

# 1.3 — Data Visualization with ggplot2

ECON 480 • Econometrics • Fall 2020

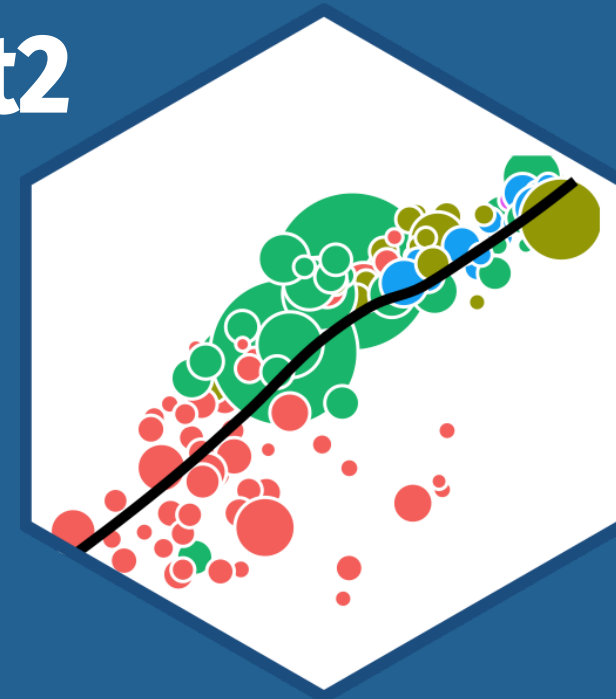
Ryan Safner

Assistant Professor of Economics

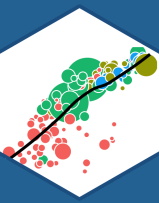
✉ [safner@hood.edu](mailto:safner@hood.edu)

🔗 [ryansafner/metricsF20](https://ryansafner/metricsF20)

🌐 [metricsF20.classes.ryansafner.com](https://metricsF20.classes.ryansafner.com)



# Outline



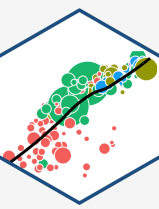
Plotting in Base R

ggplot2 and the tidyverse

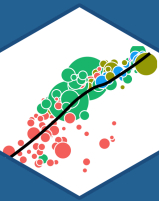
Plot Layers

Some Troubleshooting

# Graphics and Statistics

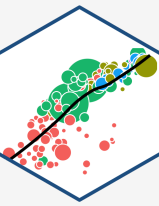


- Admittedly, we still need to cover basic descriptive statistics and data fundamentals
  - continuous, discrete, cross-sectional, time series, panel data
  - mean, median, variance, standard deviation
  - random variables, distributions, PDFs, Z-scores
  - bargraphs, boxplots, histograms, scatterplots
- All of this is coming in 2 weeks as we return to statistics and econometric theory
- But let's start with the fun stuff right away, even if you don't fully know the *reasons*: **data visualization**



# Plotting in Base R

# Our Data Source



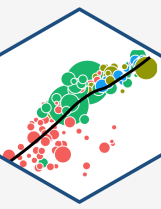
- For our examples, we'll use a dataset `mpg` from the `ggplot2` library

```
library(ggplot2)
```

```
head(mpg)
```

```
## # A tibble: 6 x 11
##   manufacturer model displ  year  cyl trans      drv   cty   hwy fl   class
##   <chr>          <chr> <dbl> <int> <int> <chr>    <chr> <int> <int> <chr> <chr>
## 1 audi          a4     1.8  1999   4 auto(l5)  f     18    29 p    compa...
## 2 audi          a4     1.8  1999   4 manual(m5) f     21    29 p    compa...
## 3 audi          a4     2    2008   4 manual(m6) f     20    31 p    compa...
## 4 audi          a4     2    2008   4 auto(av)  f     21    30 p    compa...
## 5 audi          a4     2.8  1999   6 auto(l5)  f     16    26 p    compa...
## 6 audi          a4     2.8  1999   6 manual(m5) f     18    26 p    compa...
```

# Plotting in Base R



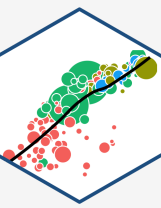
- Base R is very powerful and intuitive to plot, but not very sexy
- Basic syntax for most types of plots:

```
plot_type(my_df$variable)
```

- If using multiple variables, you can avoid typing `$` by just typing the variable names and then in another argument to the plotting function, specify `data = my_df`

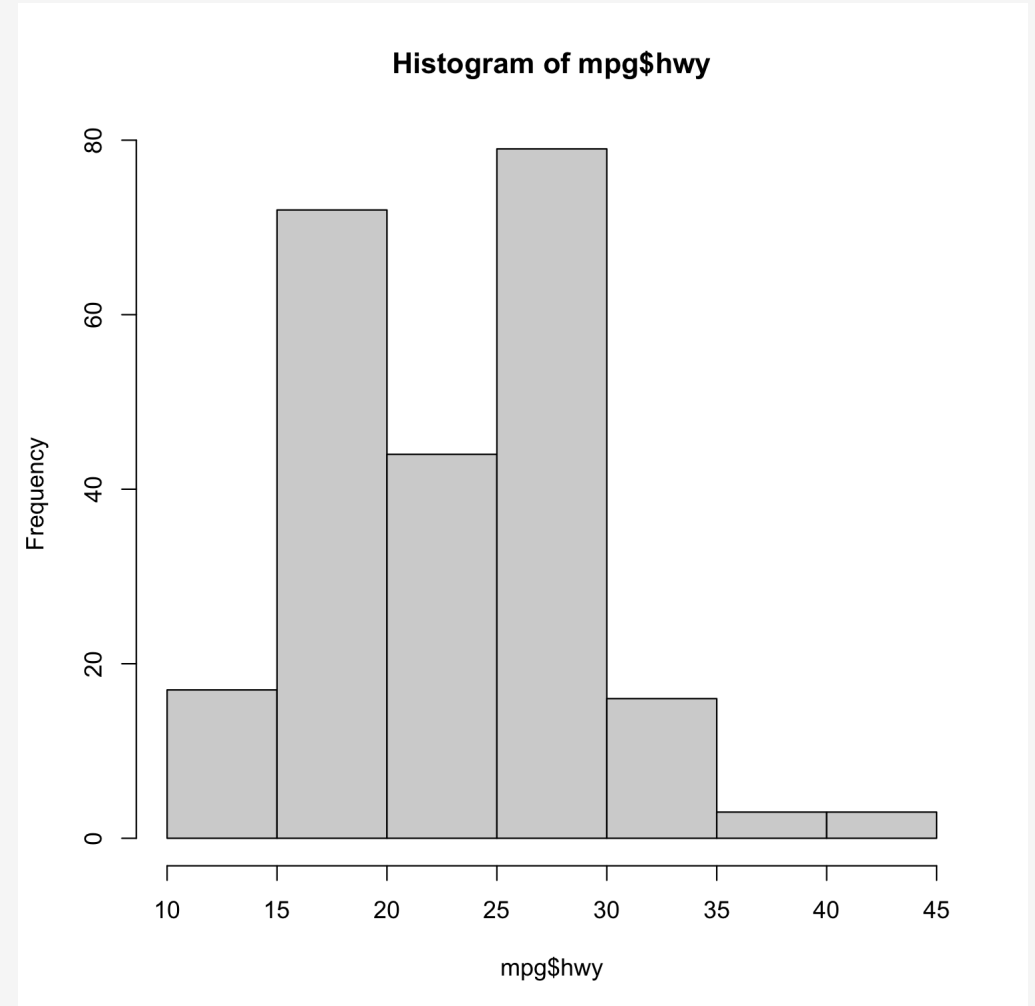
```
plot_type(my_df$variable1, my_df$variable2, data = my_df)
```

# Plotting in Base R: Histogram

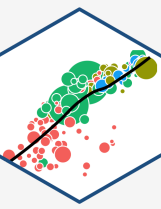


- Using the `mpg` data, plotting a **histogram** of `hwy`

```
hist(mpg$hwy)
```

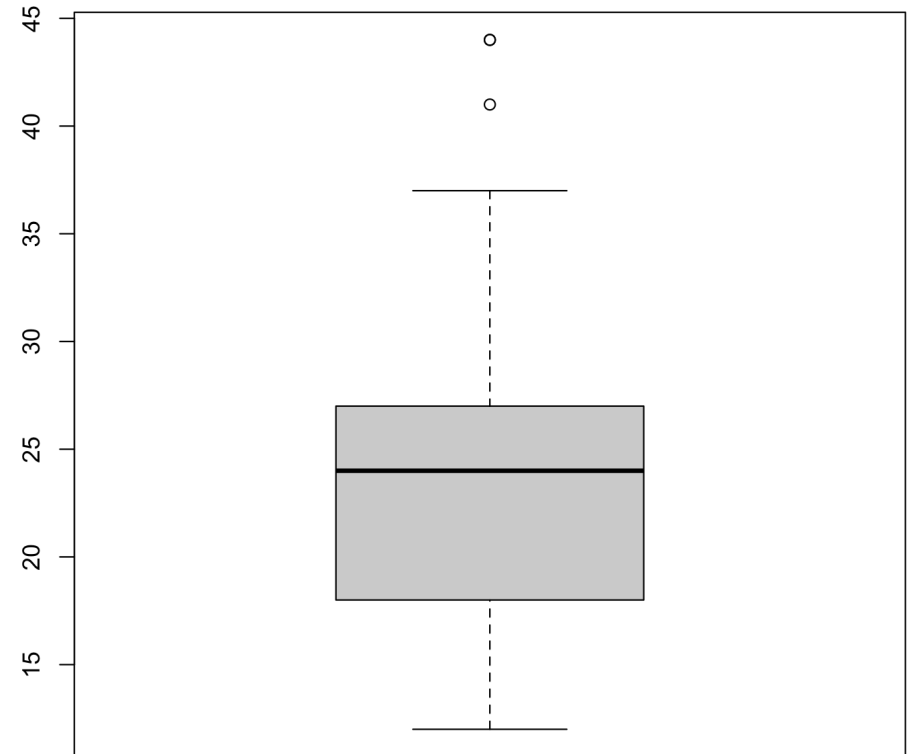


# Plotting in Base R: Boxplot



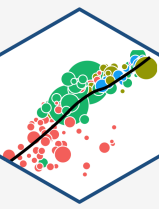
- Using the `mpg` data, plotting a **boxplot** of `hwy`

```
boxplot(mpg$hwy)
```





# Plotting in Base R: Boxplot by Category



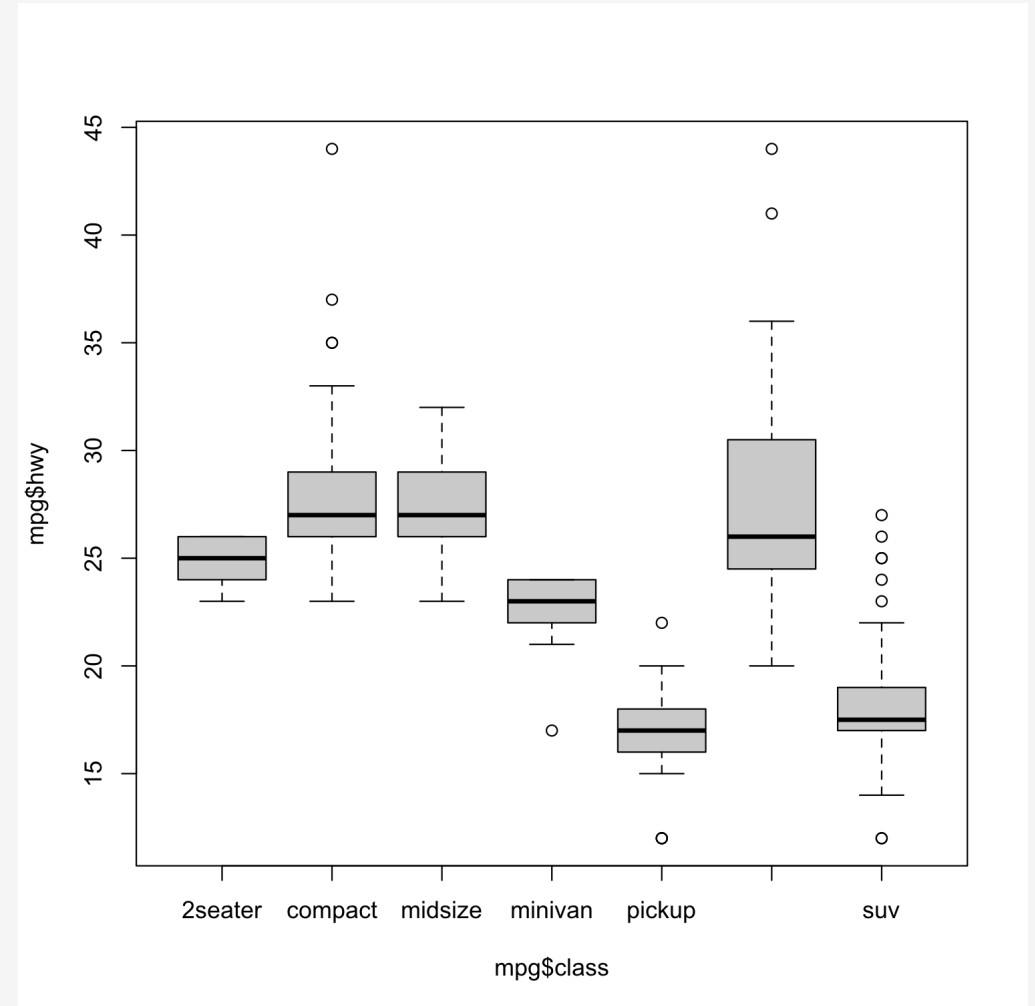
- Using the `mpg` data, plotting a **boxplot** of `hwy` by class

```
boxplot(mpg$hwy ~ mpg$class)
```

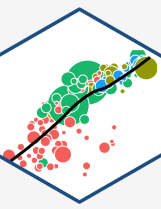
```
# second method
```

```
boxplot(mpg ~ class, data = mtcars)
```

- The `~` is part of R's **“formula notation”**:
  - Dependent variable goes to left
  - Independent variable(s) to right, separated with `+`'s
  - Think `y~x+z` means “`y` is explained by `x` and `z`”



# Plotting in Base R: Scatterplot

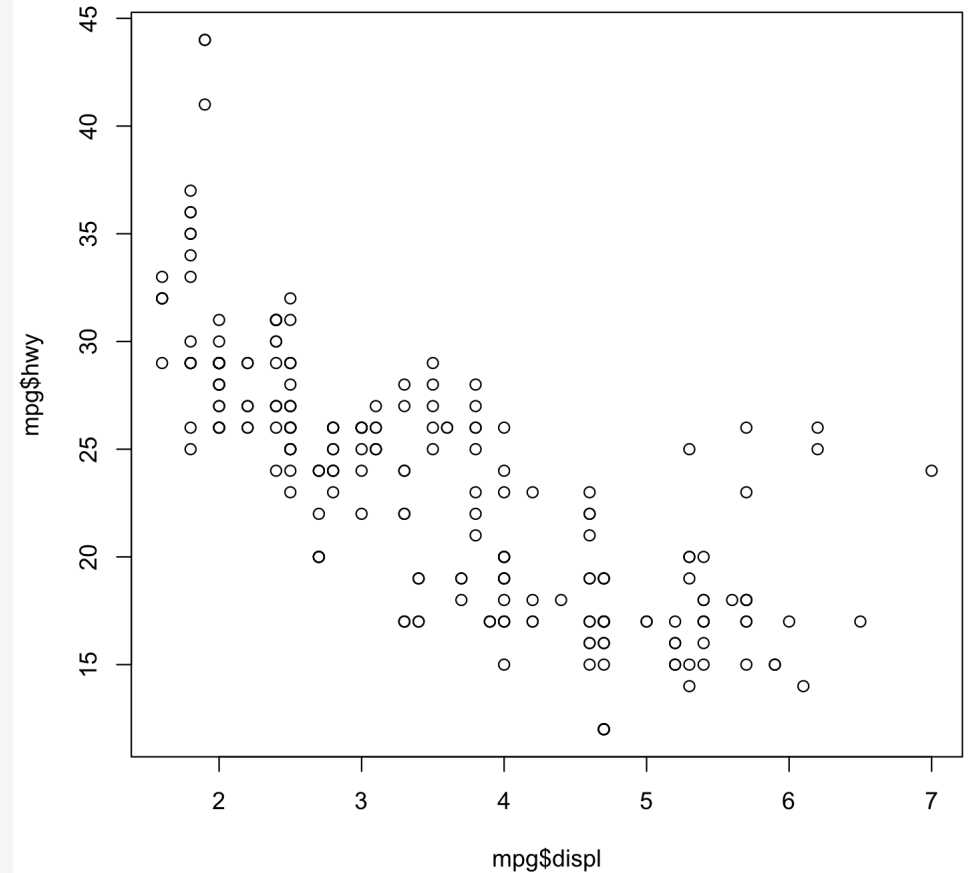


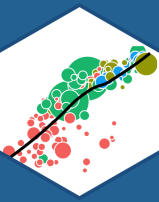
- Using the `mpg` data, plotting a **scatterplot** of `hwy` against `displ`

```
plot(mpg$hwy ~ mpg$displ)
```

```
# second method
```

```
plot(hwy ~ displ, data = mpg)
```



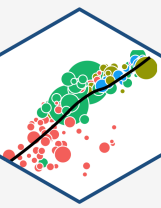


# ggplot2 and the tidyverse



tidyverse

# The tidyverse

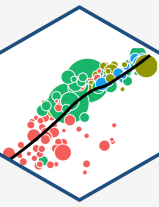


"The tidyverse is an opinionated collection of R packages designed for data science. All packages share an underlying design philosophy, grammar, and data structures.

