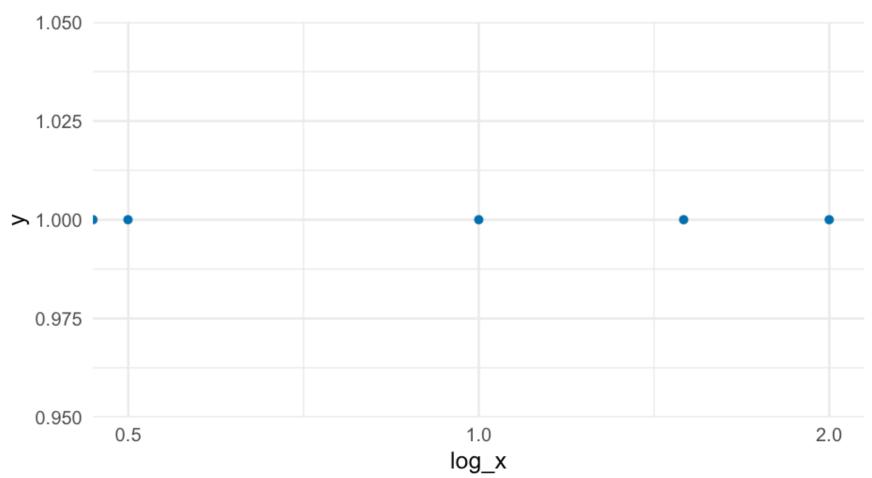
### **Scales**

# Scales



### Don't transform twice

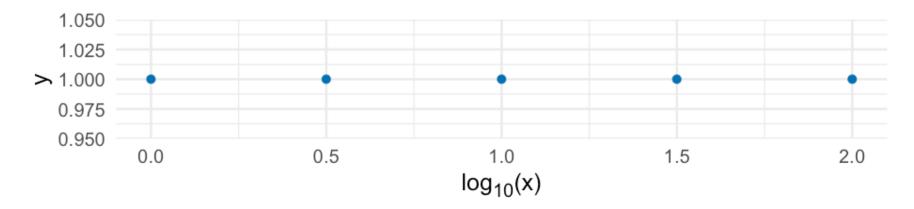
```
ggplot(d, aes(log_x, 1)) +
  geom_point(color = "#0072B2") +
  scale_x_log10()
```



# Careful with labeling

- Has the scale or the data been log transformed?
- Specify the base

```
library(ggtext)
ggplot(d, aes(log_x, 1)) +
  geom_point(color = "#0072B2") +
  labs(x = "log<sub>10</sub>(x)") +
  theme(axis.title.x = element_markdown())
```



Labels should denote the data, not the scale of the axis

```
ggplot(d, aes(x, 1)) +
geom_point(color = "#0072B2") +
scale_x_log10()

1.050
1.025
> 1.000
0.975
0.950
1 10 100
x
```

Labeling the above with  $log_{10}(x)$  would be ambiguous and confusing

# Labels and captions

### Disclaimer

- APA style requires the labels be made in specific ways
- Much of the following discussion still applies
- Our book (Wilke) uses a similar style throughout

### Title

## What is the point of your figure? What are you trying to communicate

- Figures should have only one title
- Use integrated title/subtitles for sharing with a broad audience
  - Blog posts
  - Social media
  - Reports to stakeholders
- Keep figures in subtext when there's a designated format you must adhere to
- Make sure your figure has a title
  - Should not start with "This figure displays/shows..."

# Caption

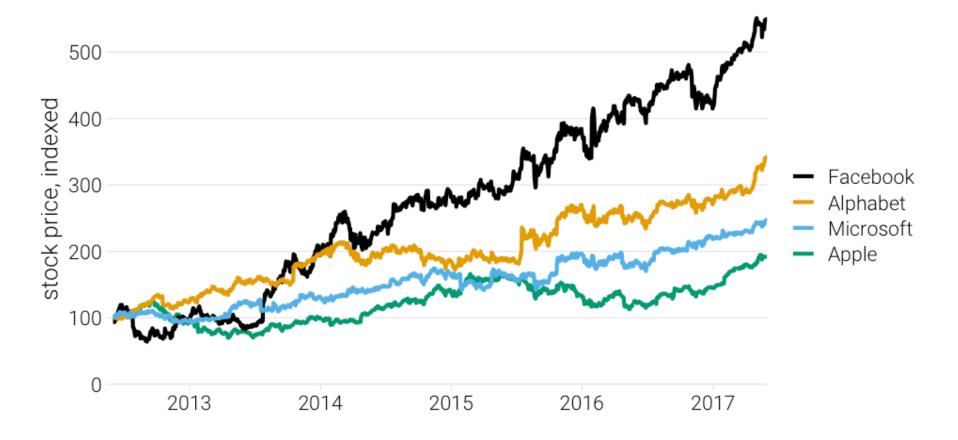
Consider stating the data source

#### Axis labels

- The title for the axis
- Critical for communication
- **Never** use variable names (very common and very poor practice)
- State the measure and the unit (if quantitative)
  - e.g., "Brain Mass (grams)", "Support for Measure (millions of people)", "Dollars spent"
  - Categorical variable likely will not need to the measurement unit

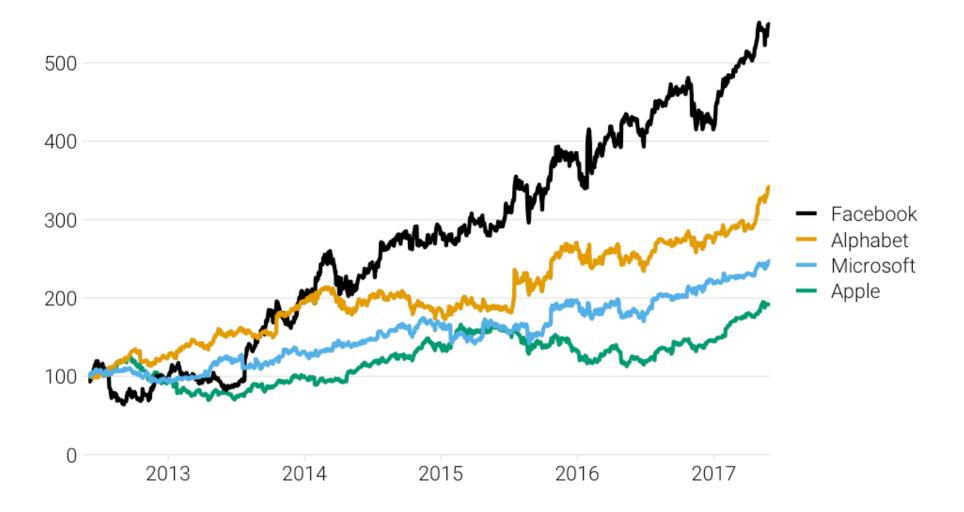
### **Omission**

- Consider omitting obvious or redundant labels
  - o Use labs(x = NULL) or labs(x = "")

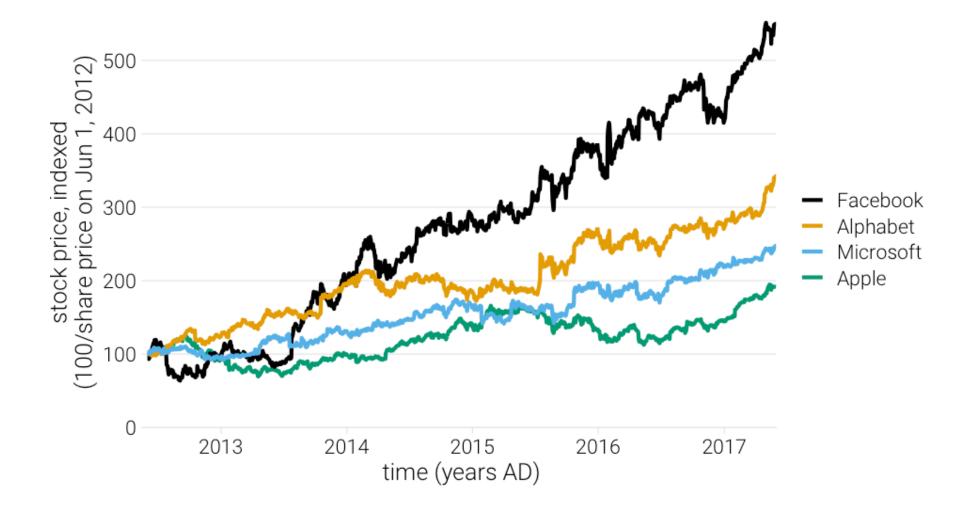


### **Omission**

• Do not omit title that are not obvious



### Don't overdo it



## Annotations

# Among the most effective

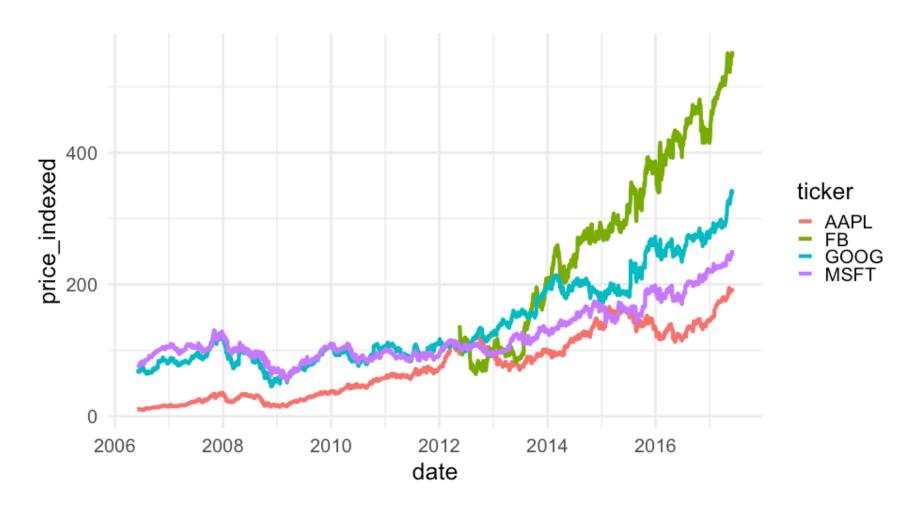
• If possible, try to remove legends, and just include annotations

