

# Data Visualisation

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# Graphics and Statistics

# Our Data Source

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

```
1 library(ggplot2)
2
3 head(mpg)
```

# A tibble: 6 × 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...

ggplot2 and the tidyverse

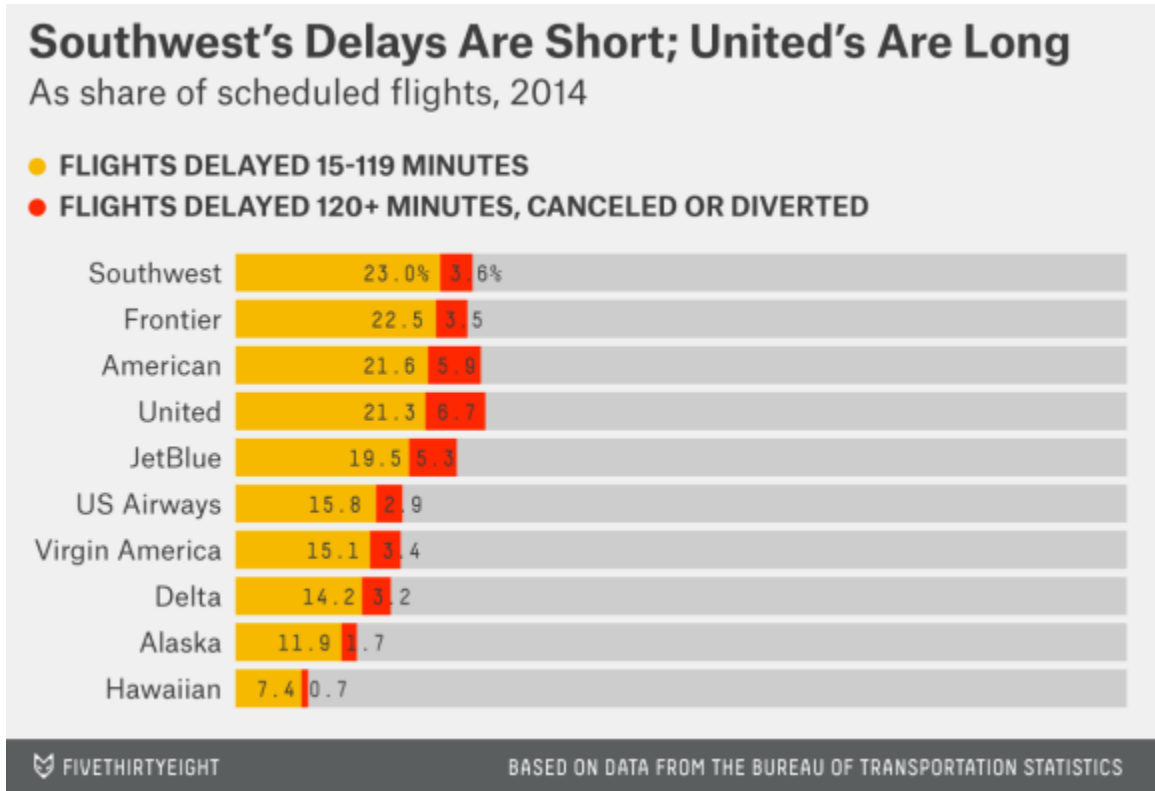
tidyverse

# ggplot2

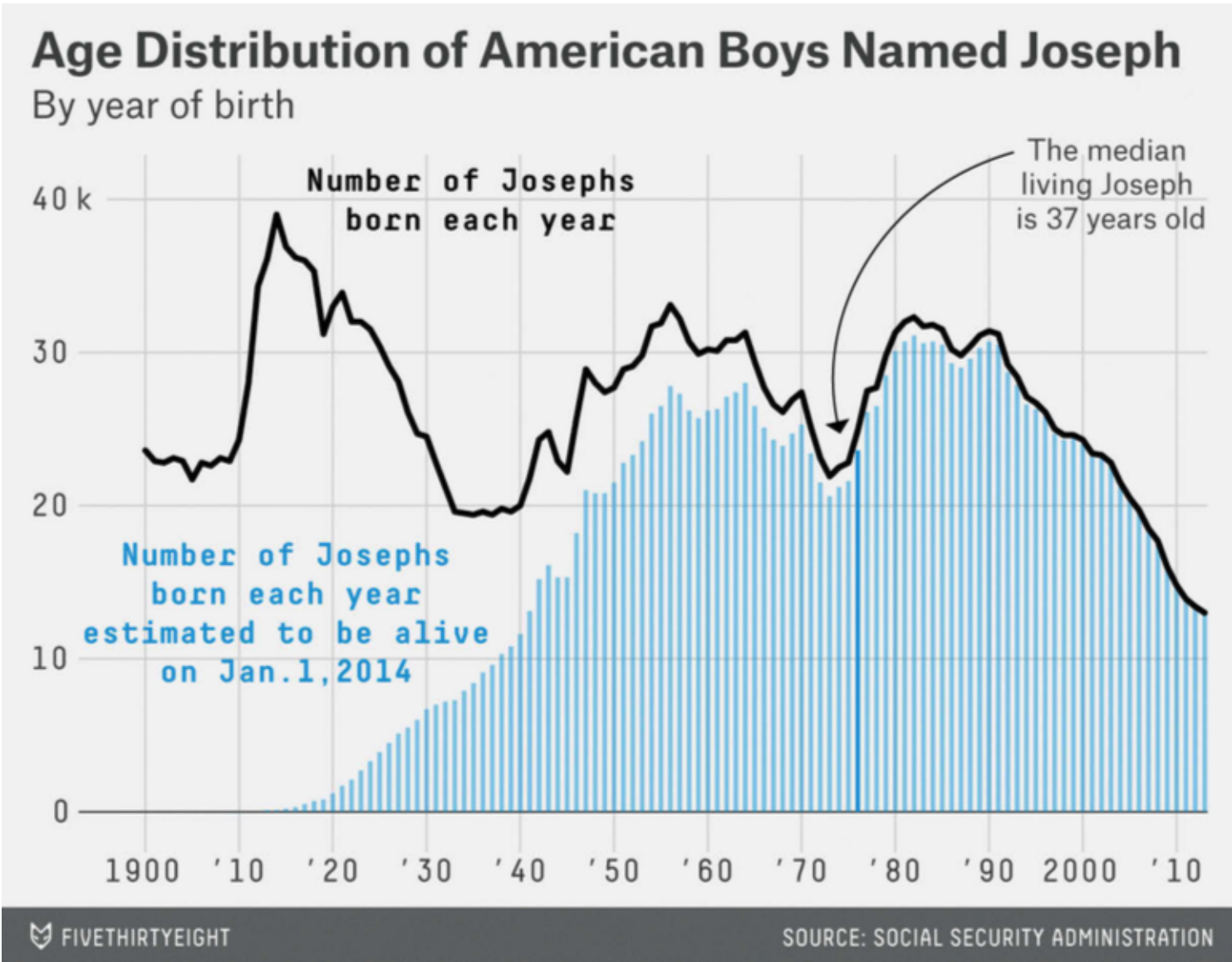
- `ggplot2` is perhaps the most popular package in `R` and a core element of the `tidyverse`
- `gg` stands for a **grammar of graphics**
- Very powerful and beautiful graphics, very customizable and reproducible, but requires a bit of a learning curve
- All those “cool graphics” you’ve seen in the New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



# ggplot: All Your Figure are Belong to Us



Source: [fivethirtyeight](#)