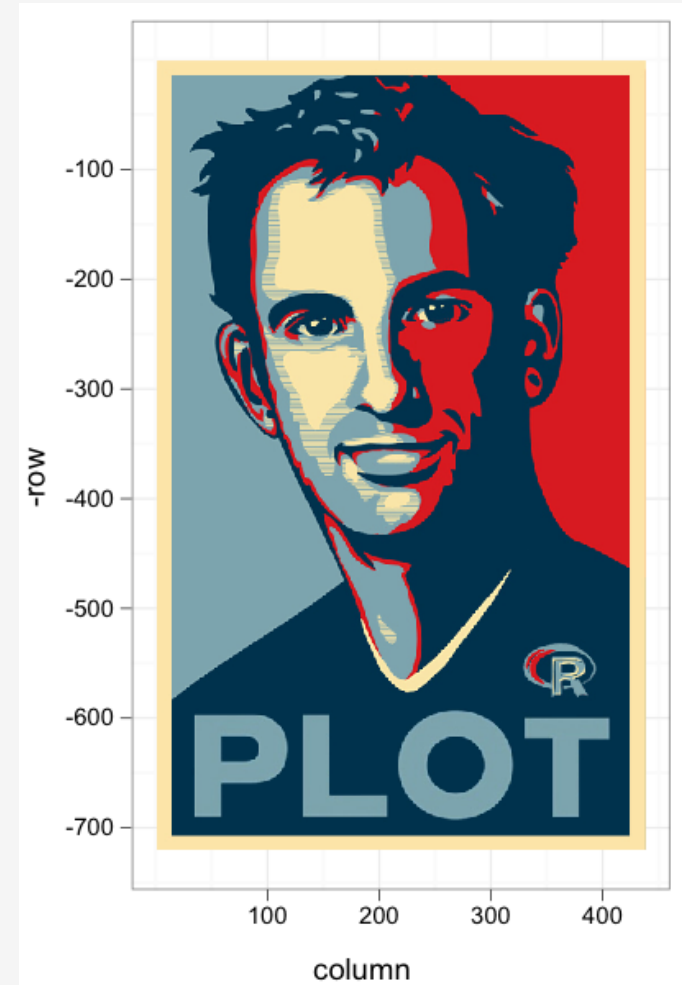
- Largely (but not only) created by Hadley Wickham
- We will look at this much more extensively next week!
- This "flavor" of R will make your coding life *so much easier!*



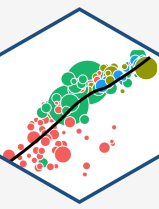
# ggplot



- `ggplot2` is perhaps the most popular package in R and a core element of the `tidyverse`
- `gg` stands for a **grammar of graphics**
- Very powerful and beautiful graphics, very customizable and reproducible, but requires a bit of a learning curve
- All those "cool graphics" you've seen in the New York Times, fivethirtyeight, the Economist, Vox, etc use the grammar of graphics



# ggplot: All Your Figure are Belong to Us

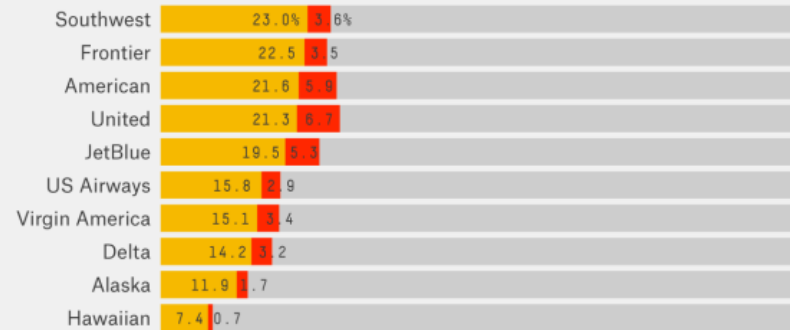


## Southwest's Delays Are Short; United's Are Long

As share of scheduled flights, 2014

● FLIGHTS DELAYED 15-119 MINUTES

● FLIGHTS DELAYED 120+ MINUTES, CANCELED OR DIVERTED



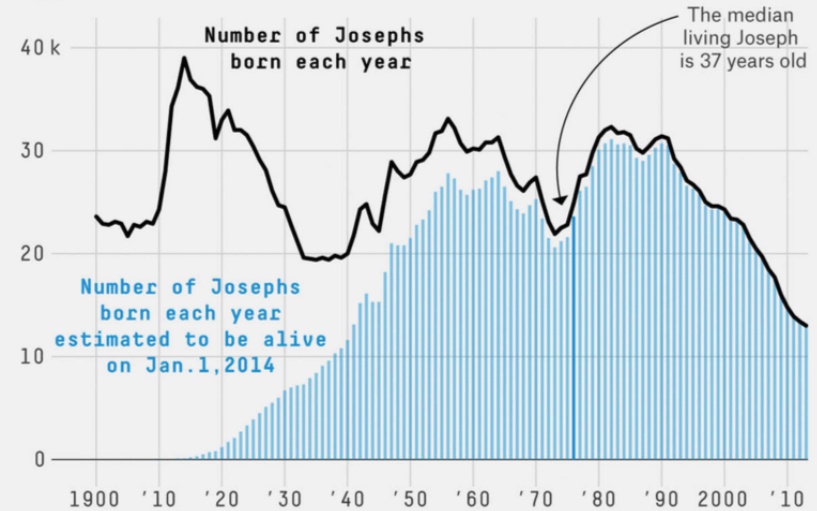
FIVETHIRTYEIGHT

BASED ON DATA FROM THE BUREAU OF TRANSPORTATION STATISTICS

Source: [fivethirtyeight](#)

## Age Distribution of American Boys Named Joseph

By year of birth

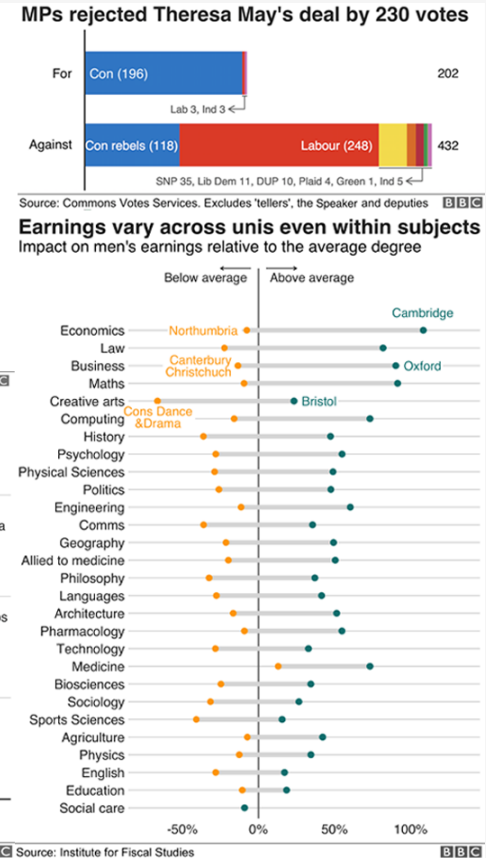
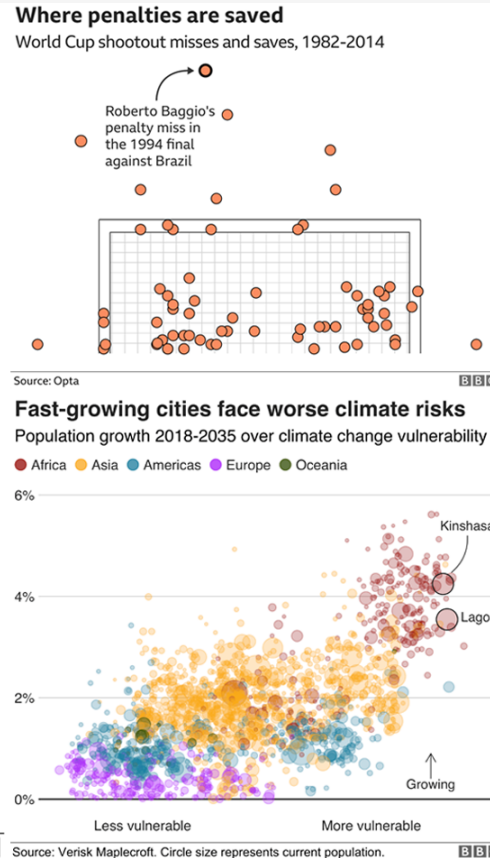
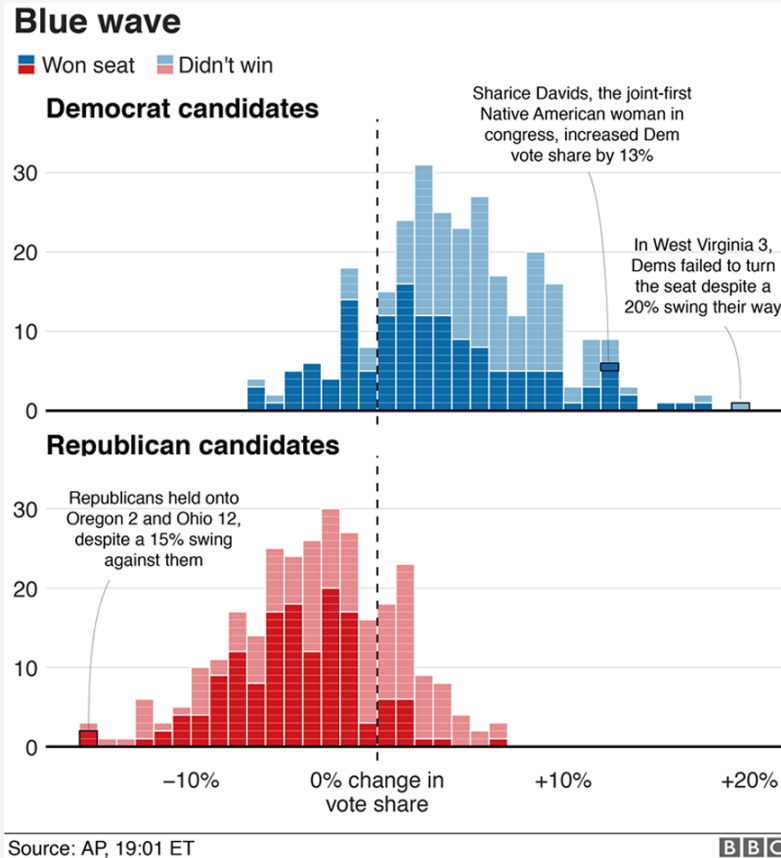
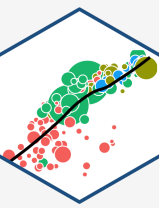


FIVETHIRTYEIGHT

SOURCE: SOCIAL SECURITY ADMINISTRATION

Source: [fivethirtyeight](#)

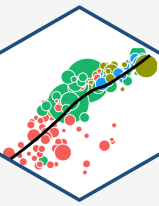
# ggplot: All Your Figure are Belong to Us



Source: [BBC's bbplot](#)



# Why Go gg?



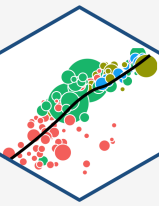
Hadley Wickham

Chief Scientist, R Studio

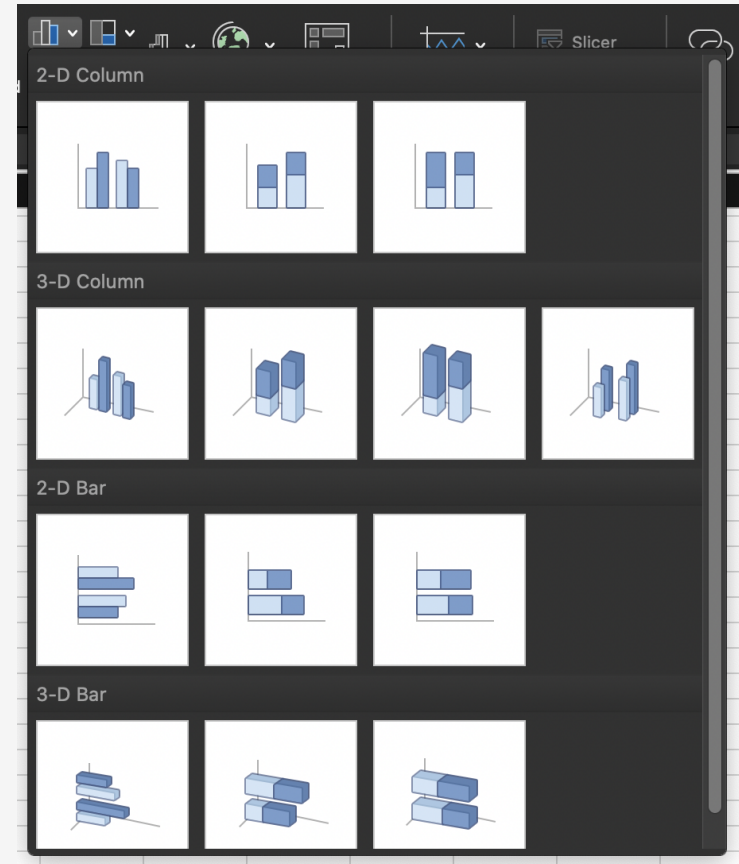
"The transferrable skills from ggplot2 are not the idiosyncracies of plotting syntax, but a powerful way of thinking about visualisation, as a way of **mapping between variables and the visual properties of geometric objects** that you can perceive."

<http://disq.us/p/sv640d>

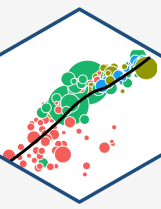
# The Grammar of Graphics (gg)



- This is a true *grammar*
- We *don't* talk about specific chart **types**
  - That you have to hunt through in Excel and reshape your data to fit it
- Instead we talk about specific chart **components**



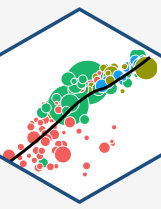
# The Grammar of Graphics (gg) I



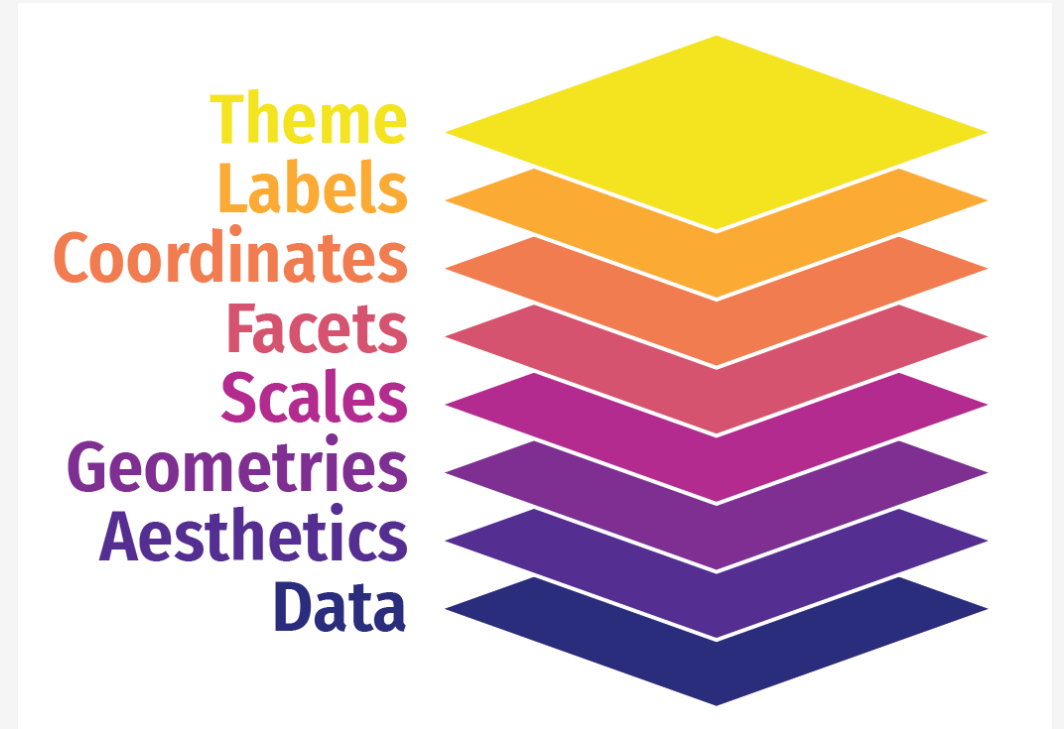
- Any graphic can be built from the same components:
  1. **Data** to be drawn from
  2. **Aesthetic mappings** from data to some visual marking
  3. **Geometric objects** on the plot
  4. **Scales** define the range of values
  5. **Coordinates** to organize location
  6. **Labels** describe the scale and markings
  7. **Facets** group into subplots
  8. **Themes** style the plot elements
- Not every plot needs *every* component, but all plots *must* have the first 3!



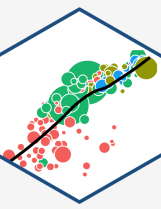
# The Grammar of Graphics (gg) II



- Any graphic can be built from the same components:
  1. `data` to be drawn from
  2. `aes` **thetic mappings** from data to some visual marking
  3. `geom` **metric objects** on the plot
  4. `scale` define the range of values
  5. `coord` **inates** to organize location
  6. `label`s describe the scale and markings
  7. `facet` group into subplots
  8. `theme` style the plot elements
- Not every plot needs *every* component, but all plots *must* have the first 3!



# The Grammar of Graphics (gg): All at Once

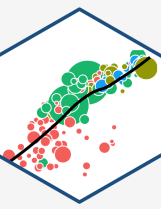


## All in one command

- Produces plot output in viewer
- Does not save plot
  - Save with **Export** menu in viewer
- Adding layers requires whole code for new plot

```
ggplot(data = mpg)+  
  aes(x = displ,  
       y = hwy)+  
  geom_point()+  
  geom_smooth()
```

# The Grammar of Graphics (gg): As R Objects



## Saving as an object

- Saves your plot as an **R** object
- Does *not* show in viewer
  - Execute the name of your object to see it
- Can add layers by calling the original plot name

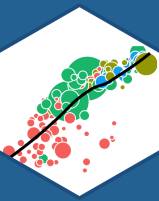
```
# make and save plot
p <- ggplot(data = mpg)+
  aes(x = displ,
       y = hwy)+
  geom_point()

p # view plot

# add a layer

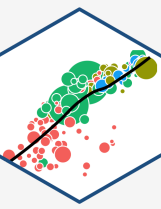
p + geom_smooth() # shows new plot

p <- p + geom_smooth() # saves and overwrites p
p2 <- p + geom_smooth() # saves as different ob
```



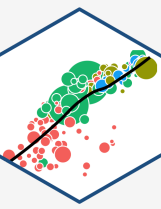
# Plot Layers

# The Grammar of Graphics





# The Grammar of Graphics (gg): Data



Data

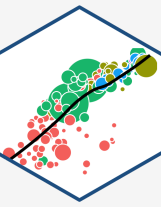
```
ggplot(data = mpg)
```

**Data** is the source of our data. As part of the `tidyverse`, `ggplot2` requires data to be "tidy"<sup>1</sup>:

1. Each variable forms a **column**
2. Each observation forms a **row**
3. Each observational unit forms a table

<sup>1</sup> Data "tidyness" is the core element of all `tidyverse` packages. Much more on all of this next class.