# Building up a plot

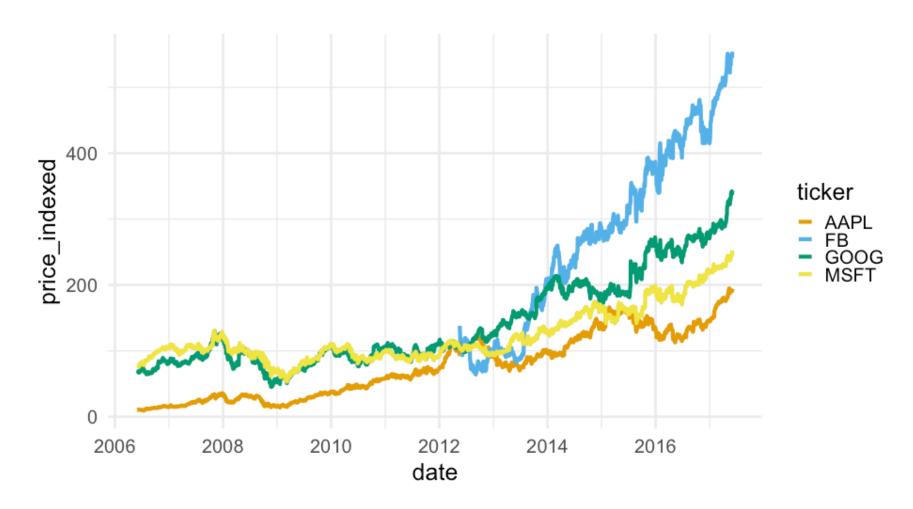
```
remotes::install_github("clauswilke/dviz.supp")
head(tech_stocks)
```

```
## # A tibble: 6 x 6
## # Groups: ticker [1]
    company ticker date
                               price index_price price_indexed
##
    <chr>
             <chr> <date>
                                            <dbl>
                                                         <dbl>
##
                                <dbl>
## 1 Alphabet GOOG 2017-06-02 975.6
                                           285.2
                                                      342.0757
## 2 Alphabet GOOG 2017-06-01 966.95
                                           285.2
                                                      339.0428
## 3 Alphabet GOOG 2017-05-31 964.86
                                           285.2
                                                      338.3100
## 4 Alphabet GOOG
                  2017-05-30 975.88
                                           285.2
                                                      342.1739
## 5 Alphabet GOOG
                  2017-05-26 971.47
                                           285.2
                                                      340.6276
## 6 Alphabet GOOG
                    2017-05-25 969.54
                                           285.2
                                                      339.9509
```

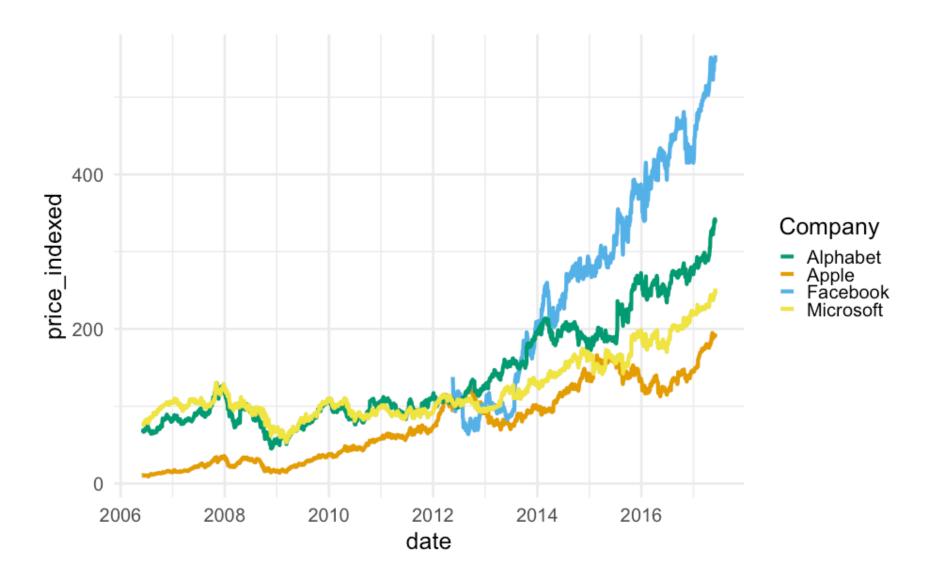
ggplot(tech\_stocks, aes(date, price\_indexed, color = ticker)) +
 geom\_line()



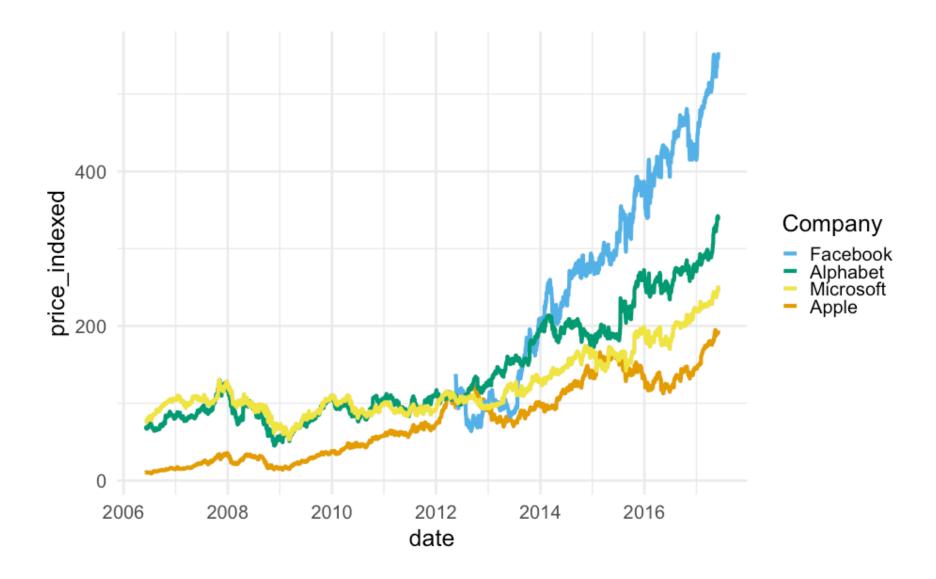
```
ggplot(tech_stocks, aes(date, price_indexed, color = ticker)) +
  geom_line() +
  scale_color_0kabeIto()
```