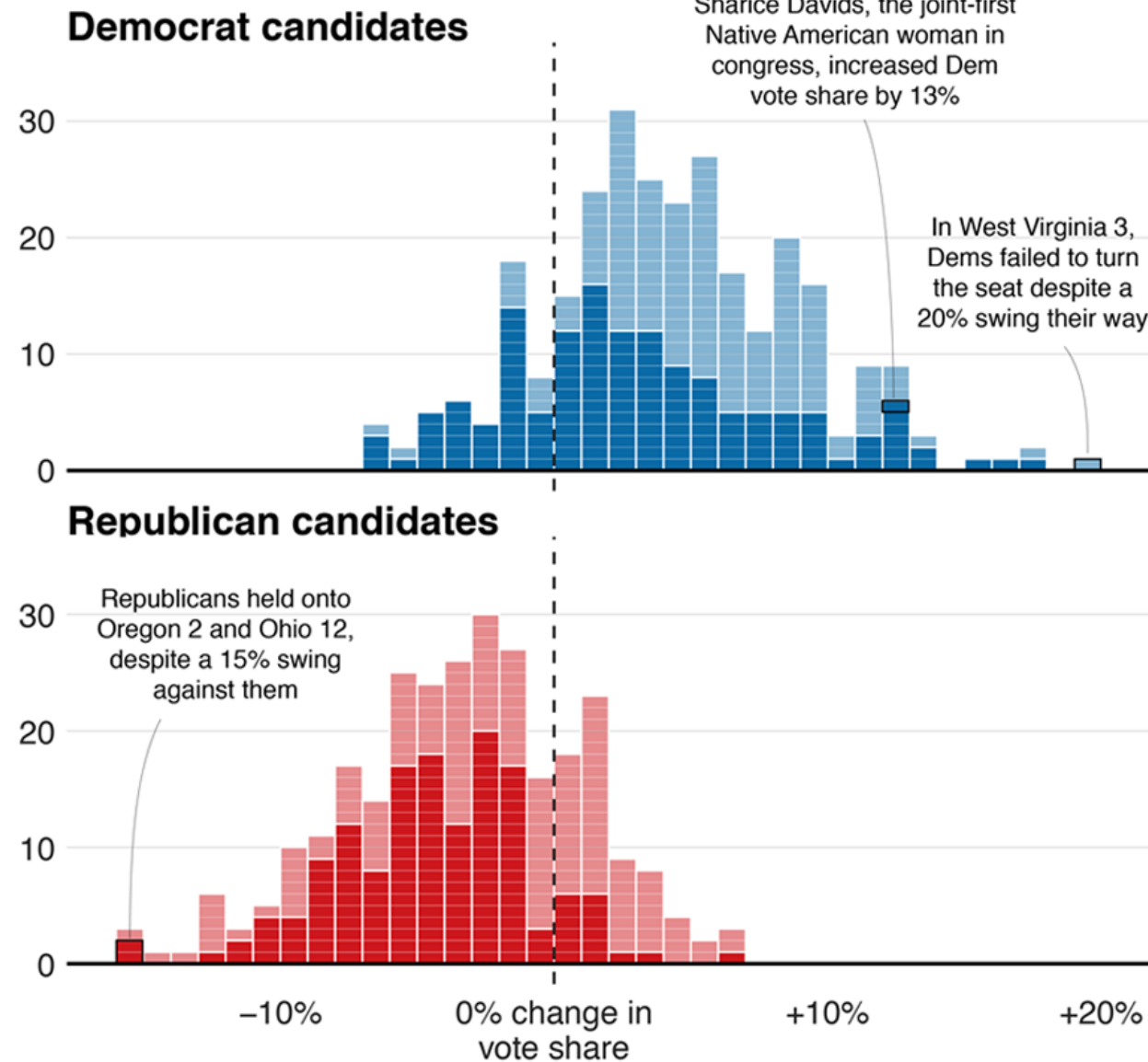


Source: [fivethirtyeight](#)