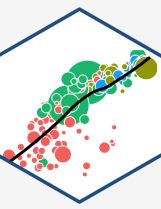
# The Grammar of Graphics (gg): Adding Layers



## Data

- Add a layer with `+` at the end of a line (never at the beginning!)
- Style recommendation: start a new line after each `+` to improve legibility!
- We will build a plot layer-by-layer

# The Grammar of Graphics (gg): Aesthetics I

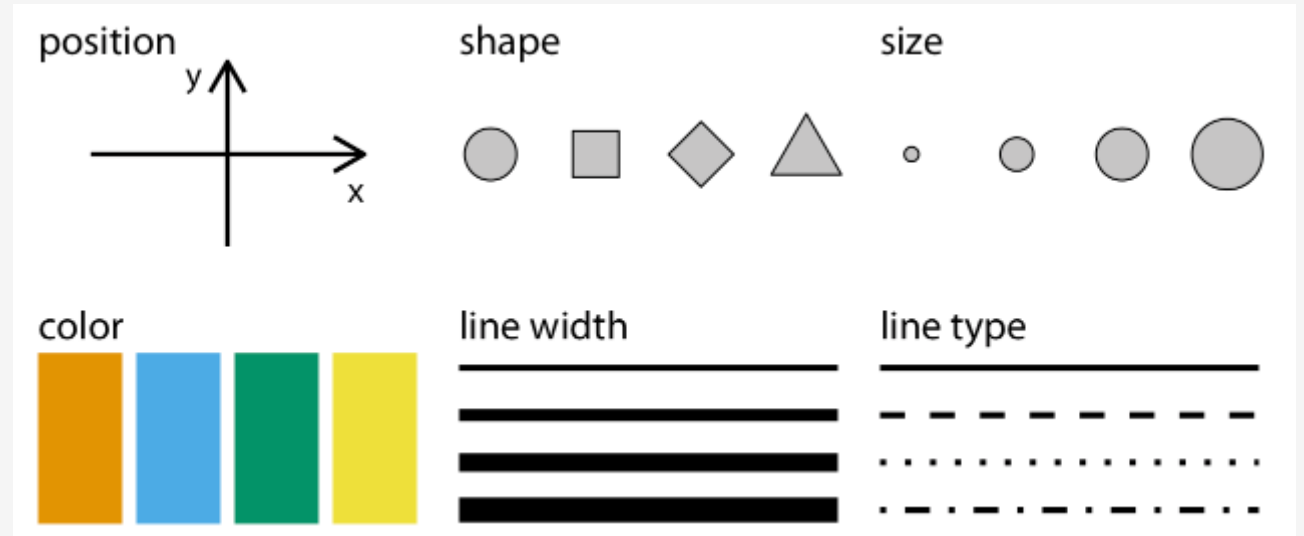


Data

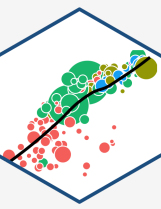
**Aesthetics** map data to visual elements or parameters

Aesthetics

+ `aes()`



# The Grammar of Graphics (gg): Aesthetics II

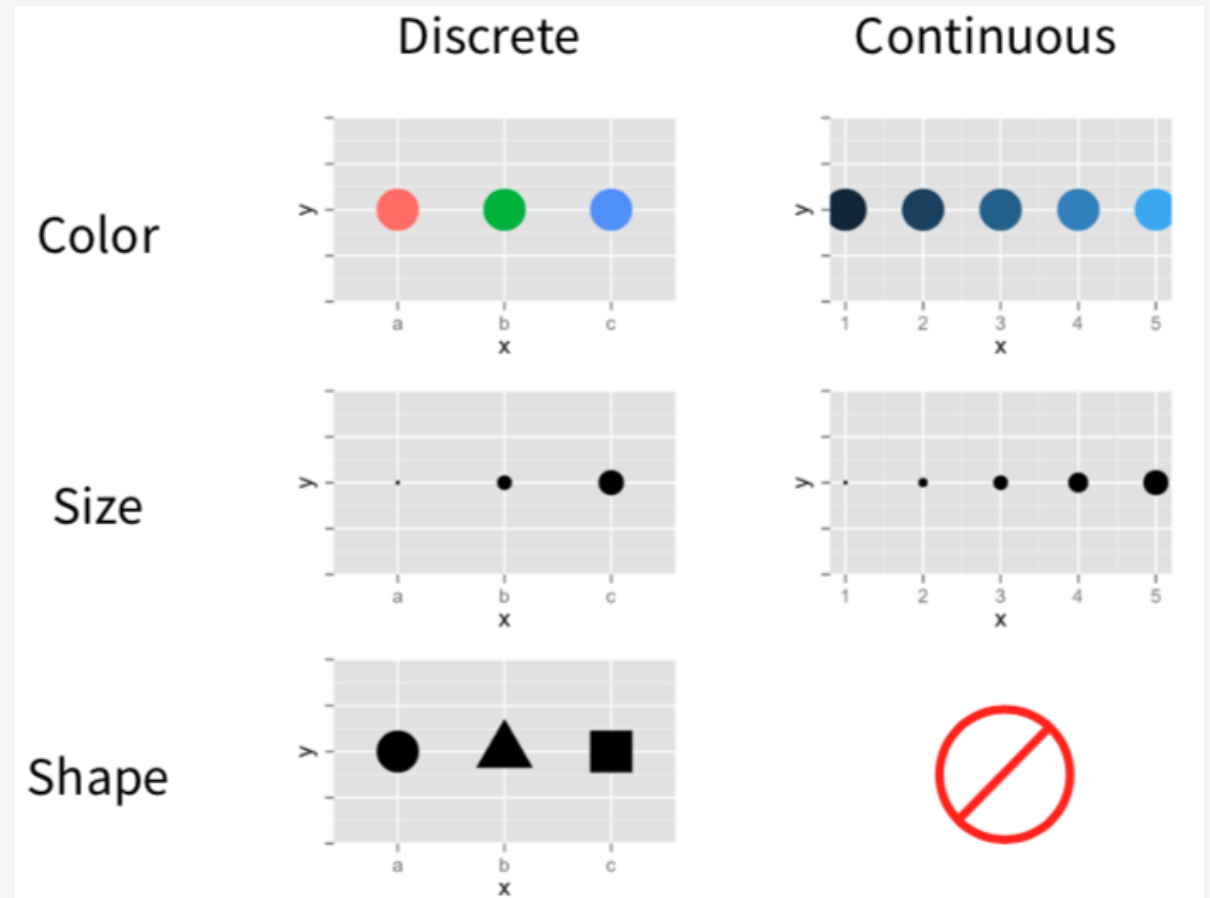


Data

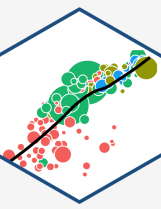
**Aesthetics** map data to visual elements or parameters

Aesthetics

+ `aes()`



# The Grammar of Graphics (gg): Aesthetics III



Data

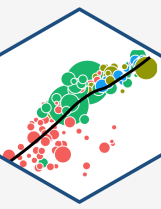
**Aesthetics** map data to visual elements or parameters

Aesthetics

- `displ`
- `hwy`
- `class`

+ `aes()`

# The Grammar of Graphics (gg): Aesthetics III



Data

**Aesthetics** map data to visual elements or parameters

Aesthetics

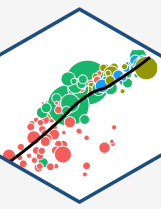
- `displ` → **x**

- `hwy` → **y**

- `class` → *shape, size, color*, etc.

+ `aes()`

# The Grammar of Graphics (gg): Aesthetics IV

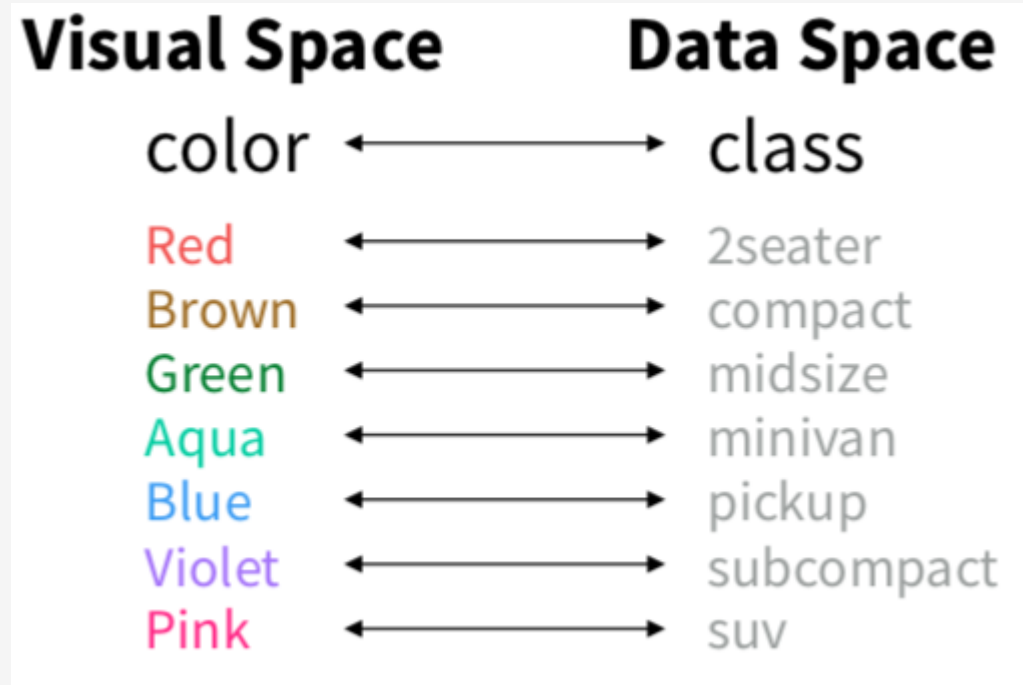


Data

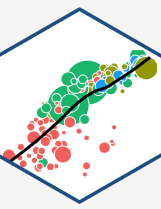
**Aesthetics** map data to visual elements or parameters

Aesthetics

+ `aes()`



# The Grammar of Graphics (gg): Aesthetics IV



Data

**Aesthetics** map data to visual elements or parameters

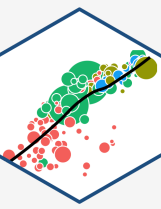
Aesthetics

+ aes()

```
aes(x = displ,  
    y = hwy,  
    color = class)
```



# The Grammar of Graphics (gg): Geoms I



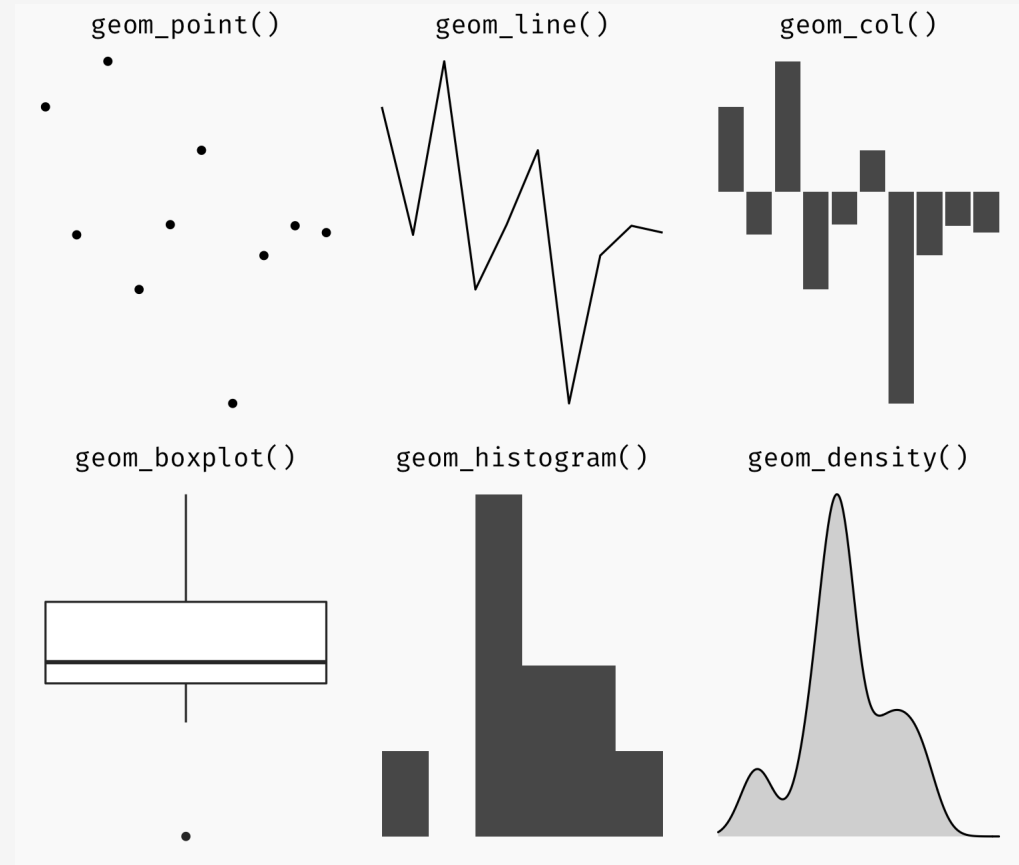
Data

**Geometric objects** displayed on the plot

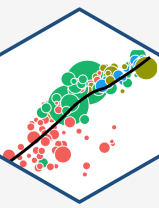
Aesthetics

Geoms

+ `geom_*()`



# The Grammar of Graphics (gg): Geoms II



Data

**Geometric objects** displayed on the plot

Aesthetics

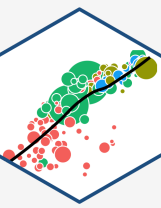
- What `geom`s you should use depends on what you want to show:

Geoms

+ `geom_*()`

Type	geom
Point	<code>geom_point()</code>
Line	<code>geom_line()</code> , <code>geom_path()</code>
Bar	<code>geom_bar()</code> , <code>geom_col()</code>
Histogram	<code>geom_histogram()</code>
Regression	<code>geom_smooth()</code>
Boxplot	<code>geom_boxplot()</code>

# The Grammar of Graphics (gg): Geoms III



Data

Aesthetics

Geoms

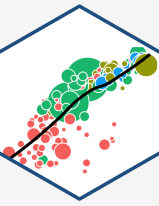
+ `geom_*()`

**Geometric objects** displayed on the plot

```
## [1] "geom_abline"      "geom_area"        "geom_bar"
## [5] "geom_blank"      "geom_boxplot"    "geom_col"
## [9] "geom_count"      "geom_crossbar"   "geom_curve"
## [13] "geom_density_2d" "geom_density2d"  "geom_dotplot"
## [17] "geom_errorbarh"  "geom_freqpoly"   "geom_hex"
## [21] "geom_hline"      "geom_jitter"     "geom_label"
## [25] "geom_linerange"  "geom_map"        "geom_path"
## [29] "geom_pointrange" "geom_polygon"    "geom_qq"
## [33] "geom_quantile"   "geom_raster"     "geom_rect"
## [37] "geom_rug"        "geom_segment"    "geom_sf"
## [41] "geom_sf_text"    "geom_smooth"     "geom_spoke"
## [45] "geom_text"       "geom_tile"       "geom_violin"
```

See <http://ggplot2.tidyverse.org/reference> for many more options

# The Grammar of Graphics (gg): Geoms IV



Data

**Geometric objects** displayed on the plot

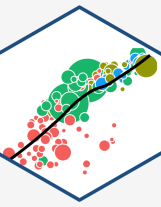
Aesthetics

Or just start typing `geom_` in R Studio!

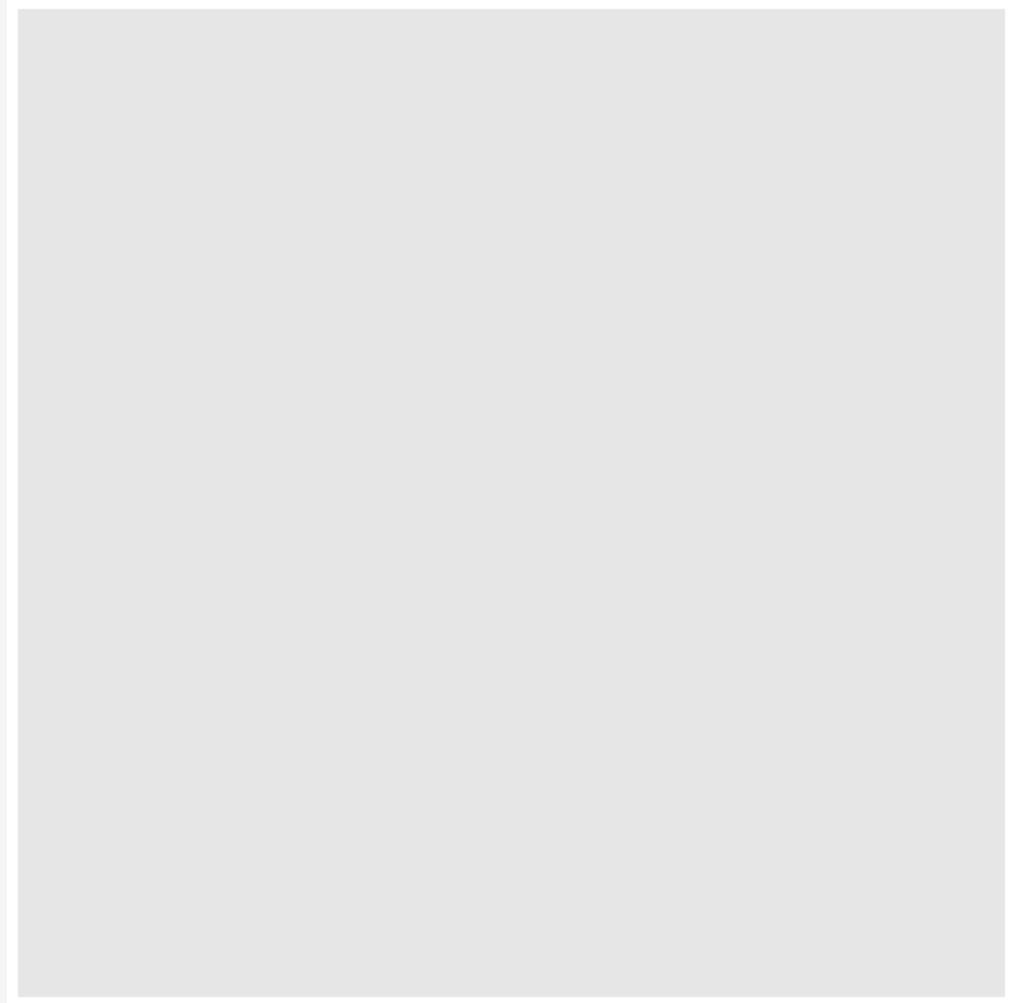
Geoms

+ `geom_*()`

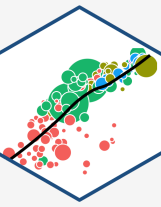
# Let's Make a Plot!



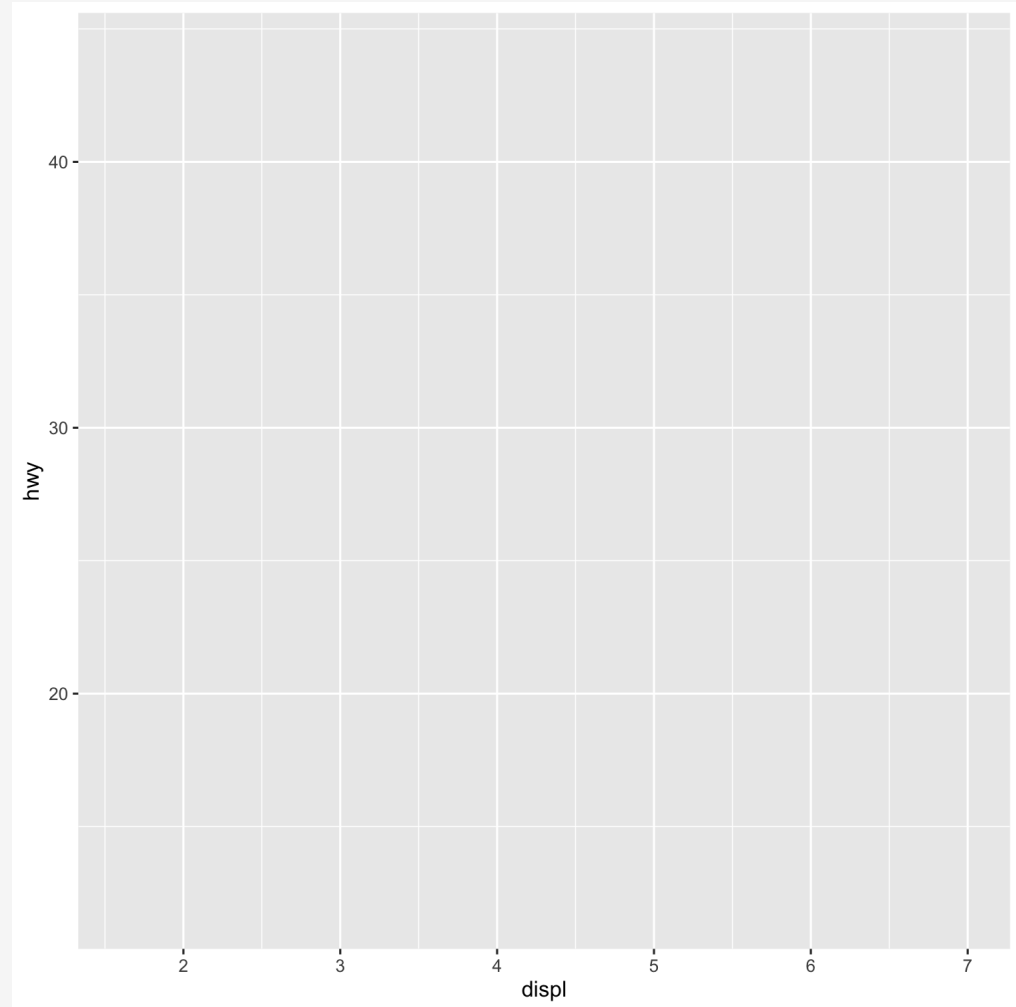
```
ggplot(data = mpg)
```



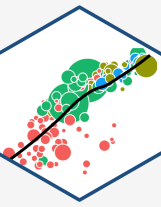
# Let's Make a Plot!