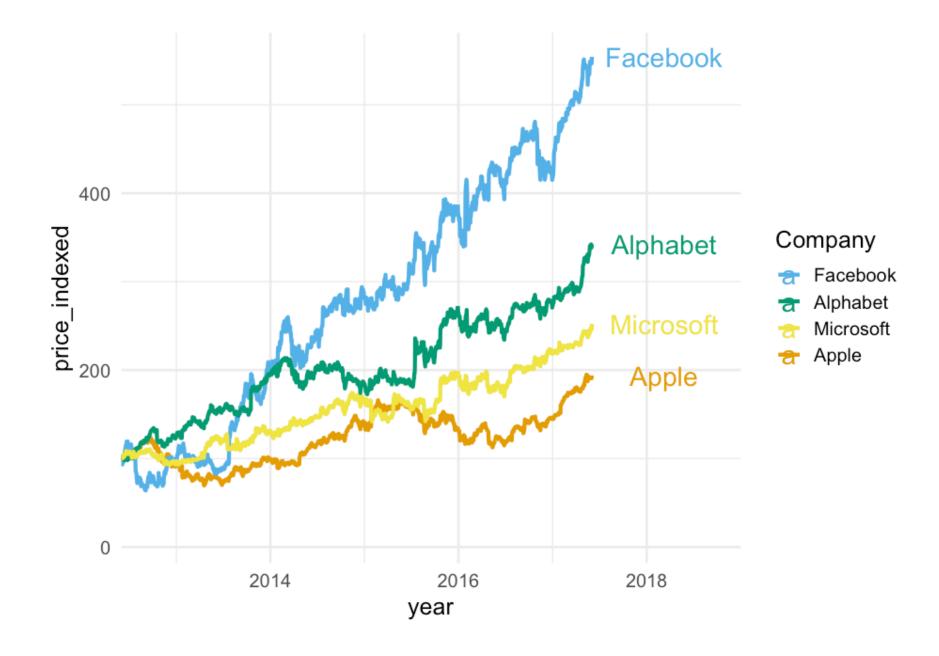


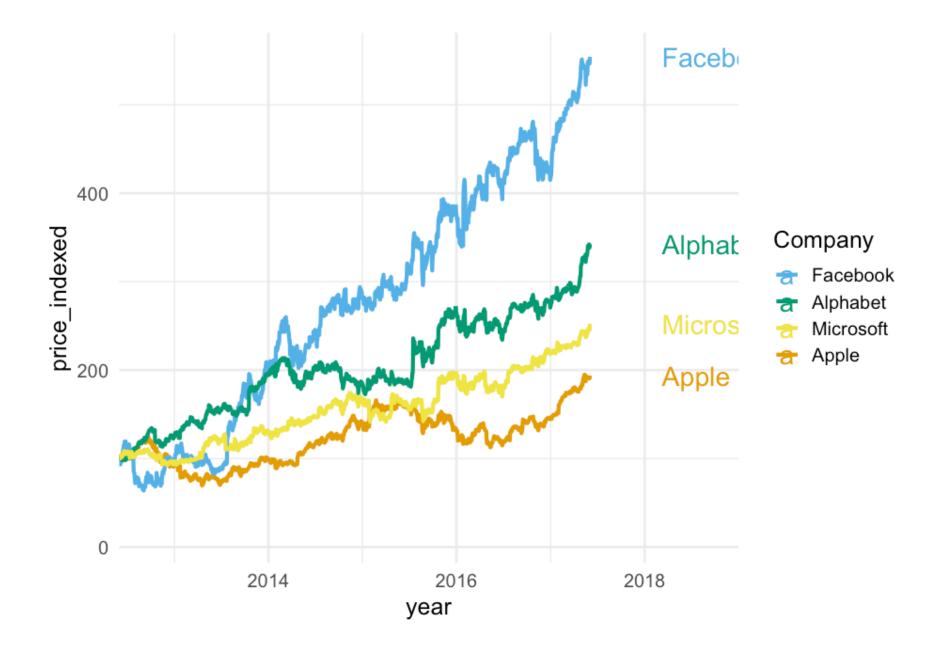
## Bad

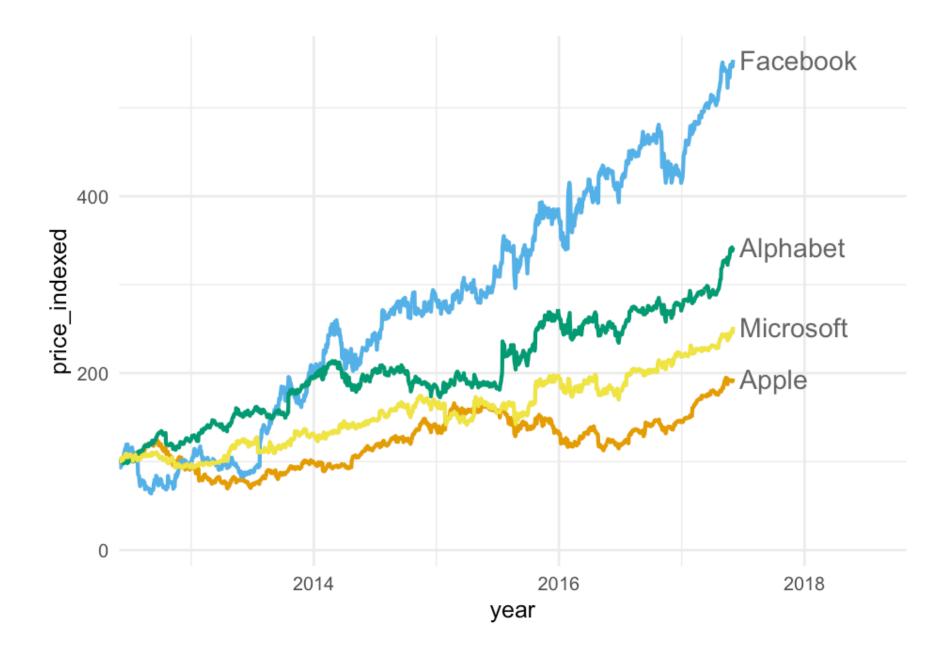


# Good

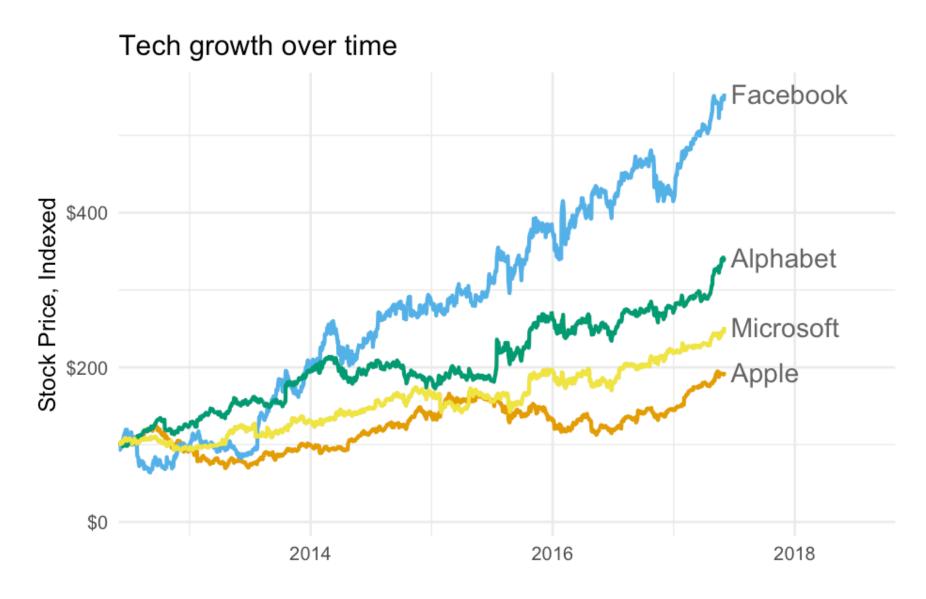








```
ggplot(tech_stocks, aes(date, price_indexed, color = ticker)) +
 geom_line() +
 scale_color_0kabeIto(name = "Company",
                       breaks = c("FB", "GOOG", "MSFT", "AAPL"),
                       labels = c("Facebook", "Alphabet", "Microsoft", "Appl
 scale_x_date(name = "",
               limits = c(ymd("2012-06-01"), ymd("2018-10-31")),
               expand = c(0,0) +
 scale v continuous(name = "Stock Price, Indexed",
                     labels = scales::dollar) +
 geom_text(data = filter(tech_stocks, date == "2017-06-02"),
            aes(y = price_indexed, label = company),
            color = "gray40",
           nudge_x = 20,
           hjust = 0,
            size = 10) +
 guides(color = "none") +
 labs(title = "Tech growth over time",
       caption = "Data from Wilke (2019): Fundamentals of Data Visualization
```