# ggplot: All Your Figure are Belong to Us

## Blue wave

■ Won seat ■ Didn't win

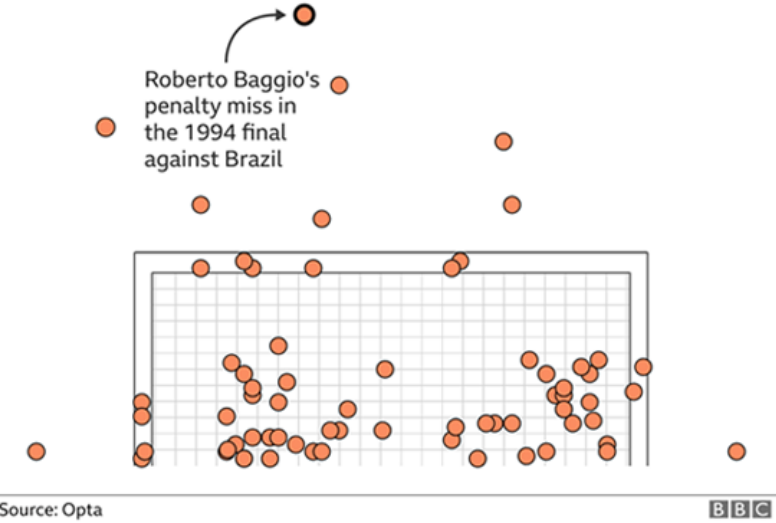


Source: AP, 19:01 ET

BBC

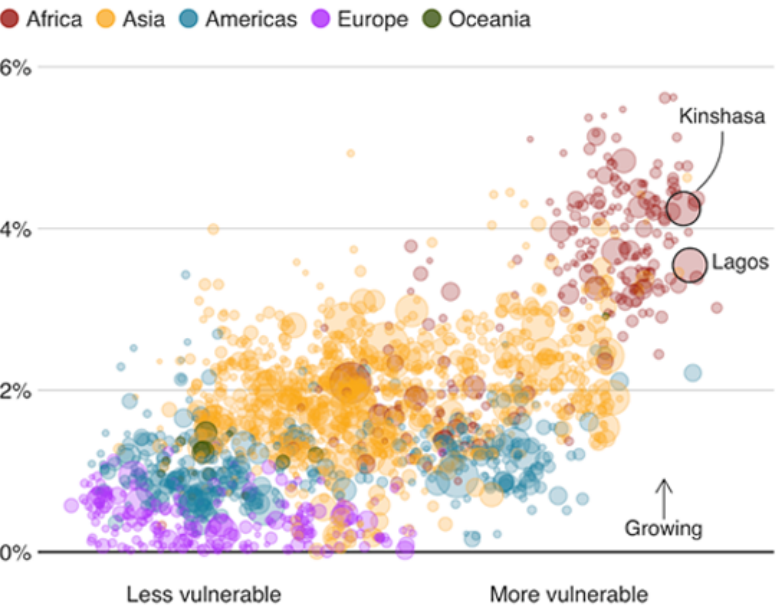
## Where penalties are saved

World Cup shootout misses and saves, 1982-2014



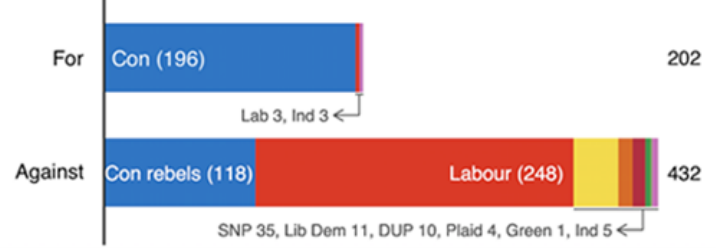
## Fast-growing cities face worse climate risks

Population growth 2018-2035 over climate change vulnerability



Source: Verisk Maplecroft. Circle size represents current population.

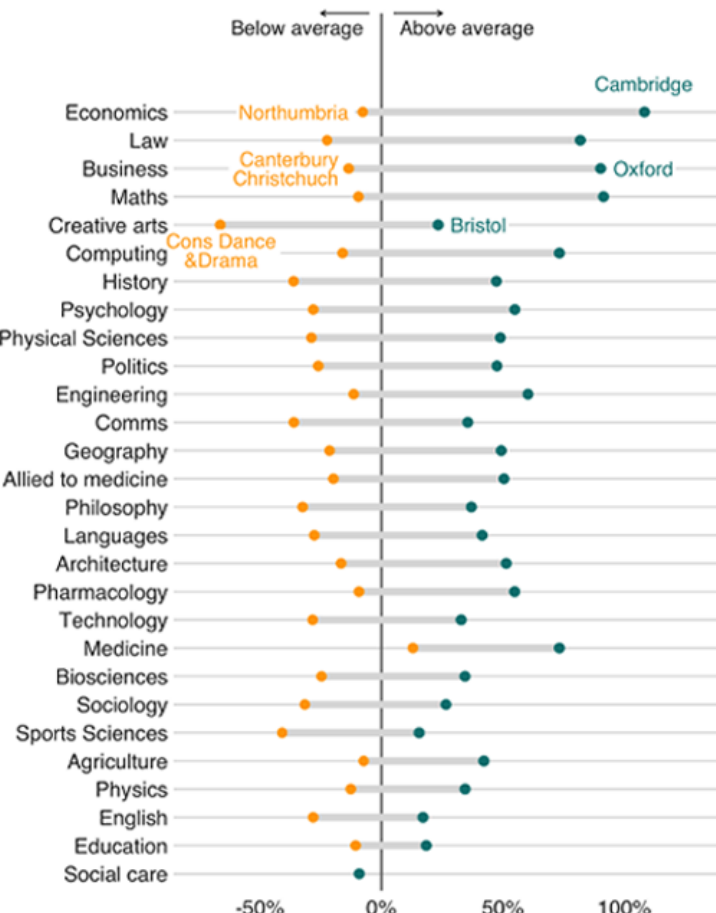
## MPs rejected Theresa May's deal by 230 votes