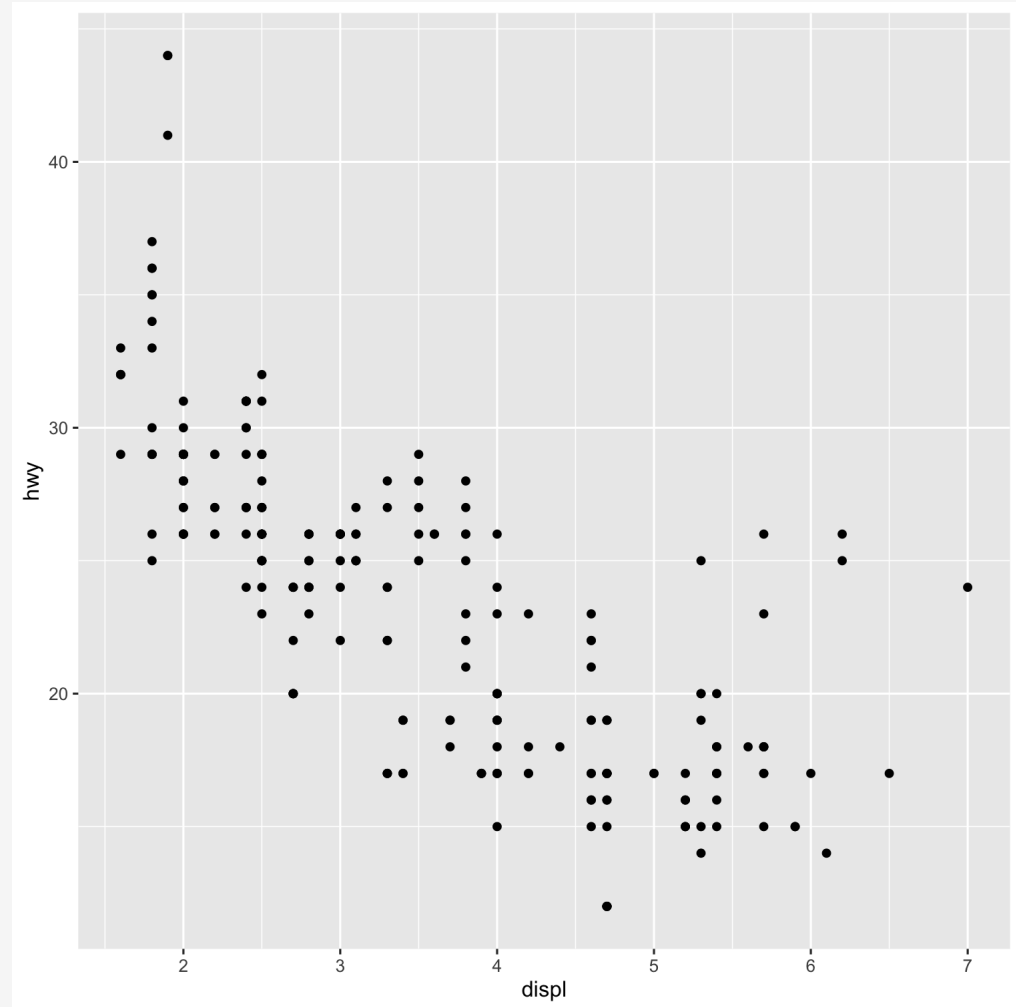
```
ggplot(data = mpg)+  
  aes(x = displ,  
      y = hwy)
```



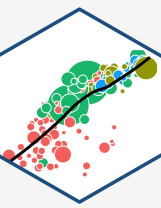
# Let's Make a Plot!



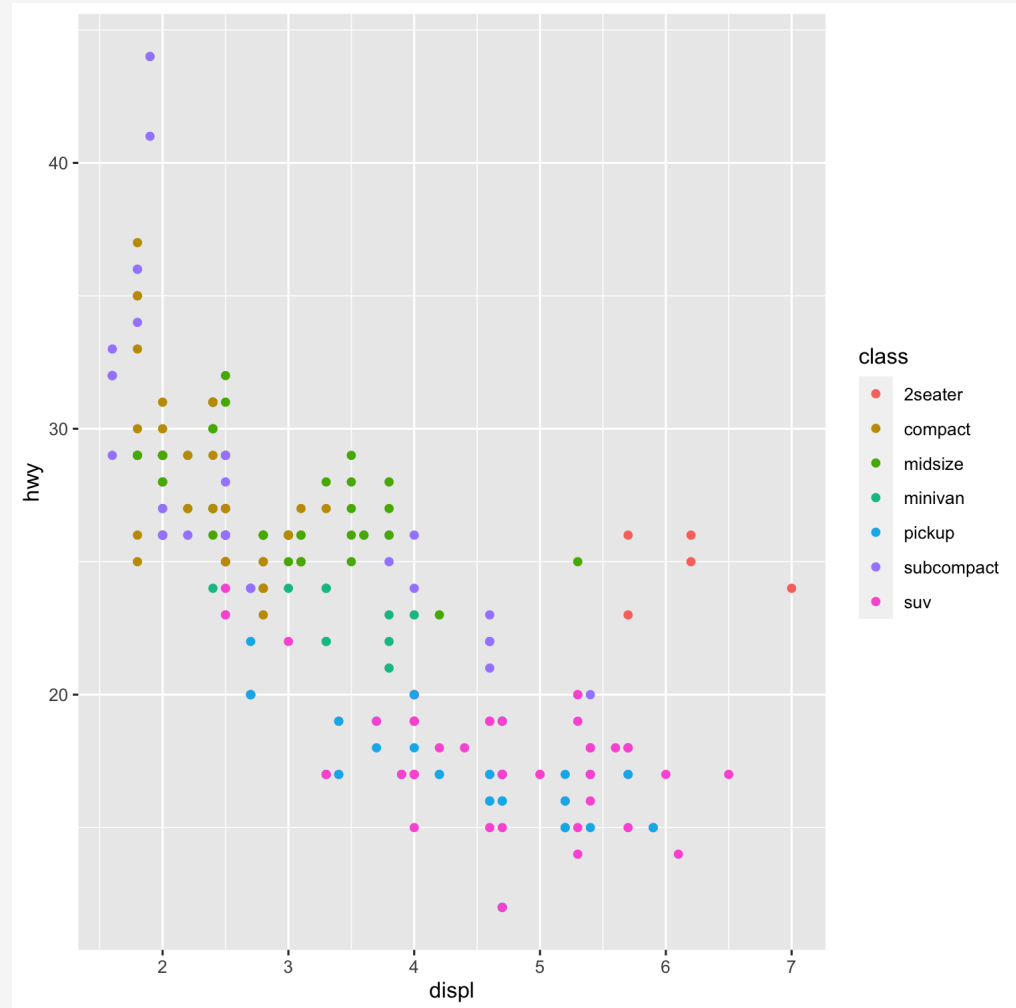
```
ggplot(data = mpg)+  
  aes(x = displ,  
      y = hwy)+  
  geom_point()
```



# Let's Make a Plot!

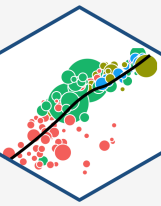


```
ggplot(data = mpg)+  
  aes(x = displ,  
      y = hwy)+  
  geom_point(aes(color = class))
```

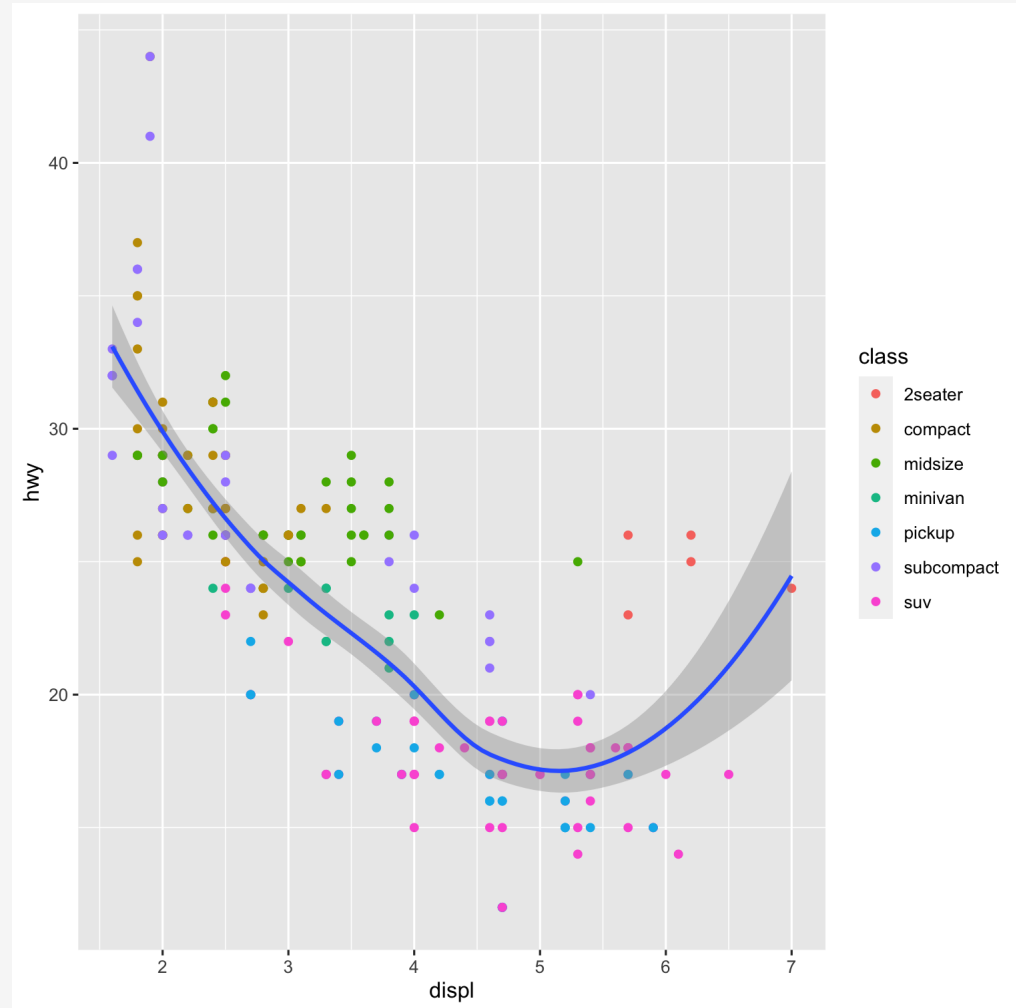




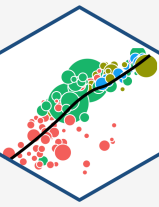
# Let's Make a Plot!



```
ggplot(data = mpg)+  
  aes(x = displ,  
       y = hwy)+  
  geom_point(aes(color = class))+  
  geom_smooth()
```



# More Geoms



Data

```
geom_*(aes, data, stat, position)
```

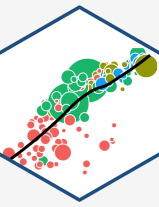
Aesthetics

Geoms

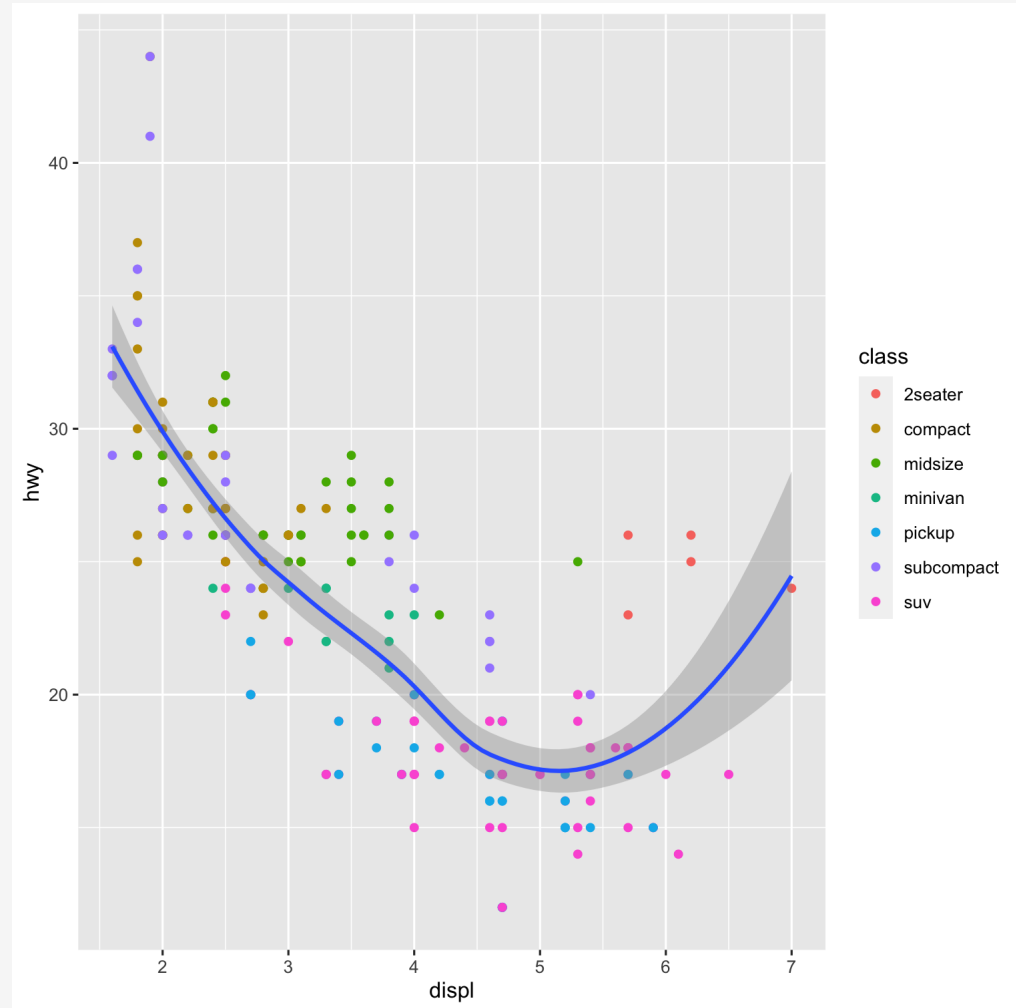
+ `geom_*()`

- `data`: geoms can have their own data
  - has to map onto global coordinates
- `aes`: geoms can have their own aesthetics
  - inherits global aesthetics by default
  - different geoms have different available aesthetics

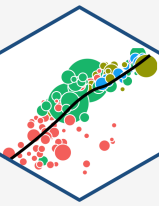
# Change Our Plot



```
ggplot(data = mpg)+  
  aes(x = displ,  
      y = hwy)+  
  geom_point(aes(color = class))+  
  geom_smooth()
```