Data from Wilke (2019): Fundamentals of Data Visualization

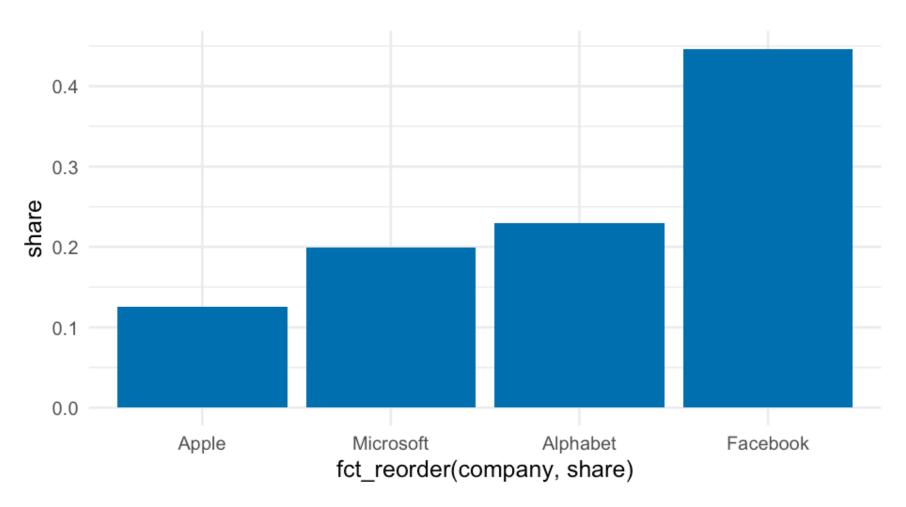
## Labeling bars

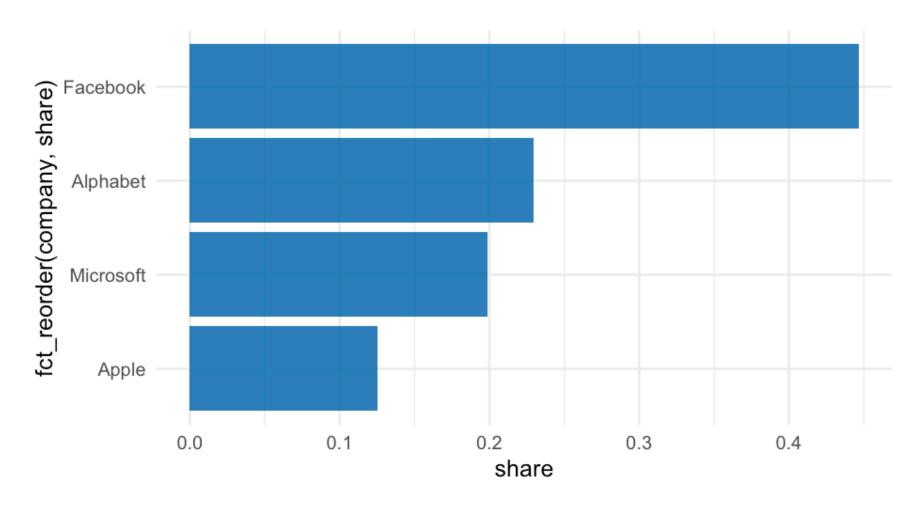
```
avs <- tech_stocks %>%
  group_by(company) %>%
  summarize(stock_av = mean(price_indexed)) %>%
  ungroup() %>%
  mutate(share = stock_av / sum(stock_av))
  avs

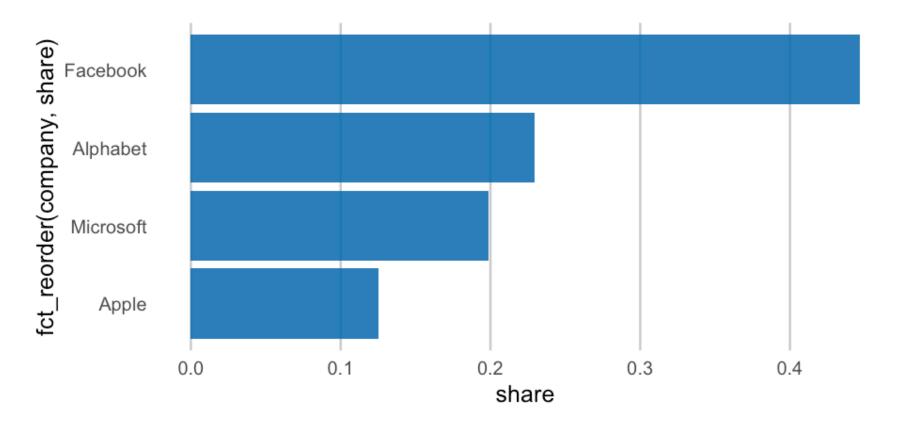
## # A tibble: 4 x 3
```

# Bar plot

```
ggplot(avs, aes(fct_reorder(company, share), share)) +
  geom_col(fill = "#0072B2")
```



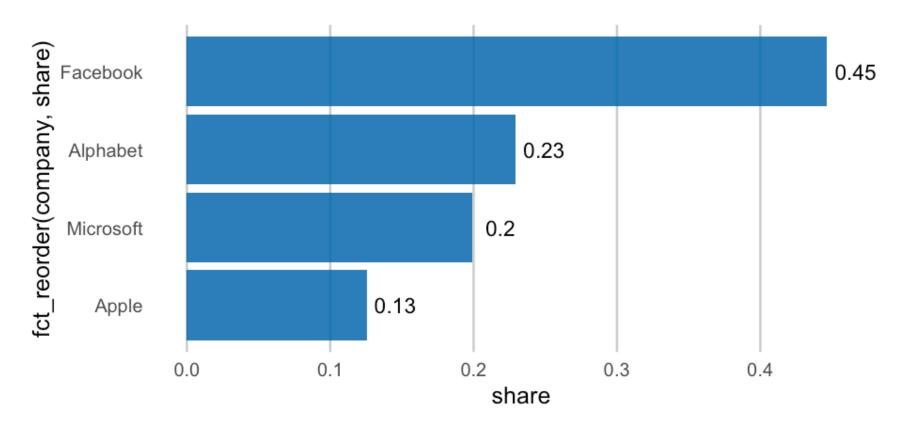




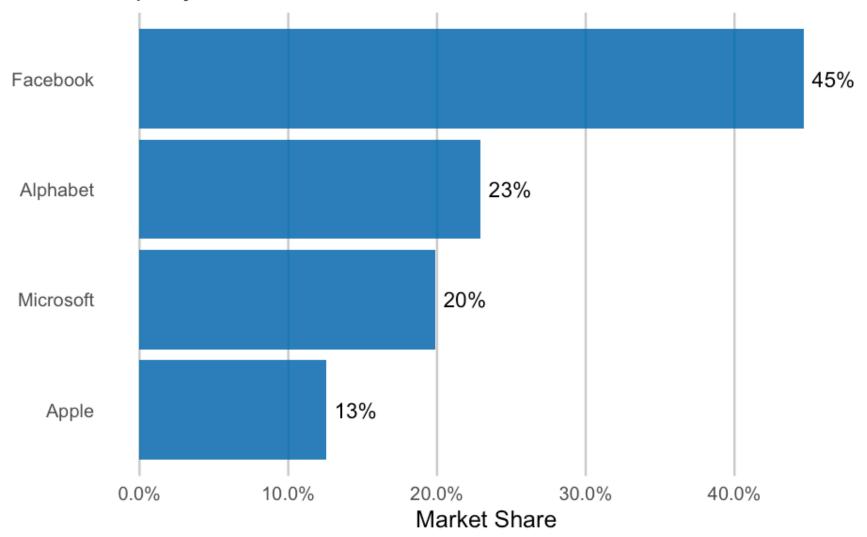
# Quick aside

Let's actually make a bar plot theme

```
bp_theme <- function(...) {
   theme_minimal(...) +
    theme(panel.grid.major.y = element_blank(),
        panel.grid.minor.x = element_blank(),
        panel.grid.major.x = element_line(color = "gray80"),
        plot.title.position = "plot")
}</pre>
```