Source: Commons Votes Services. Excludes 'tellers', the Speaker and deputies

## Earnings vary across unis even within subjects

Impact on men's earnings relative to the average degree



Source: Institute for Fiscal Studies

BBC

Source: BBC's bbplot



# Why Go gg?



Hadley Wickham  
Chief Scientist, R Studio

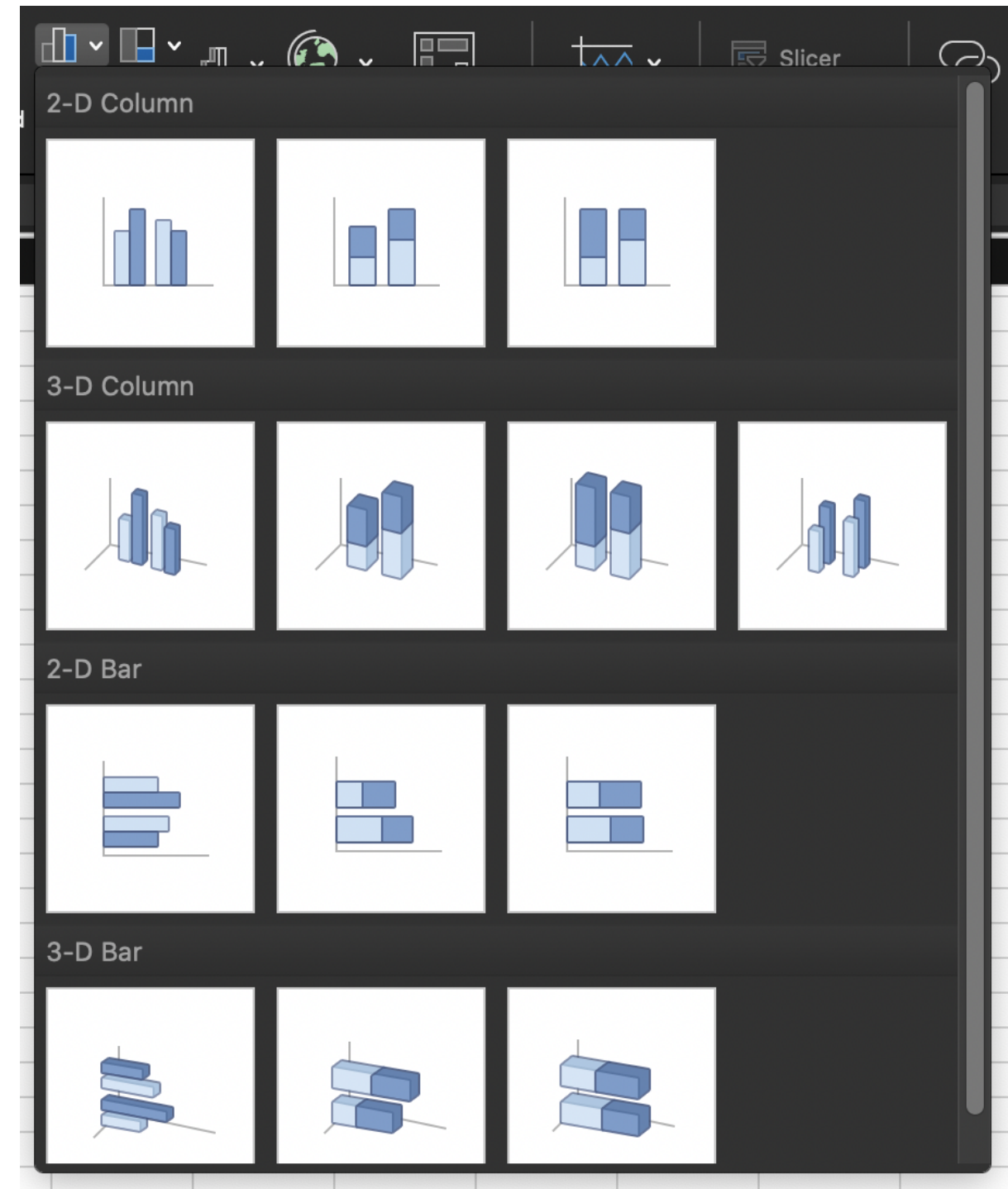
“The transferrable skills from ggplot2 are not the idiosyncracies of plotting syntax, but a powerful way of thinking about visualisation, as a way of **mapping between variables and the visual properties of geometric objects** that you can perceive.”