# More Geoms II



Data

```
geom_*(aes, data, stat, position)
```

Aesthetics

- `stat`: some geoms statistically transform data

- `geom_histogram()` uses `stat_bin()` to group observations into bins

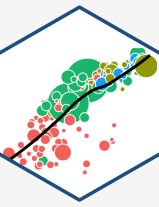
Geoms

- `position`: some adjust location of objects

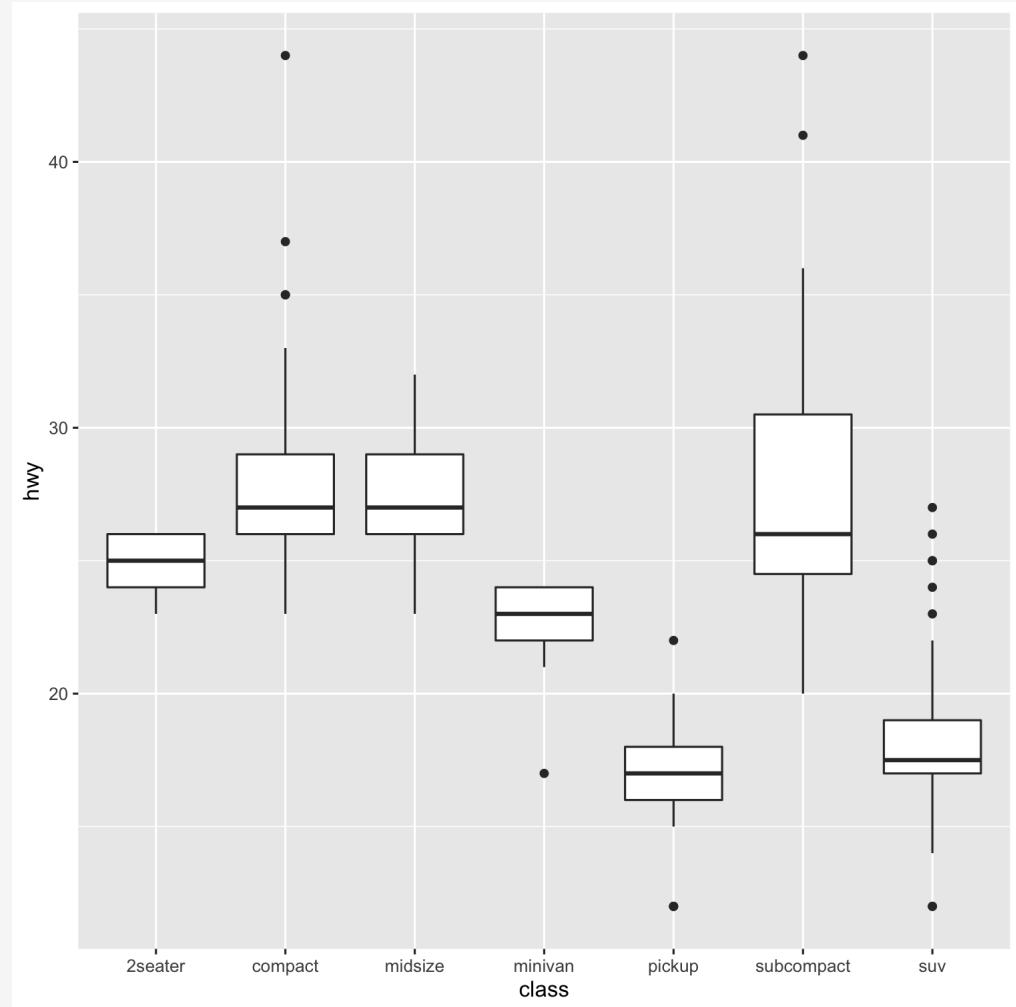
- `dodge`, `stack`, `jitter`

+ `geom_*()`

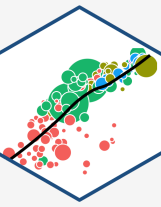
# Let's Change Our Plot



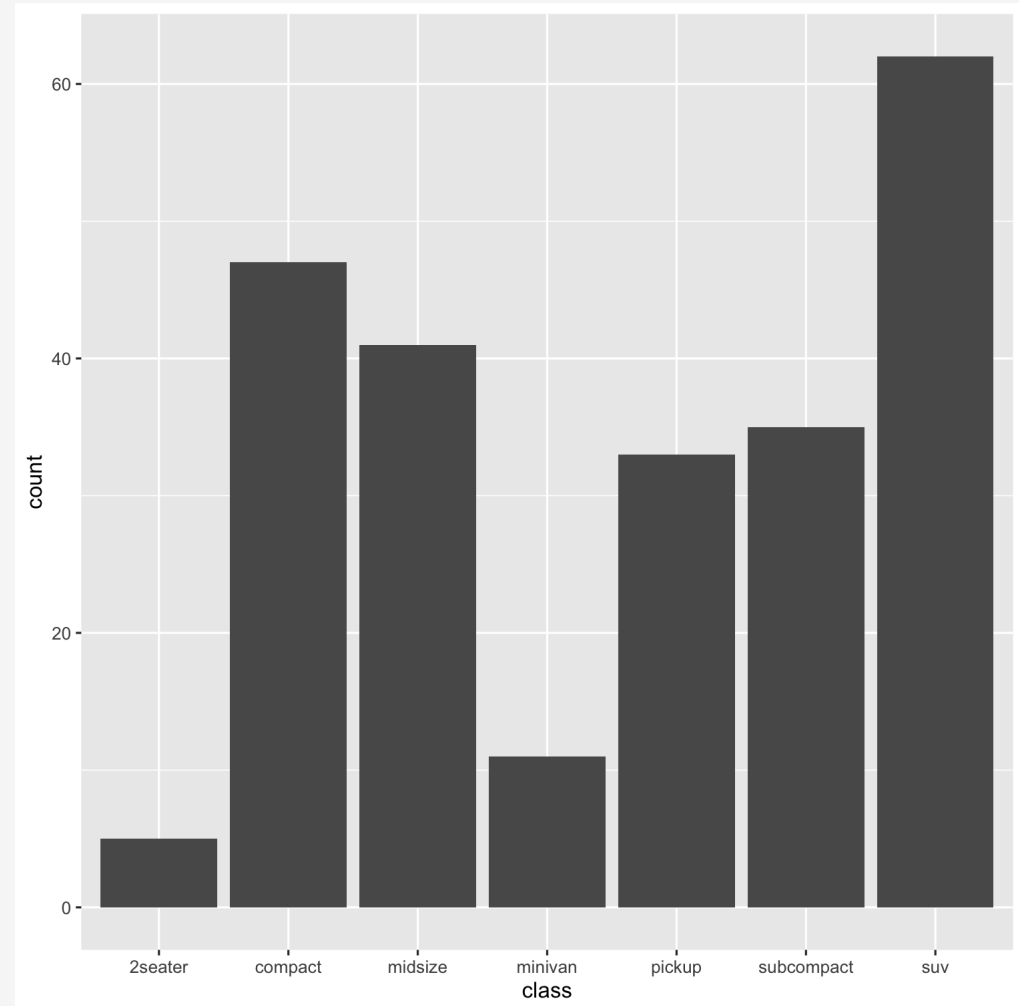
```
ggplot(data = mpg)+  
  aes(x = class,  
       y = hwy)+  
  geom_boxplot()
```



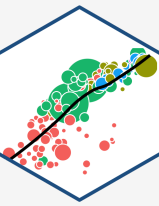
# Let's Change Our Plot



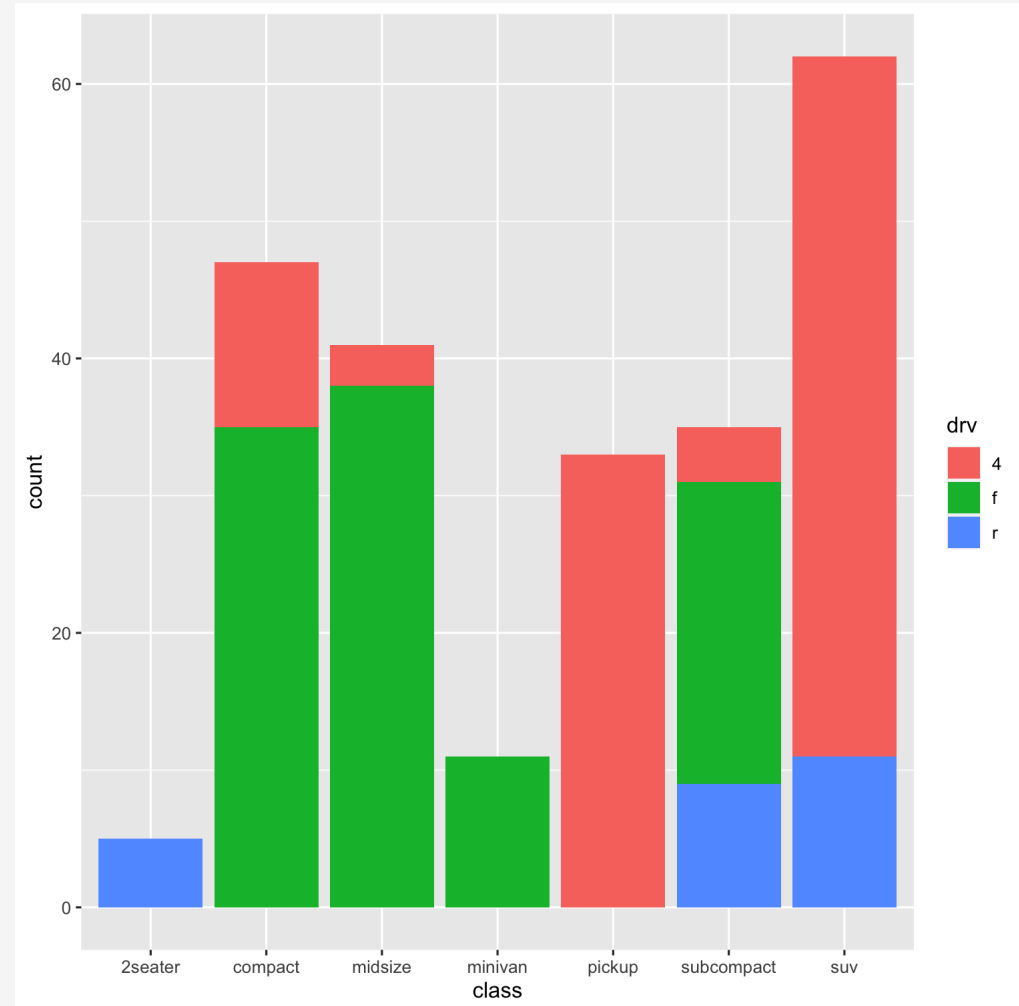
```
ggplot(data = mpg)+  
  aes(x = class)+  
  geom_bar()
```



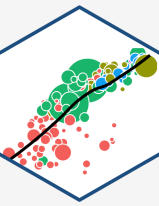
# Let's Change Our Plot



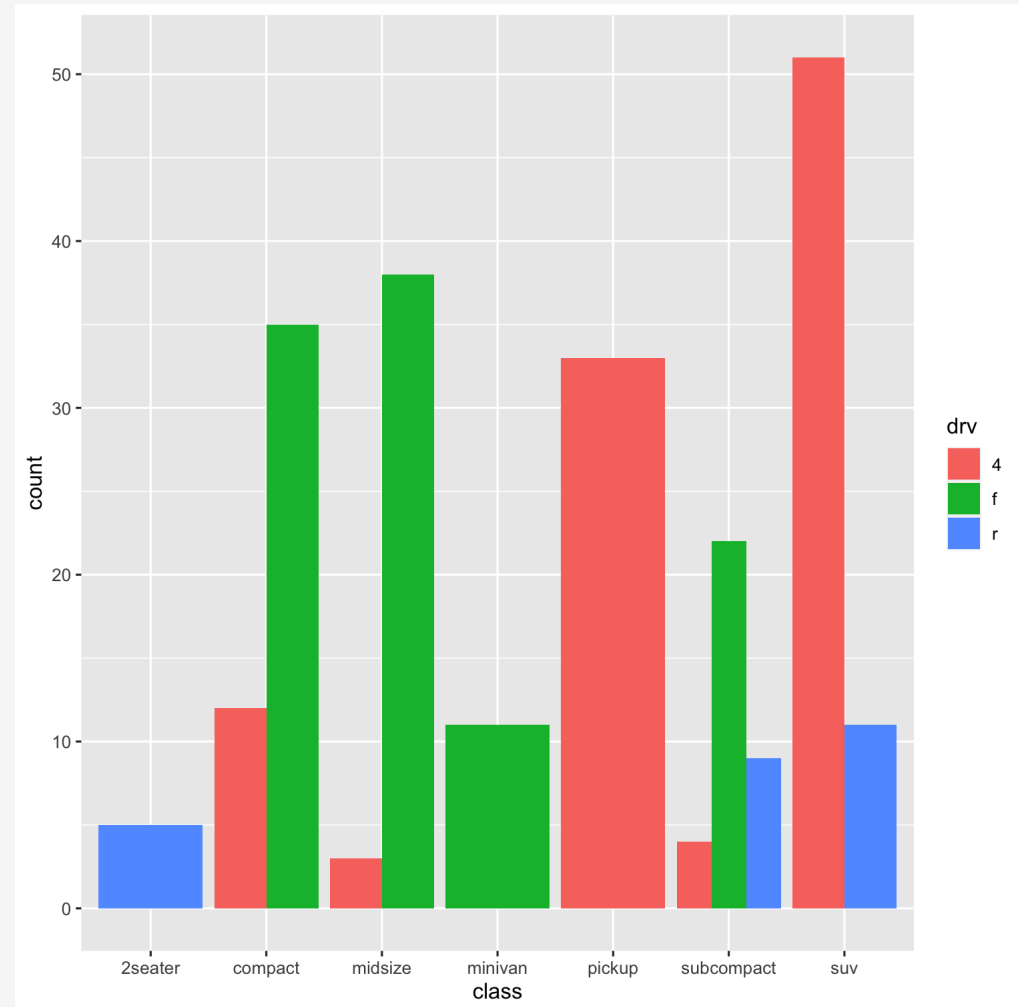
```
ggplot(data = mpg)+  
  aes(x = class,  
      fill = drv)+  
  geom_bar()
```



# Let's Change Our Plot

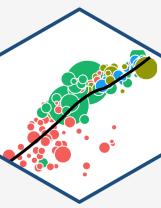


```
ggplot(data = mpg)+  
  aes(x = class,  
      fill = drv)+  
  geom_bar(position = "dodge")
```



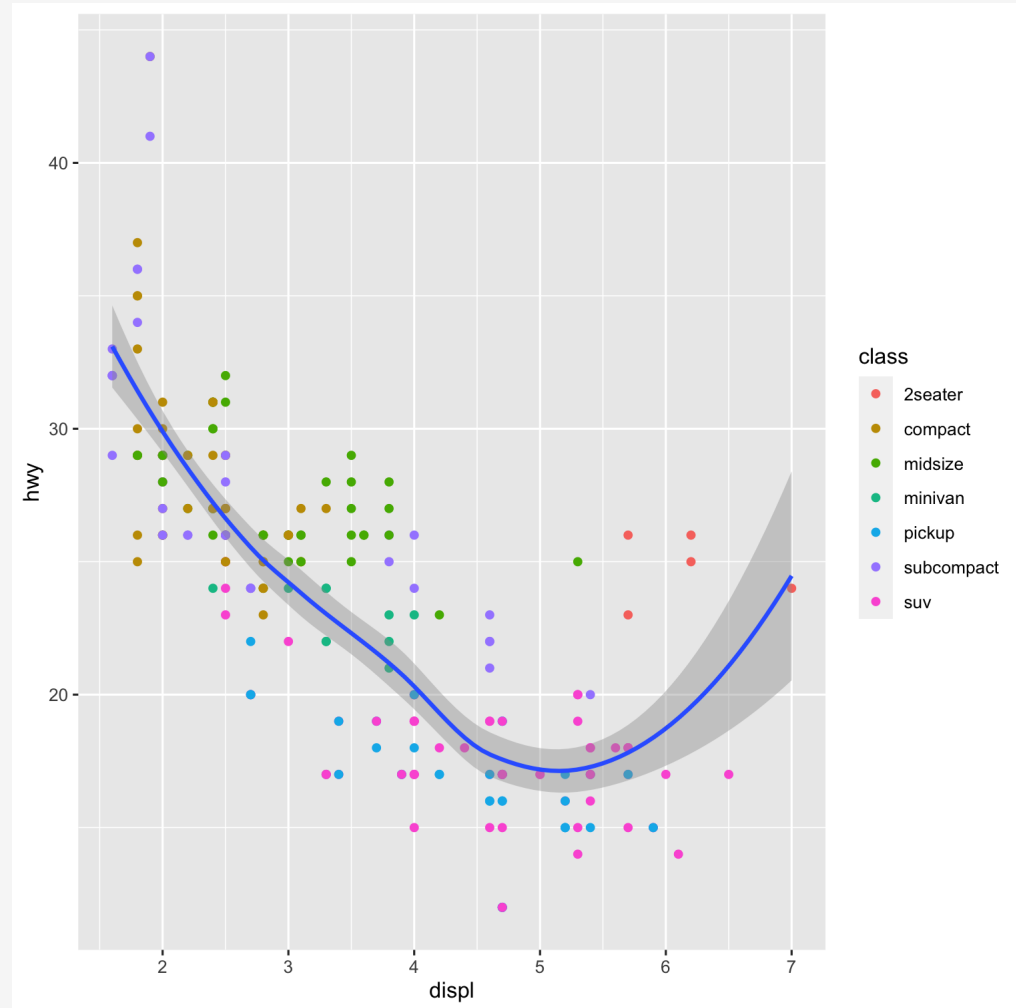


# Back to the Original (and saving it)

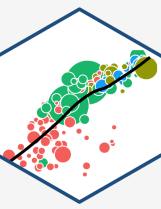


```
p <- ggplot(data = mpg)+  
  aes(x = displ,  
       y = hwy)+  
  geom_point(aes(color = class))+  
  geom_smooth()
```

```
p # show plot
```



# The Grammar of Graphics (gg): Facets I



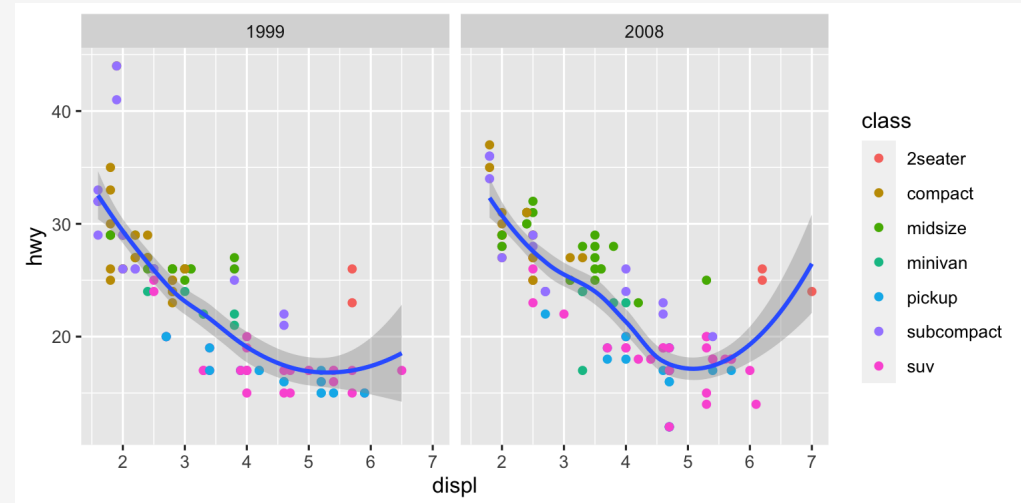
Data

Aesthetics

Geoms

Facets

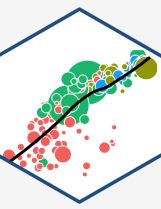
```
p + facet_wrap(~year)
```



```
+ facet_wrap()
```

```
+ facet_grid()
```

# The Grammar of Graphics (gg): Facets II



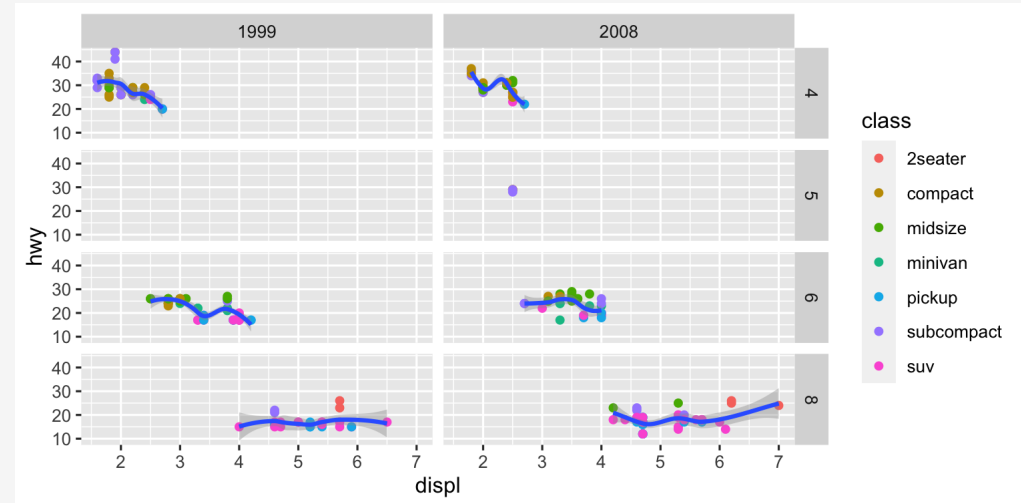
Data

Aesthetics

Geoms

Facets

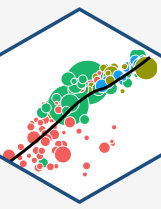
```
p + facet_grid(cyl~year)
```



```
+ facet_wrap()
```

```
+ facet_grid()
```