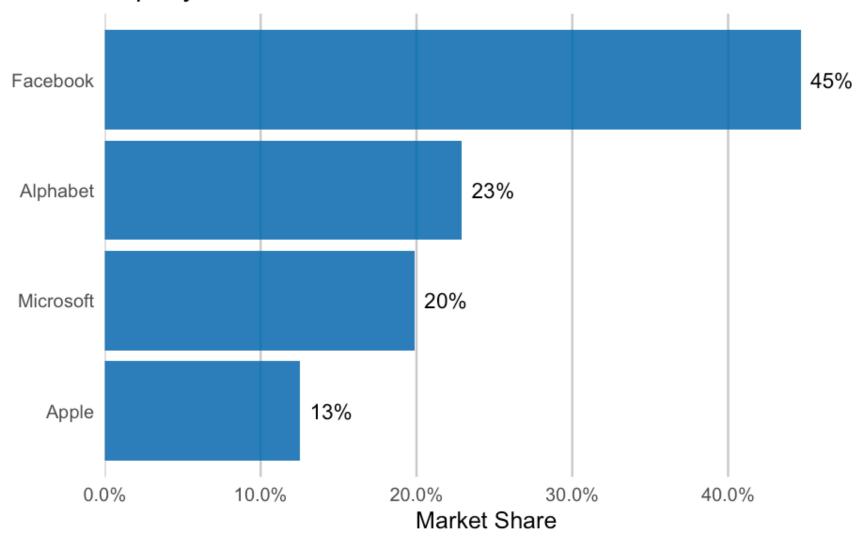


#### Tech company market control



Data from Clause Wilke Book: Fundamentals of Data Visualizations

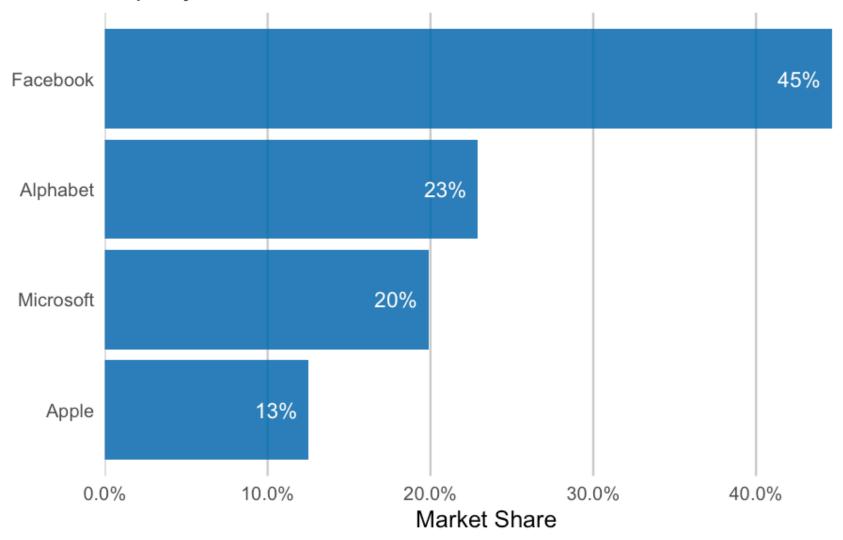
#### Tech company market control



Data from Clause Wilke Book: Fundamentals of Data Visualizations

### Last alternative

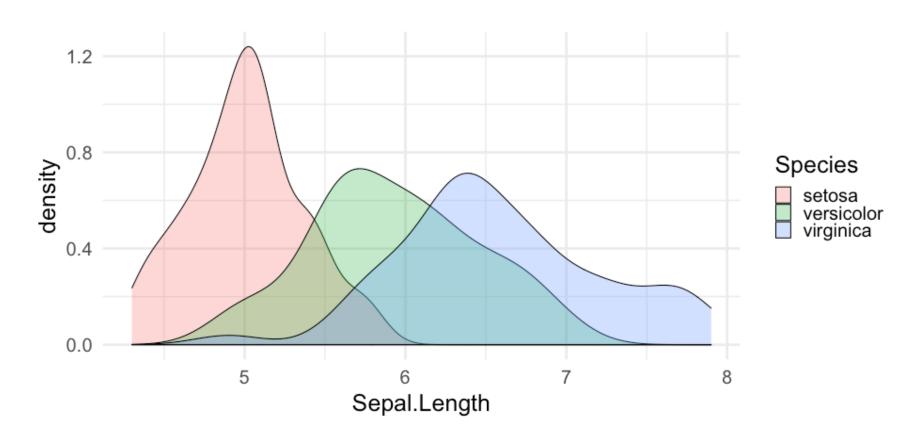
#### Tech company market control



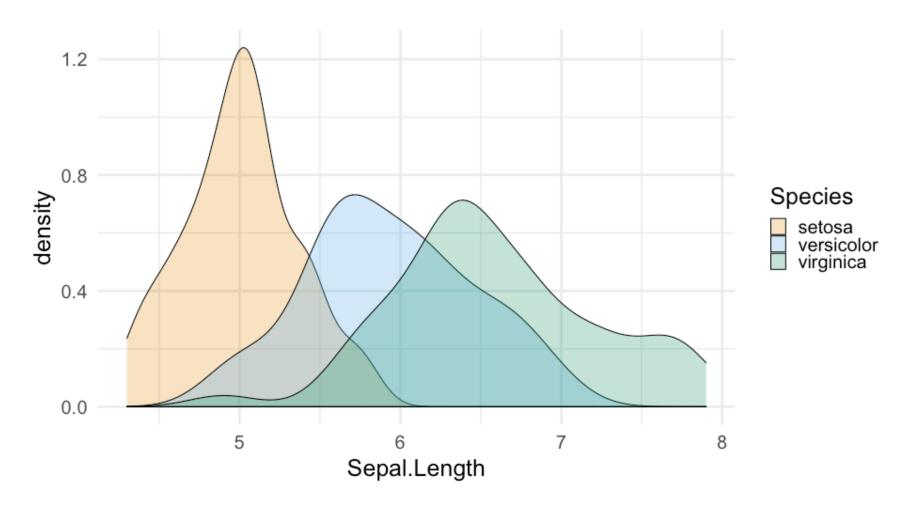
Data from Clause Wilke Book: Fundamentals of Data Visualizations

## Distributions

```
ggplot(iris, aes(Sepal.Length, fill = Species)) +
  geom_density(alpha = 0.3)
```

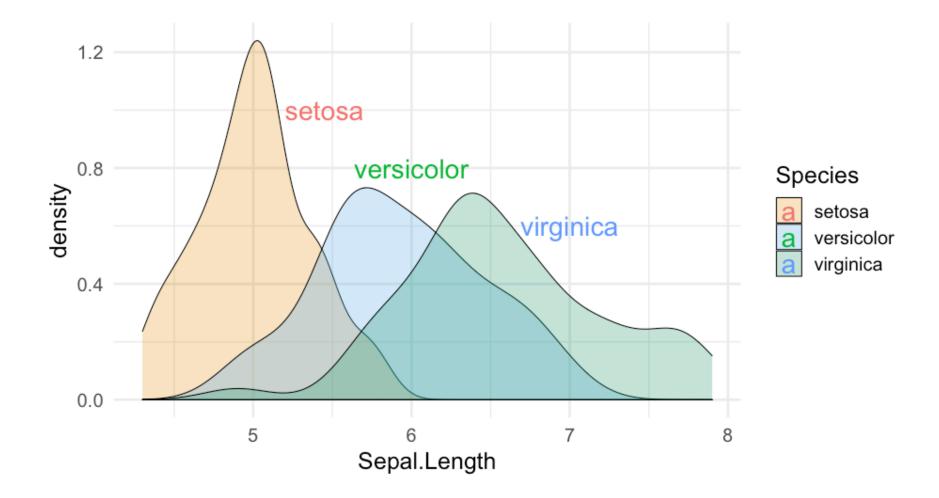


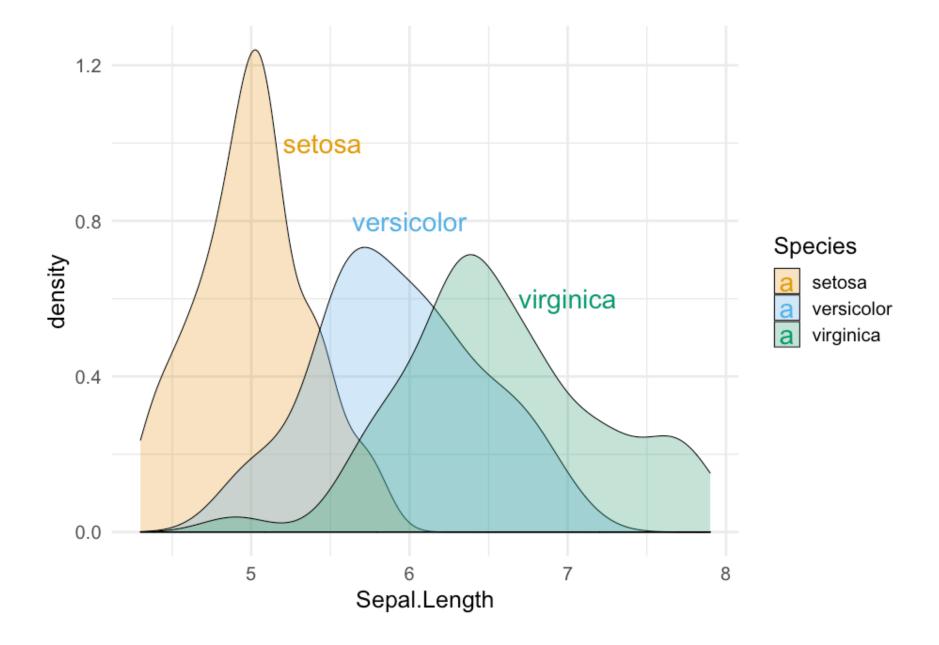
```
ggplot(iris, aes(Sepal.Length, fill = Species)) +
  geom_density(alpha = 0.3) +
  scale_fill_OkabeIto()
```