# The Grammar of Graphics (gg): Labels



Data

Aesthetics

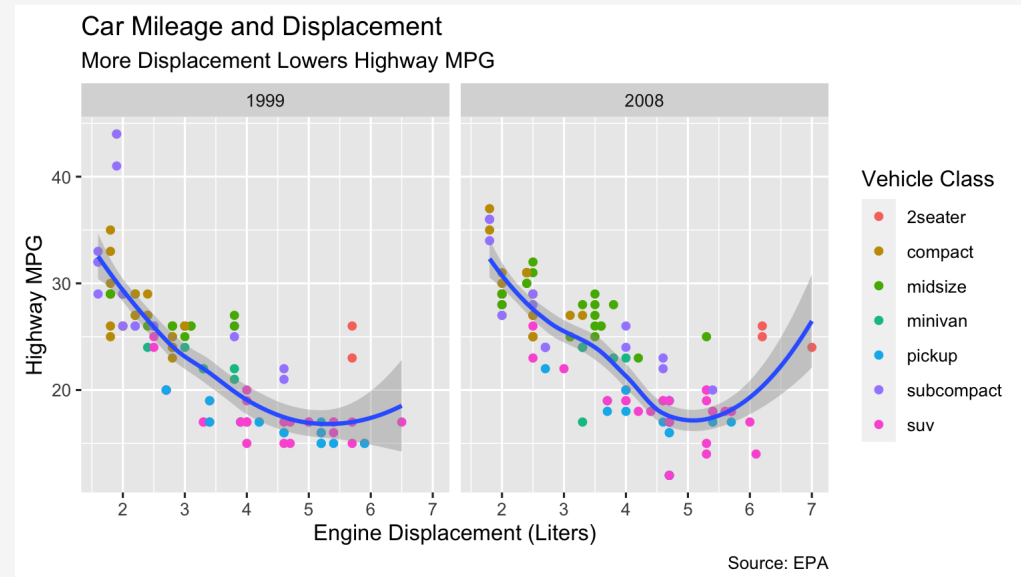
Geoms

Facets

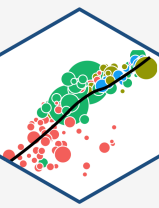
Labels

+ `labs()`

```
p + facet_wrap(~year)+  
  labs(x = "Engine Displacement (Liters)",  
       y = "Highway MPG",  
       title = "Car Mileage and Displacement",  
       subtitle = "More Displacement Lowers Highway MPG",  
       caption = "Source: EPA",  
       color = "Vehicle Class")
```



# The Grammar of Graphics (gg): Scales



Data

```
scale + _ + <aes> + _ + <type> + ()
```

Aesthetics

- `<aes>`: parameter you want to adjust

Geoms

- `<type>`: type of parameter

Facets

- I want to change my discrete x-axis:

```
scale_x_discrete()
```

Labels

- I want to change my continuous y-axis:

```
scale_y_continuous()
```

Scales

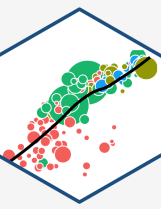
- I want to rescale x-axis to log: `scale_x_log10()`

- I want to use a different color palette:

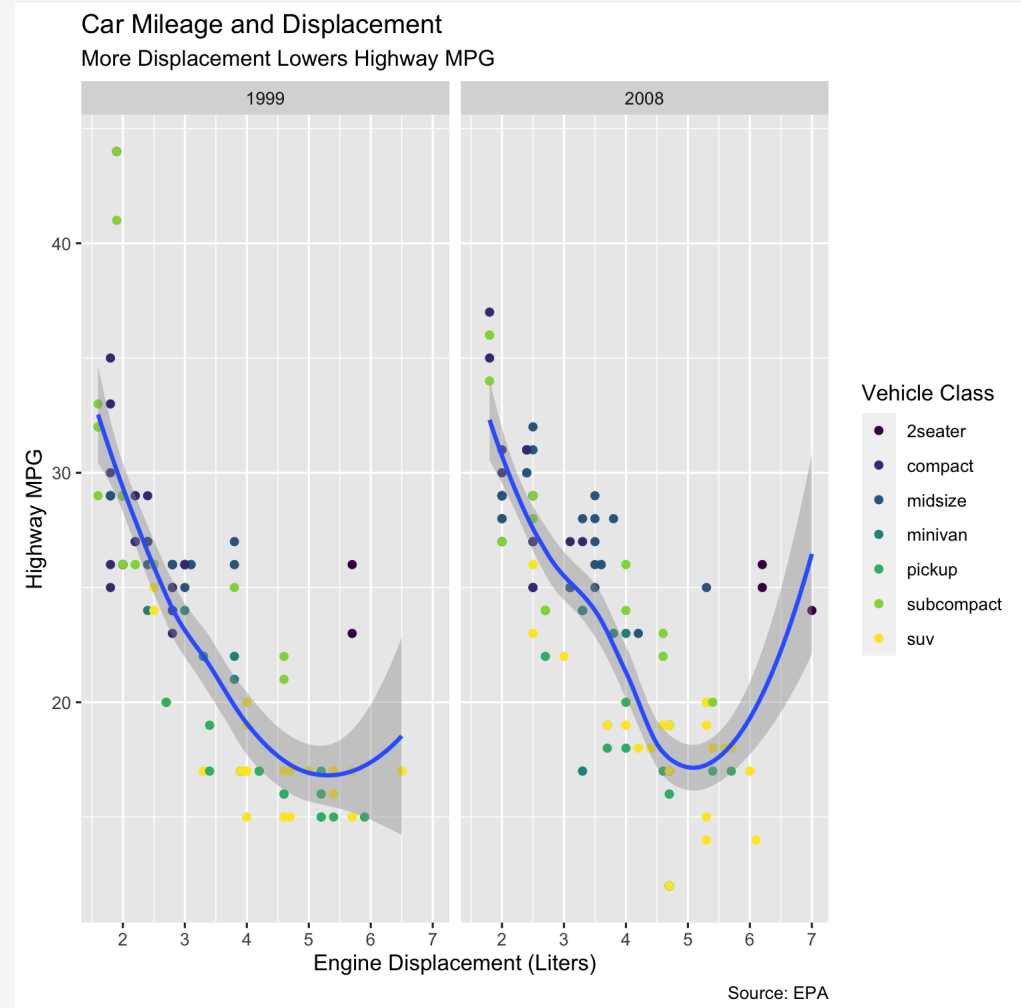
```
scale_fill_discrete(),
```

```
+ scale_*_*()
```

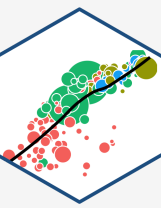
# The Grammar of Graphics (gg): Scales



```
ggplot(data = mpg)+  
  aes(x = displ,  
      y = hwy)+  
  geom_point(aes(color = class))+  
  geom_smooth()+  
  facet_wrap(~year)+  
  labs(x = "Engine Displacement (Liter",  
       y = "Highway MPG",  
       title = "Car Mileage and Displa",  
       subtitle = "More Displacement L",  
       caption = "Source: EPA",  
       color = "Vehicle Class")+  
  scale_color_viridis_d()
```



# The Grammar of Graphics (gg): Themes



Data

**Theme** changes appearance of plot decorations (things not mapped to data)

Aesthetics

- Some themes that come with `ggplot2`:

Geoms

- + `theme_bw()`

Facets

- + `theme_dark()`

Labels

- + `theme_gray()`

Scales

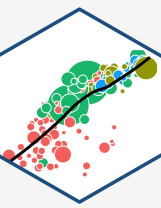
- + `theme_minimal()`

Theme

- + `theme_light()`

- + `theme_classic()`

# The Grammar of Graphics (gg): Themes



Data

**Theme** changes appearance of plot decorations (things not mapped to data)

Aesthetics

- Many parameters we could change

Geoms

- Global options: `line`, `rect`, `text`, `title`

Facets

- `axis`: x-, y-, or other axis title, ticks, lines

Labels

- `legend`: plot legends for fill or color

Scales

- `panel`: actual plot area

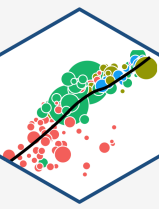
Theme

- `plot`: whole image

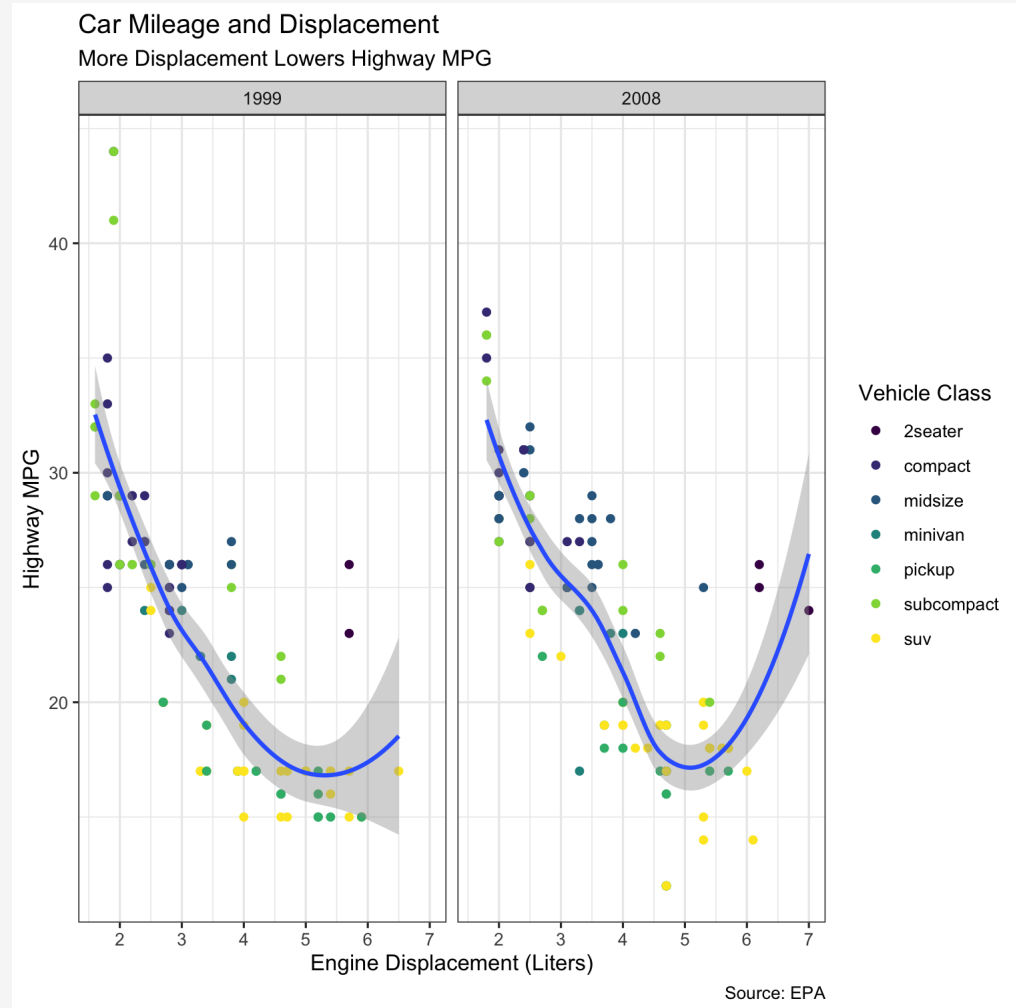
- `strip`: facet labels



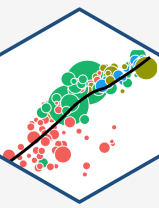
# The Grammar of Graphics (gg): Themes



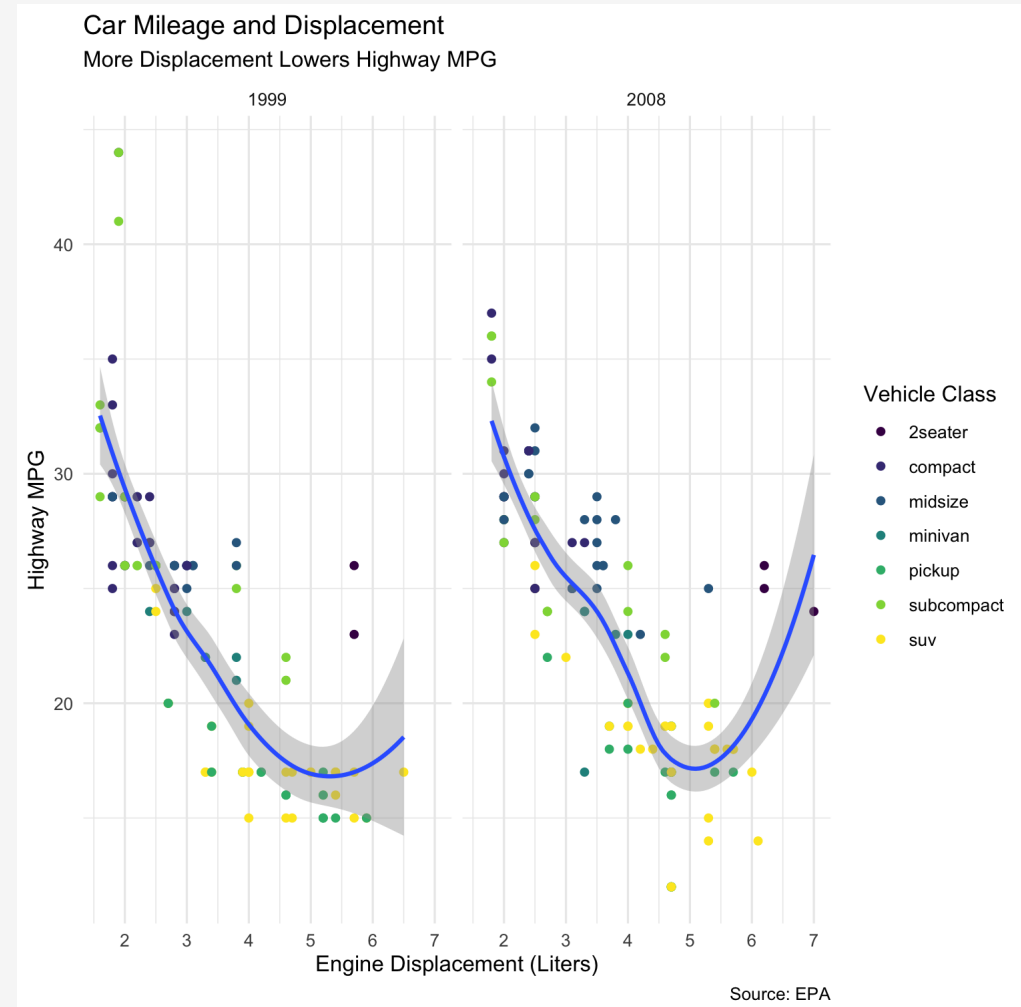
```
ggplot(data = mpg)+  
  aes(x = displ,  
      y = hwy)+  
  geom_point(aes(color = class))+  
  geom_smooth()+  
  facet_wrap(~year)+  
  labs(x = "Engine Displacement (Liter",  
       y = "Highway MPG",  
       title = "Car Mileage and Displa",  
       subtitle = "More Displacement L",  
       caption = "Source: EPA",  
       color = "Vehicle Class")+  
  scale_color_viridis_d()+  
  theme_bw()
```



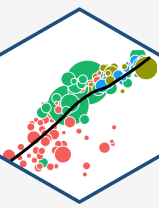
# The Grammar of Graphics (gg): Themes II



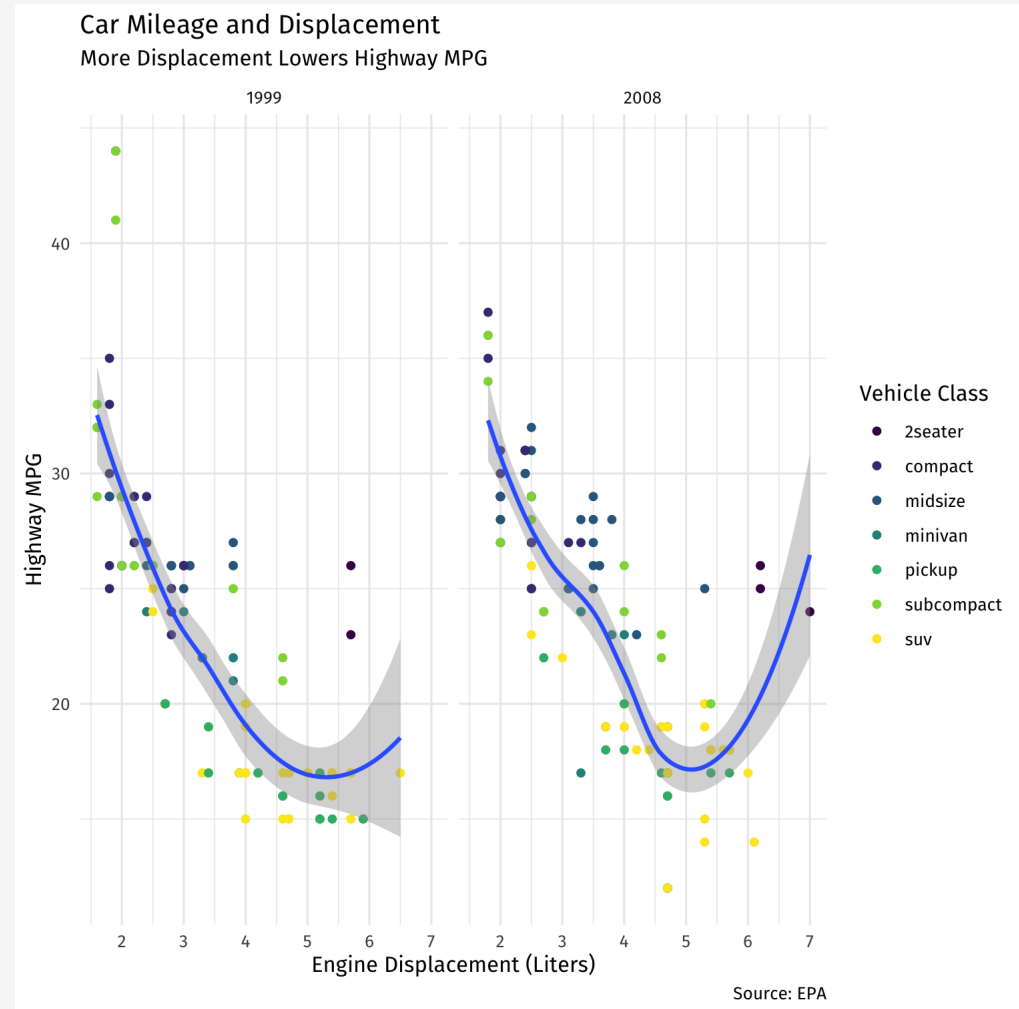
```
ggplot(data = mpg)+  
  aes(x = displ,  
      y = hwy)+  
  geom_point(aes(color = class))+  
  geom_smooth()+  
  facet_wrap(~year)+  
  labs(x = "Engine Displacement (Liter",  
       y = "Highway MPG",  
       title = "Car Mileage and Displa",  
       subtitle = "More Displacement L",  
       caption = "Source: EPA",  
       color = "Vehicle Class")+  
  scale_color_viridis_d()+  
  theme_minimal()
```



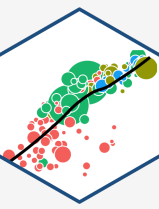
# The Grammar of Graphics (gg): Themes III