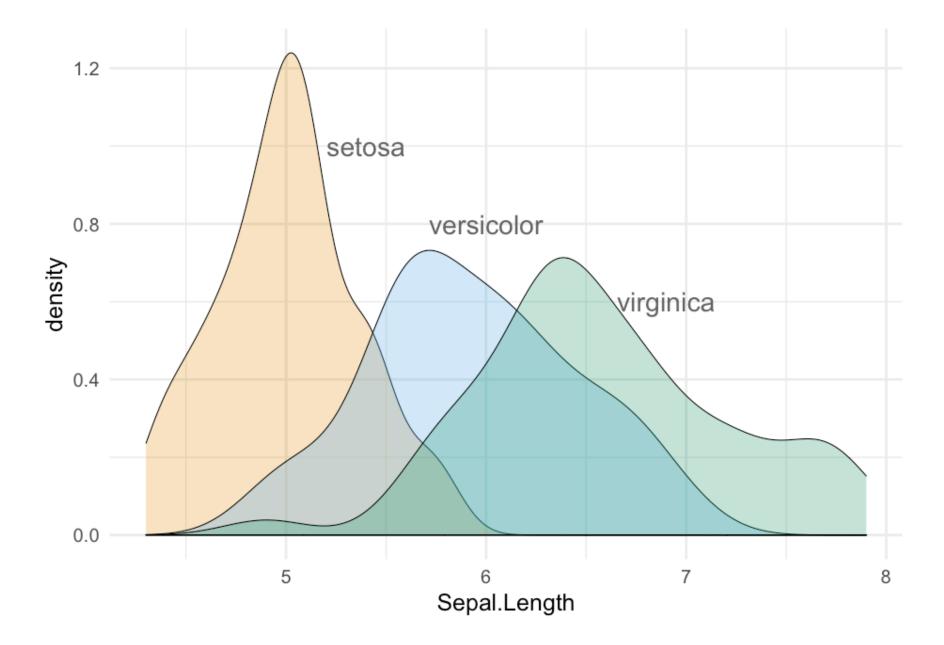


# Labeling

#### One method





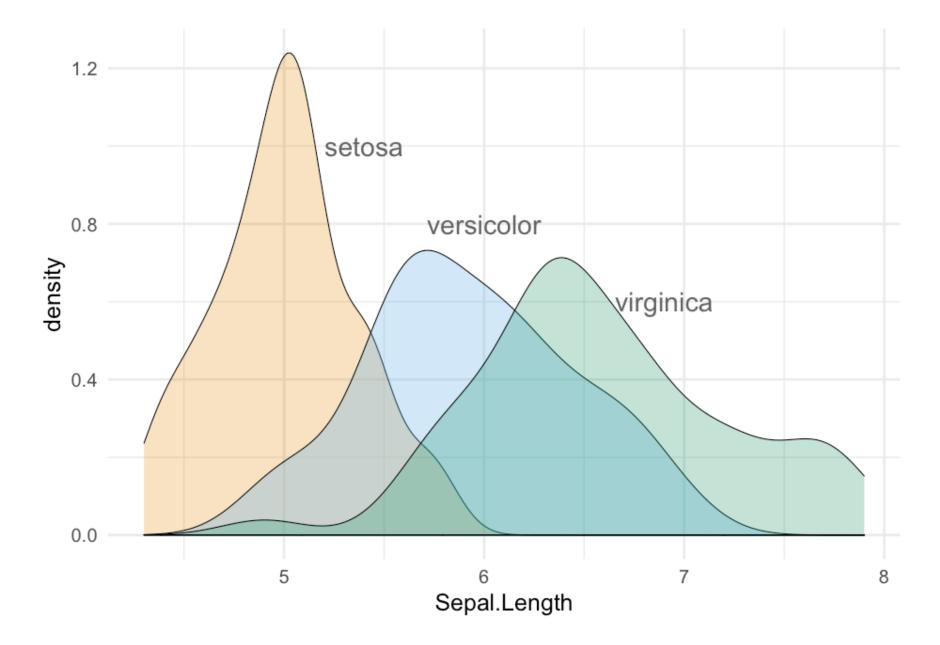


## Other options

- Rather than using a new data frame, you could use multiple calls to annotate.
- One is not necessarily better than the other, but I prefer the data frame method
- Keep in mind you can **always** use multiple data sources within a single plot
  - Each layer can have its own data source
  - Common in geographic data in particular

## Annotate example

```
ggplot(iris, aes(Sepal.Length, fill = Species)) +
  geom_density(alpha = 0.3) +
  scale_fill_OkabeIto() +
  scale_color_OkabeIto() +
  annotate("text", label = "setosa", x = 5.45, y = 1, color = "gray40") +
  annotate("text", label = "versicolor", x = 6, y = 0.8, color = "gray40") +
  annotate("text", label = "virginica", x = 7, y = 0.6, color = "gray40") +
  guides(fill = "none")
```

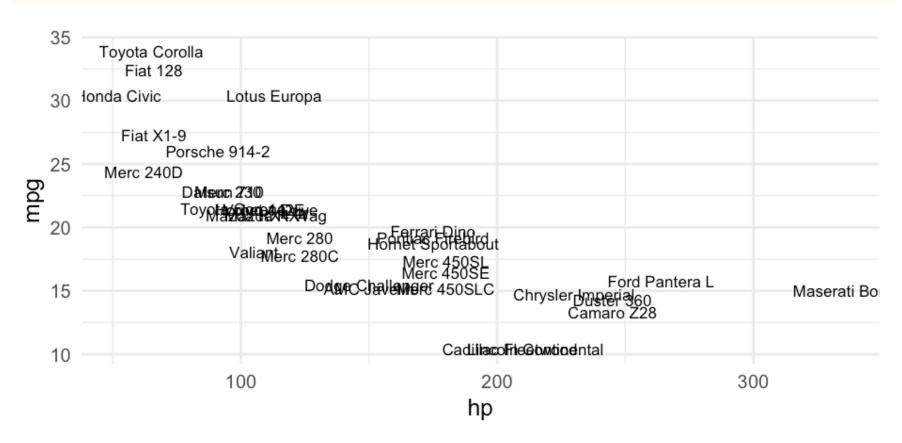


# ggrepel

# Plot text directly

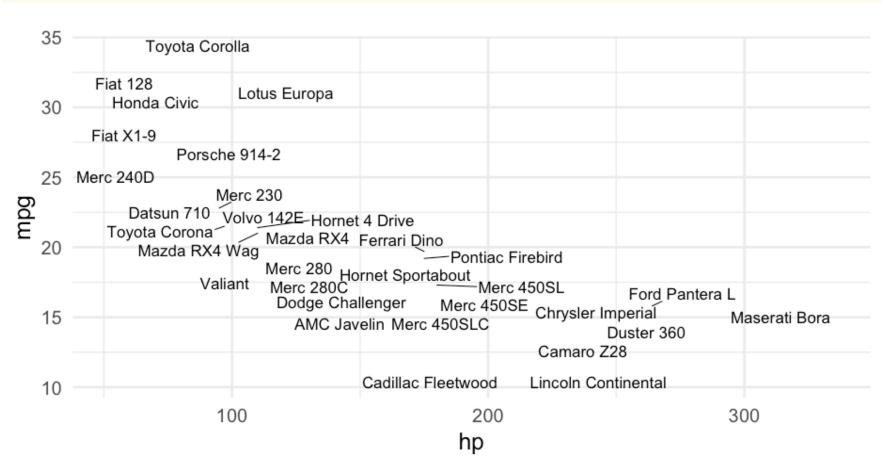
```
cars <- rownames_to_column(mtcars)

ggplot(cars, aes(hp, mpg)) +
  geom_text(aes(label = rowname))</pre>
```

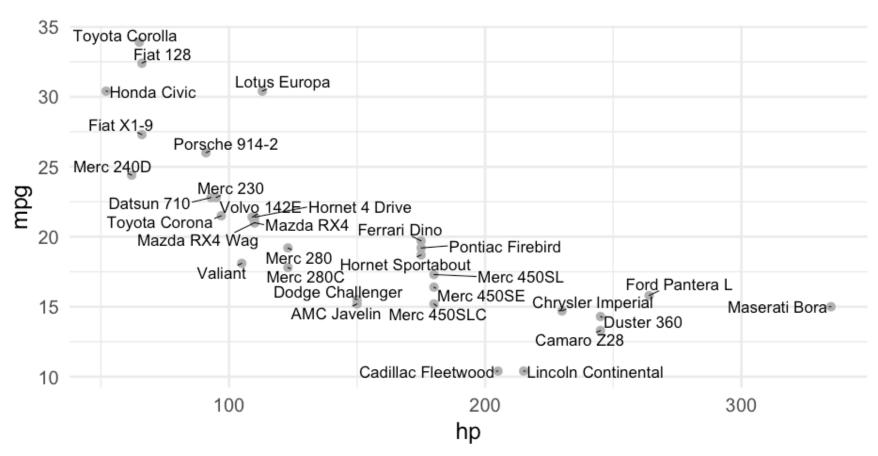


## Repel text

```
library(ggrepel)
ggplot(cars, aes(hp, mpg)) +
  geom_text_repel(aes(label = rowname))
```



# Slightly better



## Common use cases

- Label some sample data that makes some theoretical sense (we've seen this before)
- Label outliers
- Label points from a specific group (e.g., similar to highlighting can be used in conjunction)

### Some new data

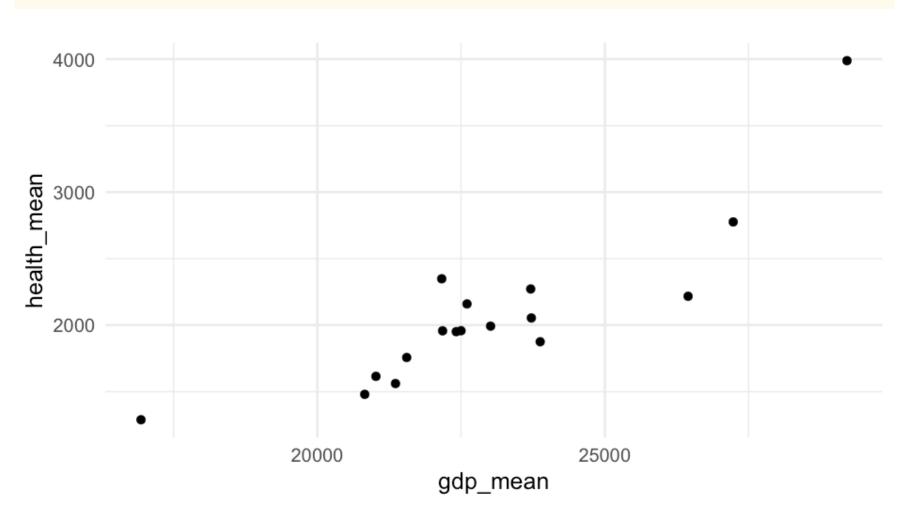
```
remotes::install_github("kjhealy/socviz")
library(socviz)
```