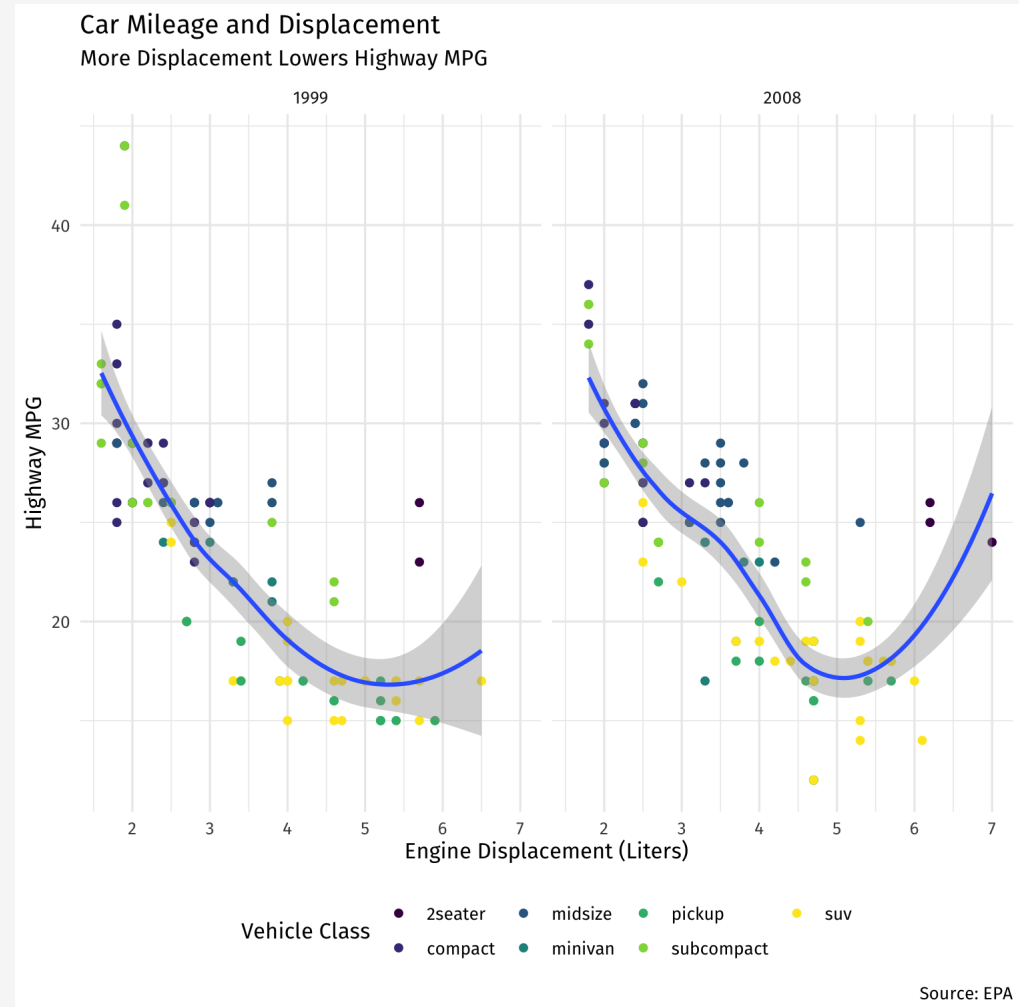
```
ggplot(data = mpg)+  
  aes(x = displ,  
      y = hwy)+  
  geom_point(aes(color = class))+  
  geom_smooth()+  
  facet_wrap(~year)+  
  labs(x = "Engine Displacement (Liter  
      y = "Highway MPG",  
      title = "Car Mileage and Displa  
      subtitle = "More Displacement L  
      caption = "Source: EPA",  
      color = "Vehicle Class")+  
  scale_color_viridis_d()+  
  theme_minimal()+  
  theme(text = element_text(family = "
```



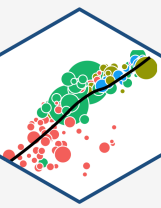
# The Grammar of Graphics (gg): Themes III



```
ggplot(data = mpg)+  
  aes(x = displ,  
      y = hwy)+  
  geom_point(aes(color = class))+  
  geom_smooth()+  
  facet_wrap(~year)+  
  labs(x = "Engine Displacement (Liter  
      y = "Highway MPG",  
      title = "Car Mileage and Displa  
      subtitle = "More Displacement L  
      caption = "Source: EPA",  
      color = "Vehicle Class")+  
  scale_color_viridis_d()+  
  theme_minimal()+  
  theme(text = element_text(family = "  
      legend.position="bottom")
```



# The Grammar of Graphics (gg): Themes (ggthemes)



Data

- `ggthemes` package adds some other nice themes

Aesthetics

```
# install if you don't have it  
# install.packages("ggthemes")  
library("ggthemes") # load package
```

Geoms

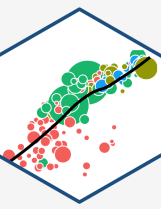
Facets

Labels

Scales

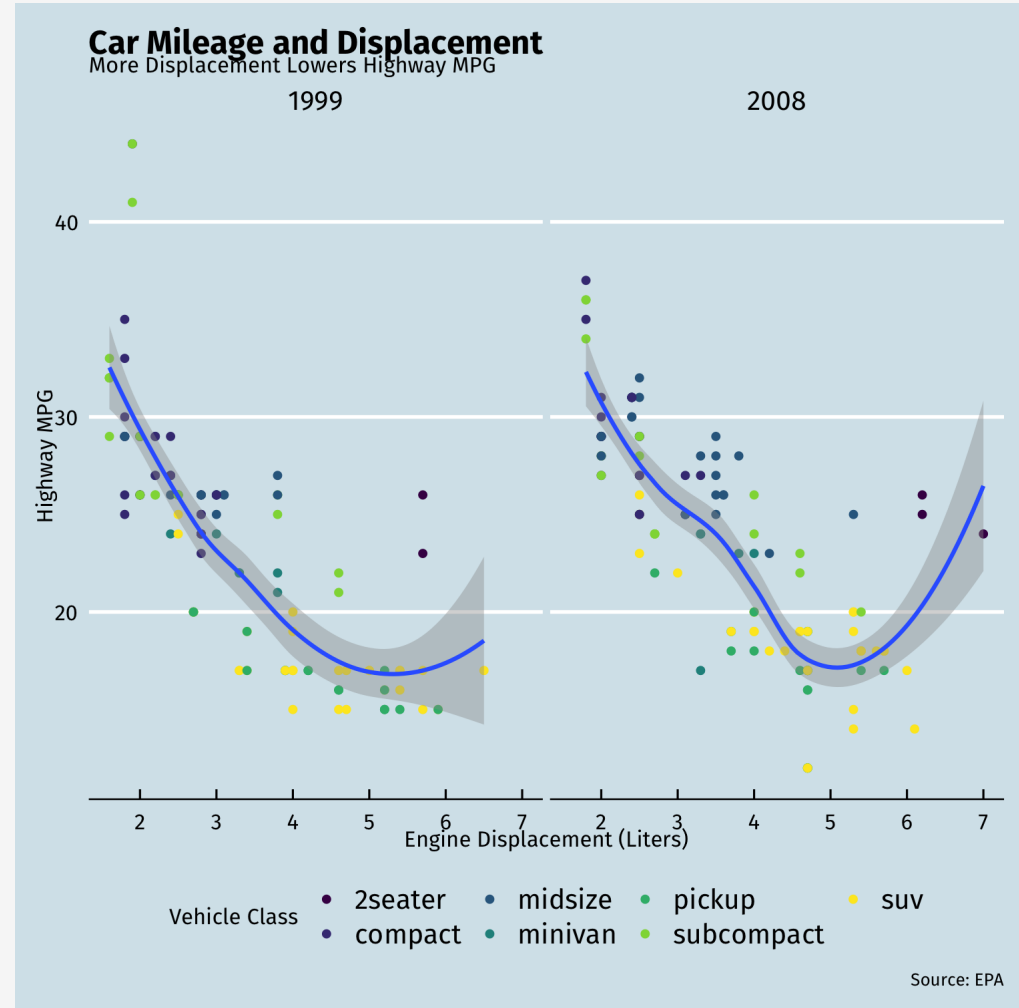
Theme

# The Grammar of Graphics (gg): Themes IV

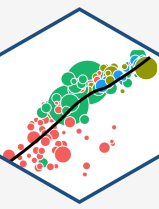


```
library("ggthemes")
```

```
ggplot(data = mpg)+  
  aes(x = displ,  
      y = hwy)+  
  geom_point(aes(color = class))+  
  geom_smooth()+  
  facet_wrap(~year)+  
  labs(x = "Engine Displacement (Liter",  
       y = "Highway MPG",  
       title = "Car Mileage and Displa",  
       subtitle = "More Displacement L",  
       caption = "Source: EPA",  
       color = "Vehicle Class")+  
  scale_color_viridis_d()+  
  theme_economist()+  
  theme(text = element_text(family = "  
      legend.position="bottom")
```

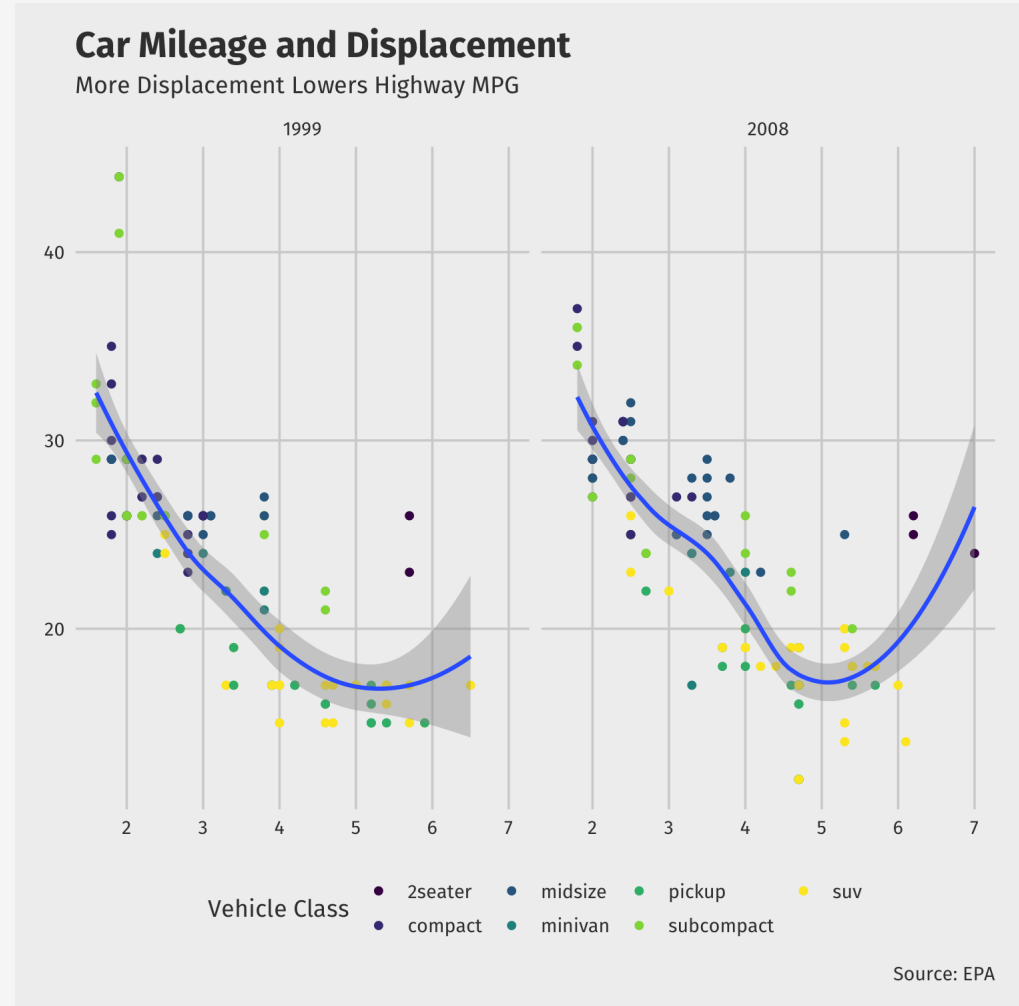


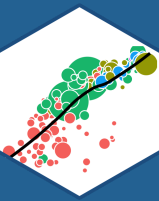
# The Grammar of Graphics (gg): Themes V



```
library("ggthemes")
```

```
ggplot(data = mpg)+  
  aes(x = displ,  
       y = hwy)+  
  geom_point(aes(color = class))+  
  geom_smooth()+  
  facet_wrap(~year)+  
  labs(x = "Engine Displacement (Liter",  
       y = "Highway MPG",  
       title = "Car Mileage and Displa",  
       subtitle = "More Displacement L",  
       caption = "Source: EPA",  
       color = "Vehicle Class")+  
  scale_color_viridis_d()+  
  theme_fivethirtyeight()+  
  theme(text = element_text(family = "  
      legend.position="bottom")
```

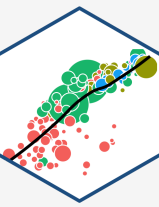




# Some Troubleshooting



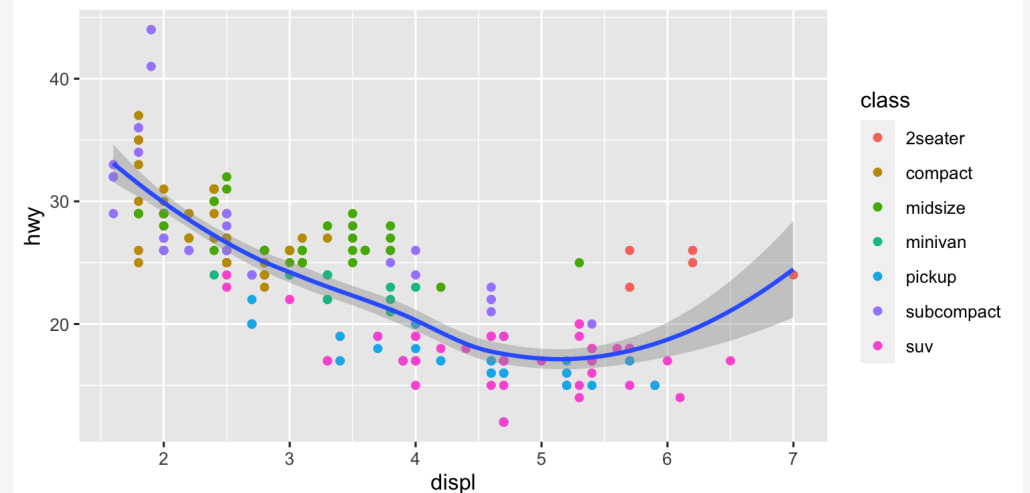
# Global vs. Local Aesthetics



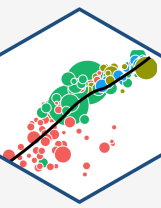
- `aes()` can go in base (`data`) layer and/or in individual `geom()` layers
- All `geoms` will inherit global `aes` from `data` layer unless overridden

```
# ALL GEOMS will map data to colors
ggplot(data = mpg, aes(x = displ,
                       y = hwy,
                       color = class))+
  geom_point()+
  geom_smooth()
```

```
# ONLY points will map data to colors
ggplot(data = mpg, aes(x = displ,
                       y = hwy))+
  geom_point(aes(color = class))+
  geom_smooth()
```



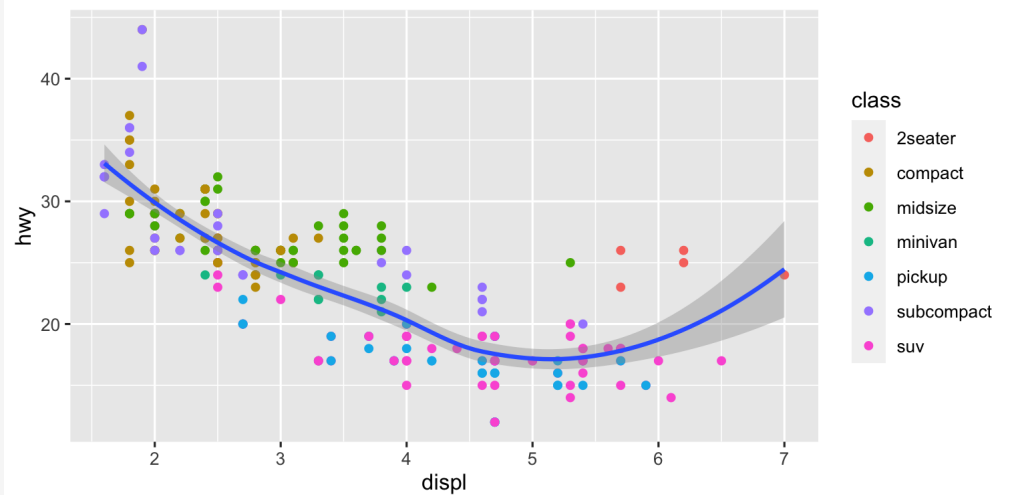
# Mapped vs. Set Aesthetics



- `aes` aesthetics such as `size` and `color` can be mapped from data or set to a single value
- Map *inside* of `aes()`, set *outside* of `aes()`

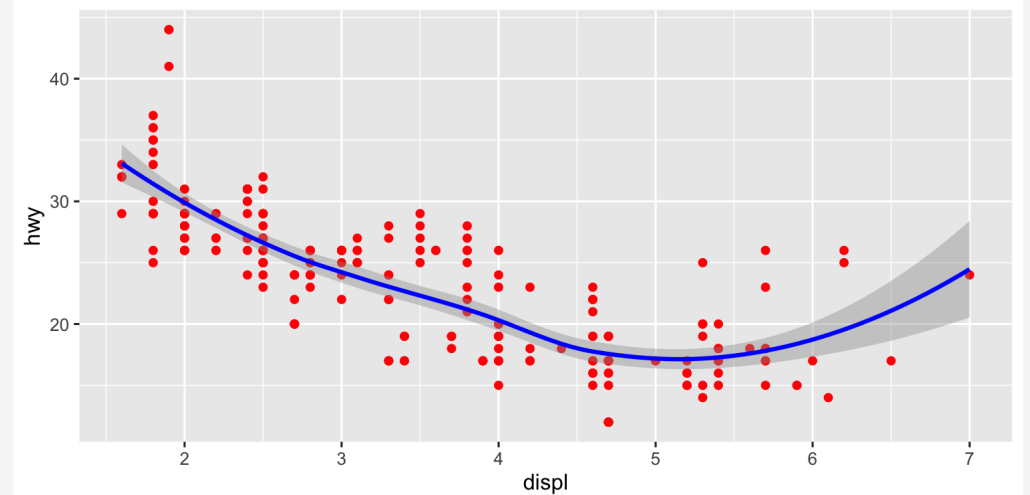
```
# Point colors are mapped from class data
```

```
ggplot(data = mpg, aes(x = displ,  
                       y = hwy))+  
  geom_point(aes(color = class))+  
  geom_smooth()
```

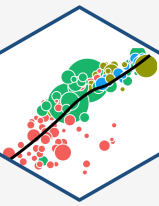


```
# Point colors are all set to blue
```