#### by\_country

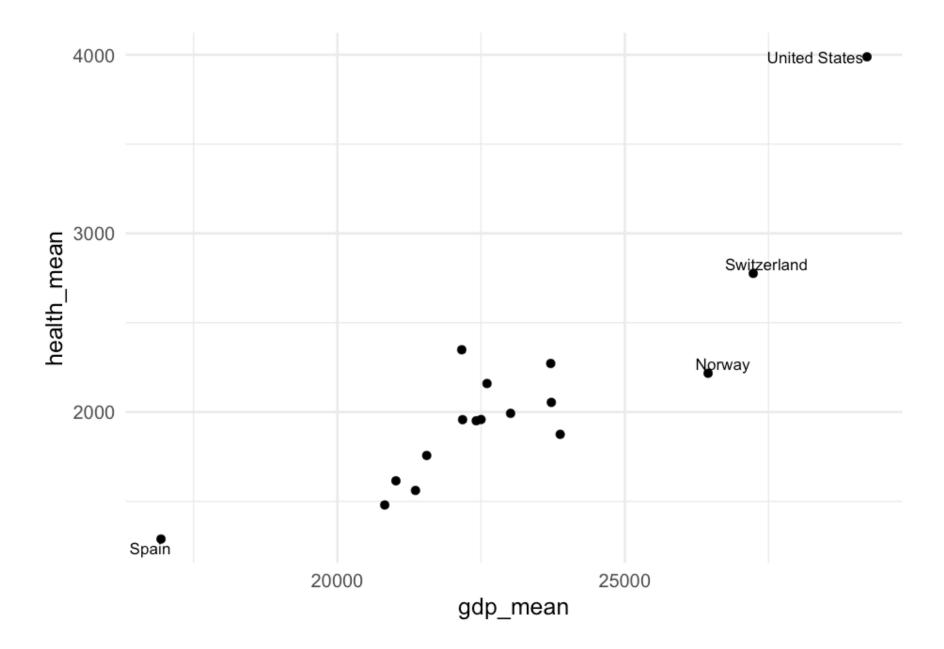
```
## # A tibble: 17 x 8
            consent_law [2]
## # Groups:
    consent_law country donors_mean donors_sd gdp_mean health_mean roads_mean
##
##
    <chr>
                <chr>
                              <dbl>
                                        <dbl>
                                                 <dbl>
                                                             <dbl>
                                                                        <dbl>
## 1 Informed
                Austra...
                           10.635
                                    1.142808 22178.54
                                                          1957.5
                                                                    104.8757
## 2 Informed
                Canada
                           13.96667 0.7511607 23711.08
                                                          2271.929 109.2601
## 3 Informed
                Denmark
                           13.09167 1.468121 23722.31
                                                          2054.071 101.6363
## 4 Informed
                Germany 13.04167 0.6111960 22163.23
                                                          2348.75
                                                                   112.7887
## 5 Informed
                Ireland 19.79167 2.478437 20824.38
                                                          1479.929 117.7742
## 6 Informed
                Nether...
                        13.65833 1.551807 23013.15
                                                          1992.786
                                                                   76.09357
## # ... with 11 more rows, and 1 more variable: cerebvas_mean <dbl>
```

### Scatterplot

```
ggplot(by_country, aes(gdp_mean, health_mean)) +
  geom_point()
```



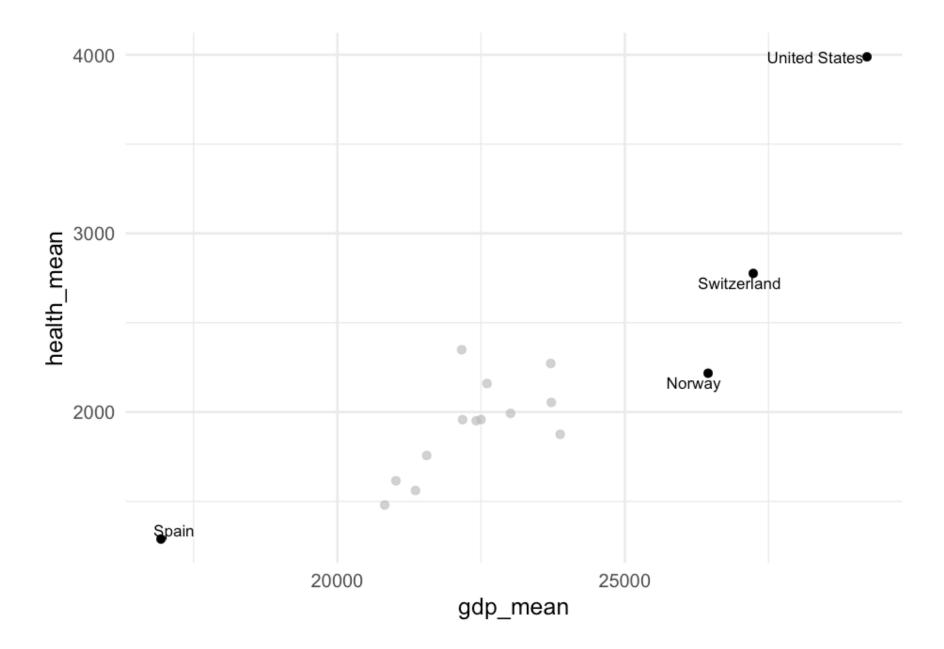
### **Outliers**



### Combine with highlighting

```
library(gghighlight)
ggplot(by_country, aes(gdp_mean, health_mean)) +
  geom_point() +
  gghighlight(gdp_mean > 25000 | gdp_mean < 20000) +
  geom_text_repel(aes(label = country))</pre>
```

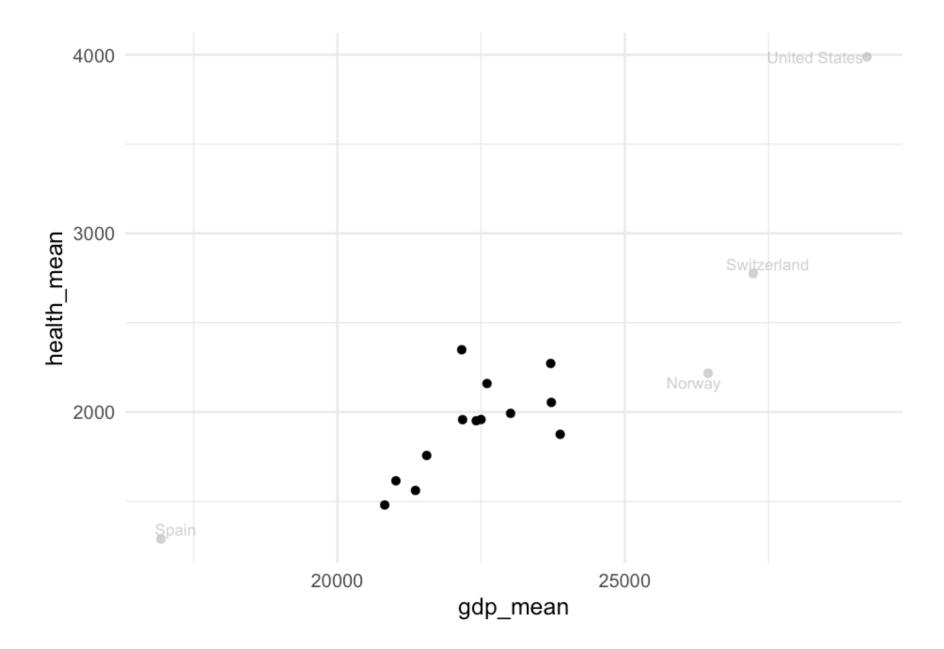
 Notice you only have to specify the points to highlight and geom\_text\_repel will then only label those points



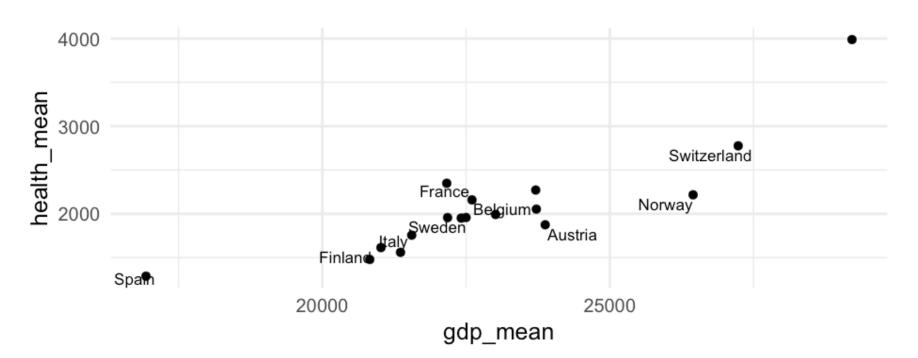
### Combine with highlighting

Switch to make outliers grayed out and labeled

Note I found the exact gray color by looking at the source code. Specifically, it is the output from ggplot2::alpha("grey", 0.7)



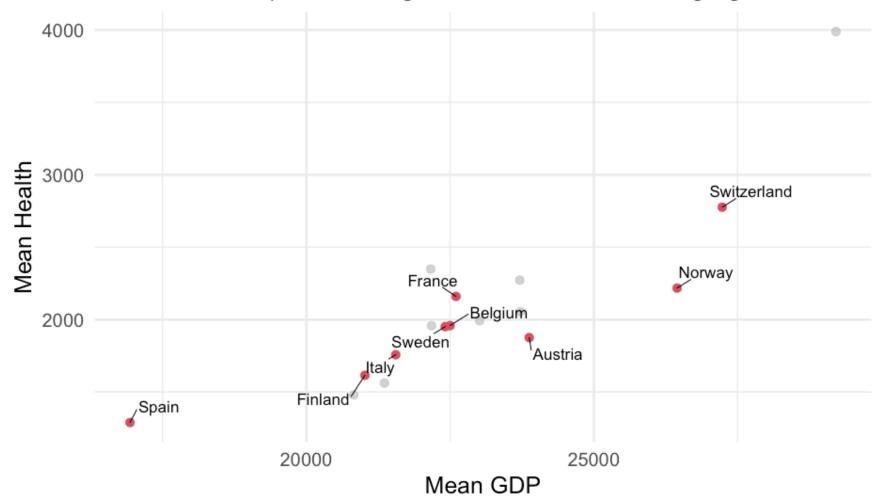
### By group



### By group

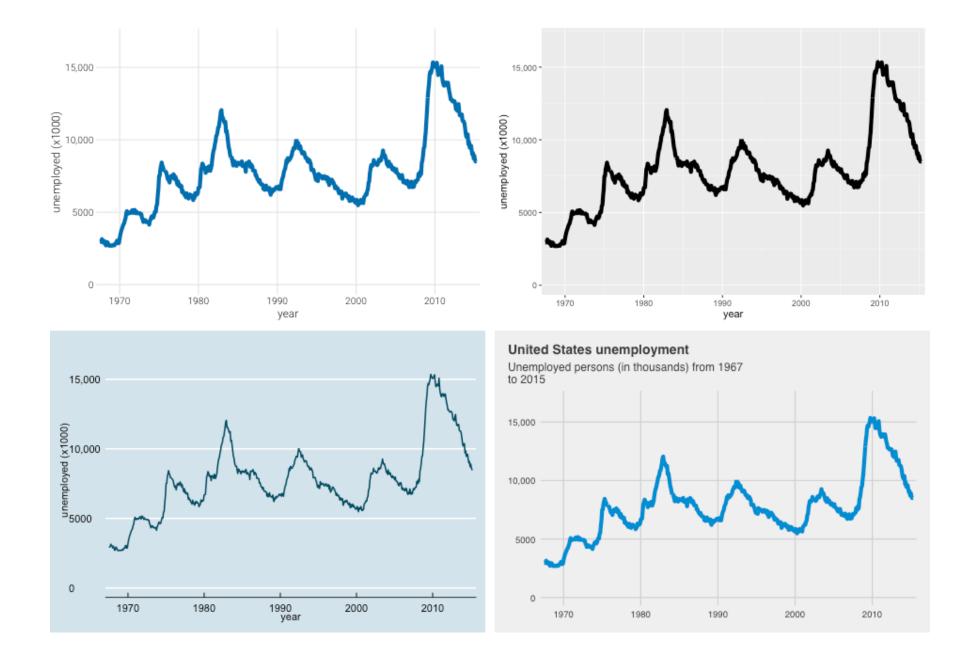
#### **GDP** and Health

Countries with a presumed organ donation consent are highlighted



Data from the General Social Science Survey, Distributed through the socviz R package

# Themes (quickly)



### ggthemes

- Great place to start. All sorts of themes.
- Includes color scales, etc., that align with themes
- You can even conform with other software
  - fit into an economics conference with theme\_stata

See the themes here

### ggthemeassist

- Another great place to start with making major modifications/creating your own custom theme
- Can't do everything, but can do a lot
- See here

(demo)

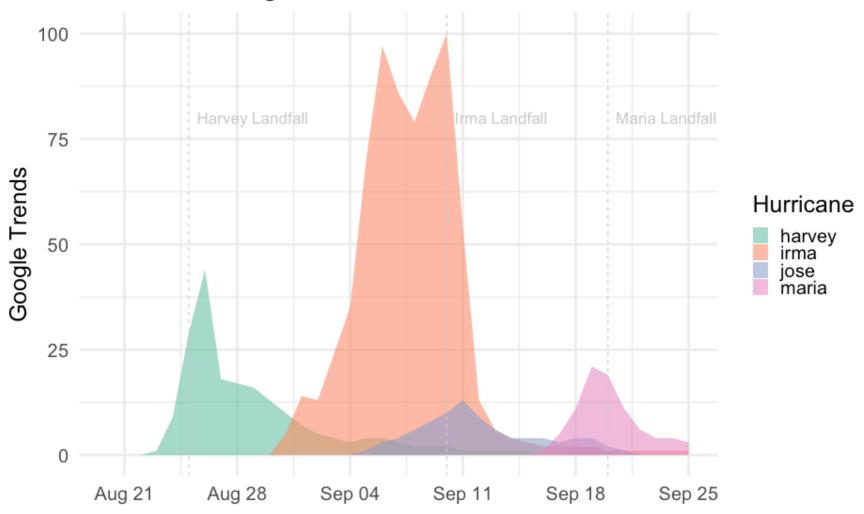
## theme() for everything else

- You can basically change your plot to look however you want through theme
- Generally a bit more complicated
- I've used ggplot for *years* and only really now gaining fluency with it

### Quick example

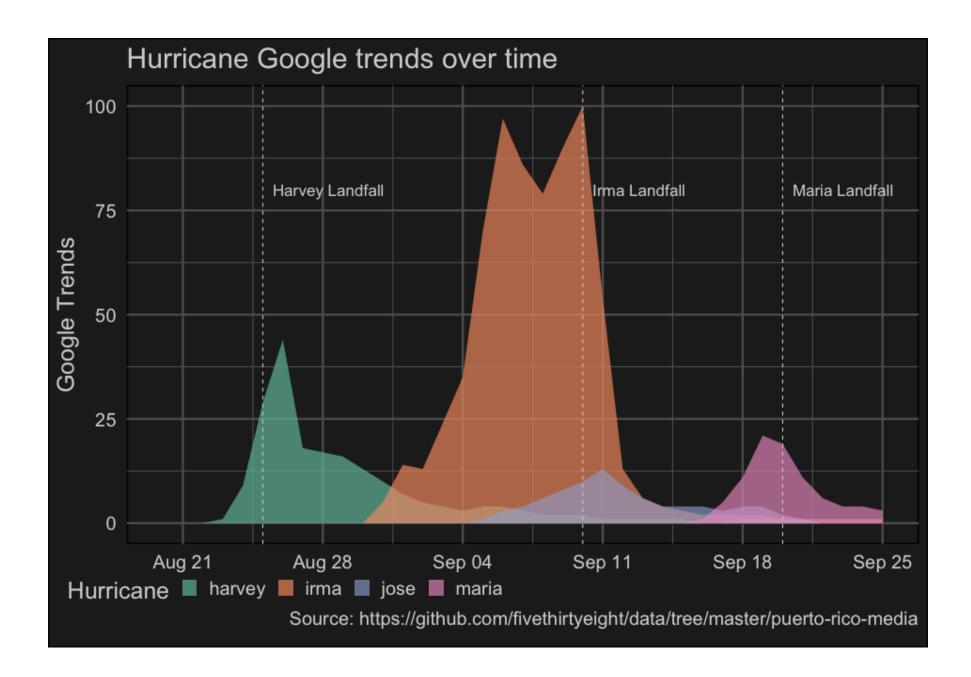
From Lab 3

#### Hurricane Google trends over time



Source: https://github.com/fivethirtyeight/data/tree/master/puerto-rico-media

```
p + theme(panel.grid.major = element line(colour = "gray30"),
          panel.grid.minor = element_line(colour = "gray30"),
          axis.text = element_text(colour = "gray80"),
          axis.text.x = element_text(colour = "gray80"),
          axis.text.y = element text(colour = "gray80"),
          axis.title = element_text(colour = "gray80"),
          legend.text = element_text(colour = "gray80"),
          legend.title = element_text(colour = "gray80"),
          panel.background = element_rect(fill = "gray10"),
          plot.background = element_rect(fill = "gray10"),
          legend.background = element_rect(fill = NA, color = NA),
          legend.position = c(0.20, -0.1),
          legend.direction = "horizontal",
          plot.margin = margin(10, 10, b = 20, 10),
          plot.caption = element_text(colour = "gray80", vjust = 1),
          plot.title = element text(colour = "gray80"))
```



# Next time Our final lab!