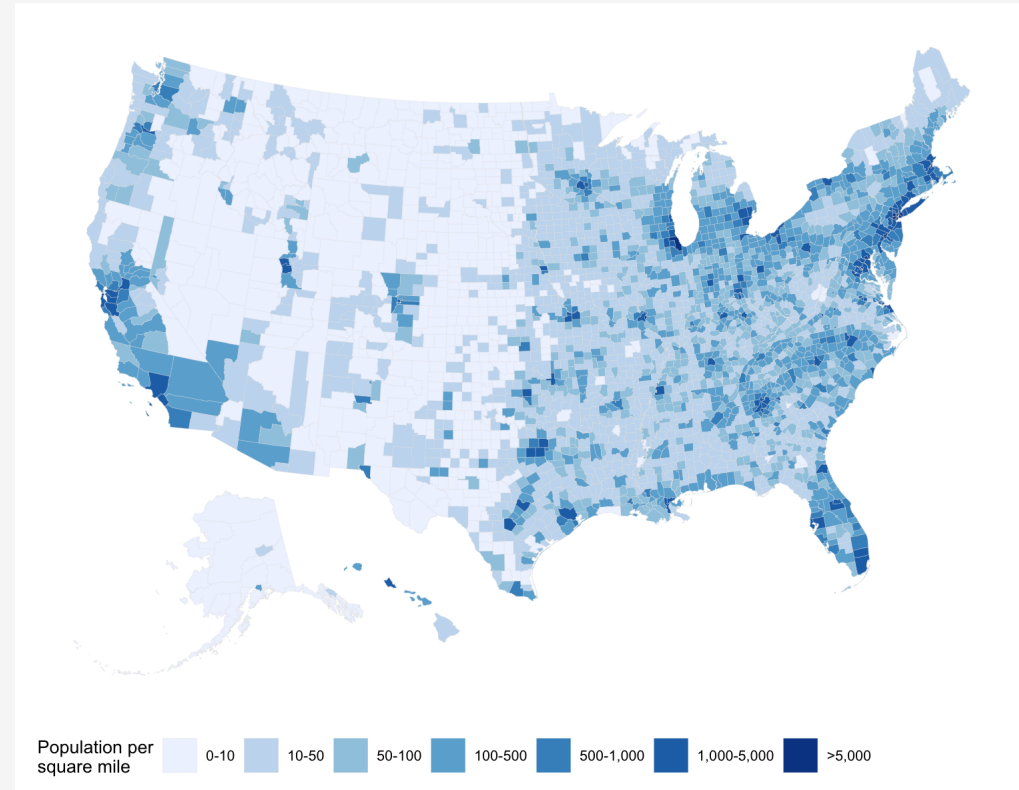
```
ggplot(data = mpg, aes(x = displ,  
                       y = hwy))+  
  geom_point(aes(), color = "red")+  
  geom_smooth(aes(), color = "blue")
```



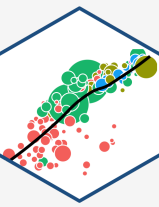
# Go Crazy I



```
# I did some (hidden) data work before
ggplot(data = county_full,
       mapping = aes(x = long, y = lat,
                    fill = pop_dens,
                    group = group)) +
  geom_polygon(color = "gray90", size = 0.5) +
  coord_equal() +
  scale_fill_brewer(palette = "Blues",
                   labels = "Population per square mile") +
  labs(fill = "Population per square mile") +
  theme_map() +
  guides(fill = guide_legend(nrow = 1,
                             title = "Population per square mile",
                             theme(legend.position = "bottom")))
```

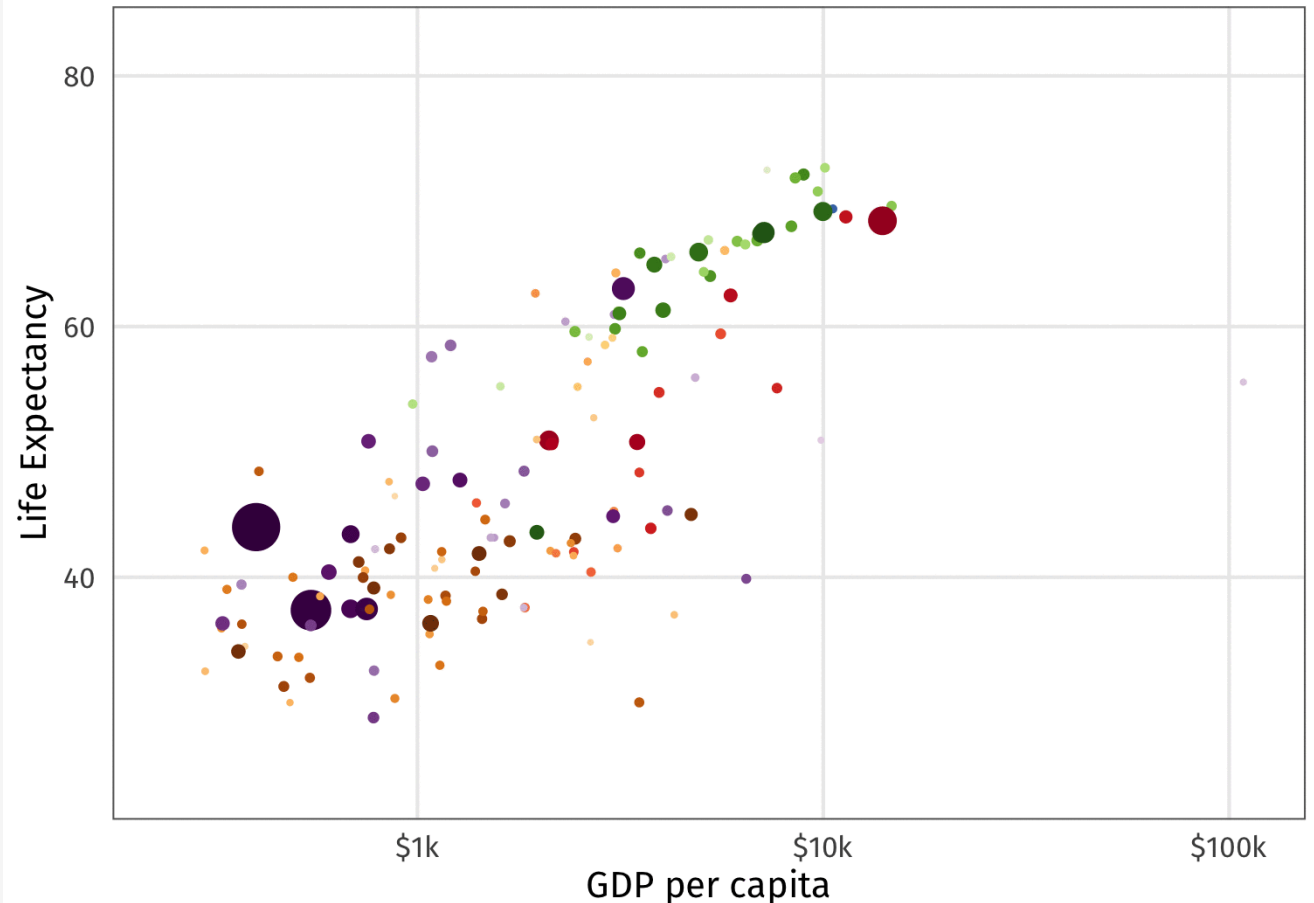


# Go Crazy II



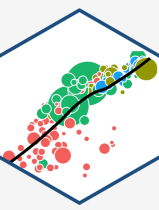
```
library("gapminder")
library("gganimate")
ggplot(gapminder) +
  aes(x = gdpPerCap, y = lifeExp, size
  geom_point() +
  guides(color = FALSE, size = FALSE)
  scale_x_log10(
    breaks = c(10^3, 10^4, 10^5),
    labels = c("$1k", "$10k", "$100k")
  scale_color_manual(values = gapminder
  scale_size(range = c(0.5, 12)) +
  labs(
    x = "GDP per capita",
    y = "Life Expectancy",
    caption = "Source: Hans Rosling's
  theme_minimal(14, base_family = "Fir
  theme(
    strip.text = element_text(size = 1
    panel.border = element_rect(fill =
```

Income and Life Expectancy - 1952



Source: Hans Rosling's [gapminder.org](http://gapminder.org)

# Data Visualization and Graphic Design Principles

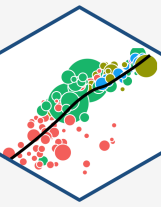


- We will return to various graphics as we cover descriptive statistics and regression
- I hope to cover some basic principles of good graphic design for figures and plots
  - If not in class, I will make a page on the website, and/or a video

Remember:



# Less is More



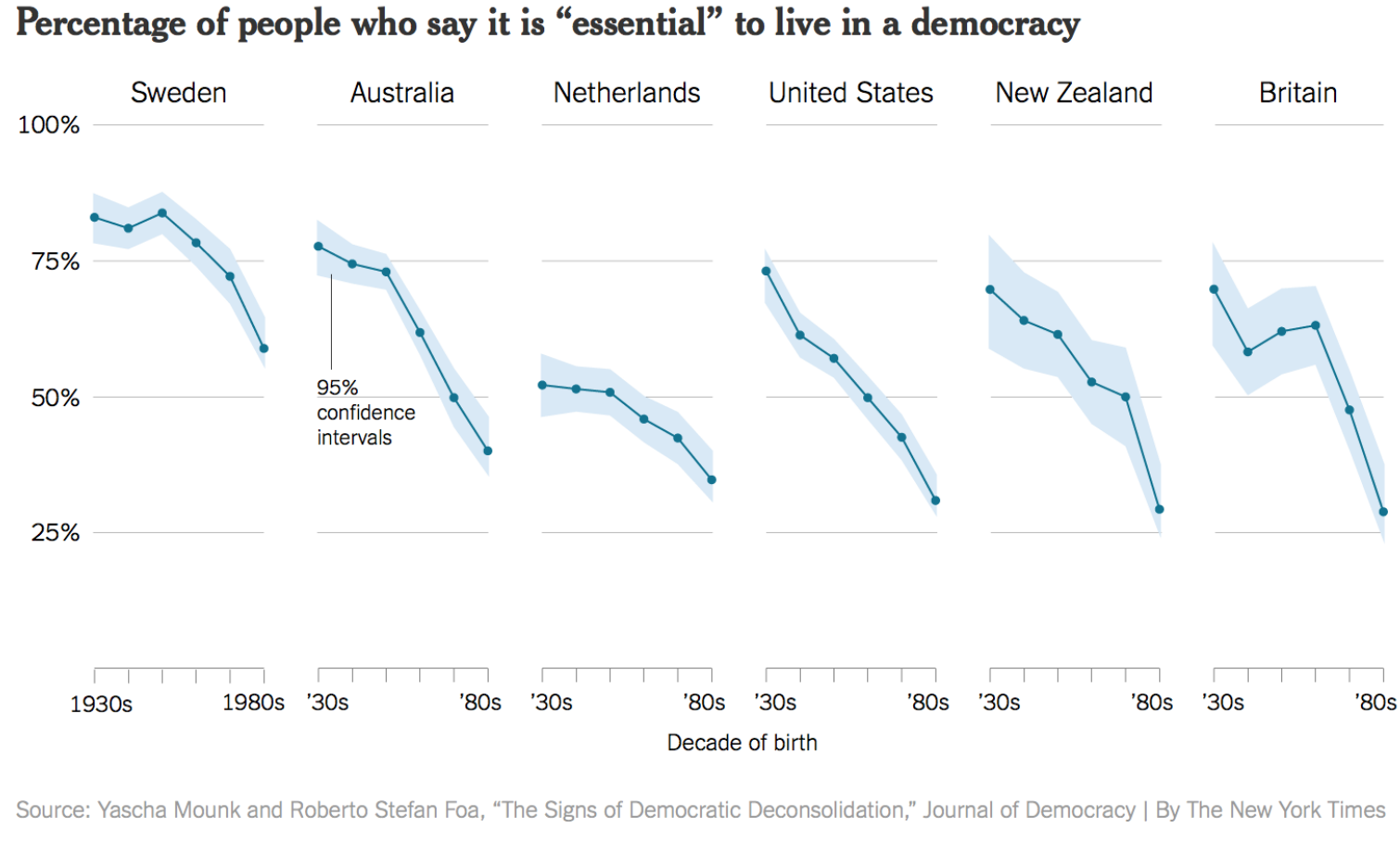
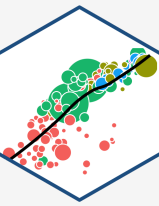
"Shoot me"



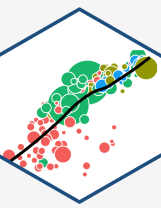
Less is More:

**Remove**  
to improve  
(the **data-ink** ratio)

# Try to Show One Trend Really Clearly



# Reference: R Studio Makes Great "Cheat Sheet"s!



## Data Visualization with ggplot2 :: CHEAT SHEET



### Basics

ggplot2 is based on the **grammar of graphics**, the idea that you can build every graph from the same components: a **data set**, a **coordinate system**, and **geoms**—visual marks that represent data points.



To display values, map variables in the data to visual properties of the geom (**aesthetics**) like **size**, **color**, and **x** and **y** locations.



Complete the template below to build a graph.

```
ggplot(data = <DATA>) +
  <GEOM_FUNCTION>(mapping = aes(<MAPPINGS>),
  stat = <STAT>, position = <POSITION>) +
  <COORDINATE_FUNCTION> +
  <FACET_FUNCTION> +
  <SCALE_FUNCTION> +
  <THEME_FUNCTION>
```

required  
Not required, sensible defaults supplied

ggplot(data = mpg, aes(x = cty, y = hwy)) Begins a plot that you finish by adding layers. Add one geom function per layer.

geom\_function(aesthetic\_mappings, data, geom) Creates a complete plot with given data, geom, and mappings. Supplies many useful defaults.

last\_plot() Returns the last plot

ggsave("plot.png", width = 5, height = 5) Saves last plot as 5 x 5 file named "plot.png" in working directory. Matches file type to file extension.

### Geoms

Use a geom function to represent data points, use the geom's aesthetic properties to represent variables. Each function returns a layer.

#### GRAPHICAL PRIMITIVES

```
a <- ggplot(economics, aes(date, unemploy))
b <- ggplot(seals, aes(x = long, y = lat))

a + geom_blank()
  (Useful for expanding limits)

b + geom_curve(aes(yend = lat + 1,
  xend = long + 1, curvature = z))
  x, yend, y, vend,
  alpha, angle, color, curvature, linetype, size

a + geom_path(linetype = "butt", linejoin = "round",
  linemitre = 1)
  x, y, alpha, color, group, linetype, size

a + geom_polygon(aes(group = group))
  x, y, alpha, color, fill, group, linetype, size

b + geom_rect(aes(xmin = long, ymin = lat, xmax =
  long + 1, ymax = lat + 1))
  - xmax, xmin, ymax,
  ymin, alpha, color, fill, linetype, size

a + geom_ribbon(aes(ymin = unemploy - 900,
  ymax = unemploy + 900))
  - x, ymax, ymin,
  alpha, color, fill, group, linetype, size
```

#### LINE SEGMENTS

```
common aesthetics: x, y, alpha, color, linetype, size

b + geom_abline(aes(intercept = 0, slope = 1))
b + geom_hline(aes(yintercept = lat))
b + geom_vline(aes(xintercept = long))

b + geom_segment(aes(yend = lat + 1, xend = long + 1))
b + geom_spoke(aes(angle = 1.1155, radius = 1))
```

#### ONE VARIABLE continuous

```
c <- ggplot(mpg, aes(hwy)); c2 <- ggplot(mpg)

c + geom_area(stat = "bin")
  x, y, alpha, color, fill, linetype, size

c + geom_density(kernel = "gaussian")
  x, y, alpha, color, fill, group, linetype, size, weight

c + geom_dotplot()
  x, y, alpha, color, fill

c + geom_freqpoly()
  x, y, alpha, color, group, linetype, size

c + geom_histogram(binwidth = 5)
  x, y, alpha, color, fill, linetype, size, weight

c2 + geom_qq(aes(sample = hwy))
  x, y, alpha, color, fill, linetype, size, weight
```

```
discrete
d <- ggplot(mpg, aes(fill))

d + geom_bar()
  x, alpha, color, fill, linetype, size, weight
```

#### TWO VARIABLES

```
continuous x, continuous y
e <- ggplot(mpg, aes(cty, hwy))

e + geom_label(aes(label = cty), nudge_x = 1,
  nudge_y = 1, check_overlap = TRUE)
  x, y, label,
  alpha, angle, color, family, fontface, hjust,
  linetype, size, vjust

e + geom_jitter(height = 2, width = 2)
  x, y, alpha, color, fill, shape, size

e + geom_point()
  x, y, alpha, color, fill, shape, size, stroke

e + geom_quantile()
  x, y, alpha, color, group, linetype, size, weight

e + geom_rug(sides = "bl")
  x, y, alpha, color, linetype, size

e + geom_smooth(method = lm)
  x, y, alpha, color, fill, group, linetype, size, weight

e + geom_text(aes(label = cty), nudge_x = 1,
  nudge_y = 1, check_overlap = TRUE)
  x, y, label,
  alpha, angle, color, family, fontface, hjust,
  linetype, size, vjust
```

#### discrete x, continuous y

```
f <- ggplot(mpg, aes(class, hwy))

f + geom_col()
  x, y, alpha, color, fill, group, linetype, size

f + geom_boxplot()
  x, y, lower, middle, upper,
  ymax, ymin, alpha, color, fill, group, linetype,
  shape, size, weight

f + geom_dotplot(binaxis = "y", stackdir =
  "center")
  x, y, alpha, color, fill, group

f + geom_violin(scale = "area")
  x, y, alpha, color, fill, group, linetype, size, weight
```

#### discrete x, discrete y

```
g <- ggplot(diamonds, aes(carat, cut))

g + geom_count()
  x, y, alpha, color, fill, shape, size, stroke

g + geom_contour(aes(z = z))
  x, y, z, alpha, color, group, linetype, size, weight
```

#### THREE VARIABLES

```
sealsSz <- with(seals, sqrt(delta_long^2 + delta_lat^2))
h <- ggplot(seals, aes(long, lat))

h + geom_raster(aes(fill = z), hjust = 0.5, vjust = 0.5,
  interpolate = FALSE)
  x, y, alpha, fill

h + geom_tile(aes(fill = z), x, y, alpha, color, fill,
  linetype, size, width
```

#### continuous bivariate distribution

```
h <- ggplot(diamonds, aes(carat, price))

h + geom_bin2d(binwidth = c(0.25, 500))
  x, y, alpha, color, fill, linetype, size, weight

h + geom_density2d()
  x, y, alpha, color, group, linetype, size

h + geom_hex()
  x, y, alpha, color, fill, size
```

#### continuous function

```
i <- ggplot(economics, aes(date, unemploy))

i + geom_area()
  x, y, alpha, color, fill, linetype, size

i + geom_line()
  x, y, alpha, color, group, linetype, size

i + geom_step(direction = "hv")
  x, y, alpha, color, group, linetype, size
```

#### visualizing error

```
df <- data.frame(grp = c("A", "B"), fit = 4.5, se = 1.2)
j <- ggplot(df, aes(grp, fit, ymin = fit - se, ymax = fit + se))

j + geom_crossbar(latten = 2)
  x, y, ymax, ymin, alpha, color, fill, group, linetype,
  size

j + geom_errorbar()
  x, ymax, ymin, alpha, color, group, linetype, size, width (also
  geom_errorbarh())

j + geom_lineangle()
  x, ymin, ymax, alpha, color, group, linetype, size

j + geom_pointrange()
  x, y, ymin, ymax, alpha, color, fill, group, linetype,
  shape, size
```

#### maps

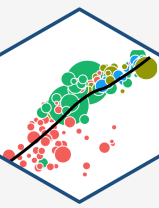
```
data <- data.frame(murder = USArrests$Murder,
  state = tolower(rownames(USArrests)))
map <- map_data("state")
k <- ggplot(data, aes(fill = murder))

k + geom_map(aes(map_id = state), map = map)
  + expand_limits(x = map$long, y = map$lat),
  map_id, alpha, color, fill, linetype, size
```





# Reference



On `ggplot2`

- **R Studio's [ggplot2 Cheat Sheet](#)**
- **`ggplot2`'s website [reference section](#)**
- Hadley Wickham's [R for Data Science book chapter on ggplot2](#)
- STHDA's [be awesome in ggplot2](#)
- r-statistic's [top 50 ggplot2 visualizations](#)

On data visualization

- **Kieran Healy's [Data Visualization: A Practical Guide](#)**
- **Claus Wilke's [Fundamentals of Data Visualization](#)**
- PolicyViz [Better Presentations](#)
- Karl Broman's [How to Display Data Badly](#)