Source



# The Grammar of Graphics (gg)

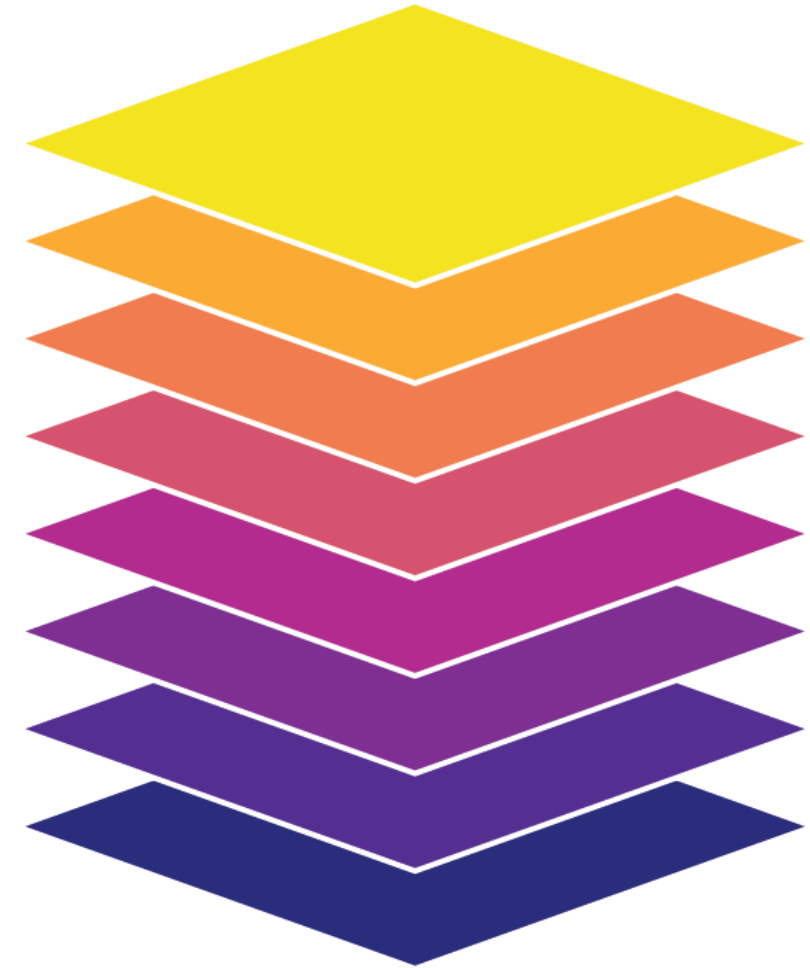
- This is a true *grammar*
- We *don't* talk about specific chart **types**
  - That you have to hunt through in Excel and reshape your data to fit it
- Instead we talk about specific chart **components**



# 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

Theme  
Labels  
Coordinates  
Facets  
Scales  
Geometries  
Aesthetics  
Data



# 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. **scale** define the range of values
  5. **coordinates** to organize location
  6. **labels** describe the scale and markings
  7. **facet** group into subplots
  8. **theme** style the plot elements

Theme  
Labels  
Coordinates  
Facets  
Scales  
Geometries  
Aesthetics  
Data



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

## All in One Command

Produces plot output in viewer

- Does not save plot (if done in console)
  - Save with **Export** menu in viewer
- Adding layers requires whole code for new plot
- Perfectly fine if it's a code chunk in a Quarto document!

```
1 ggplot(data = mpg) +  
2   aes(x = displ,  
3       y = hwy) +  
4   geom_point() +  
5   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

```
1 # make and save plot as p
2 p <- ggplot(data = mpg) +
3   aes(x = displ,
4       y = hwy) +
5   geom_point()
6
7 p # view plot
8
9 # to add a layer...
10 p + geom_smooth() # shows the new plot
11
12 p <- p + geom_smooth() # overwrites p
13 p2 <- p + geom_smooth() # saves new object
```

# Plot Layers

# The Grammar of Graphics (gg): Tidy 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



# gg: Data Layer

## Data

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

- 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

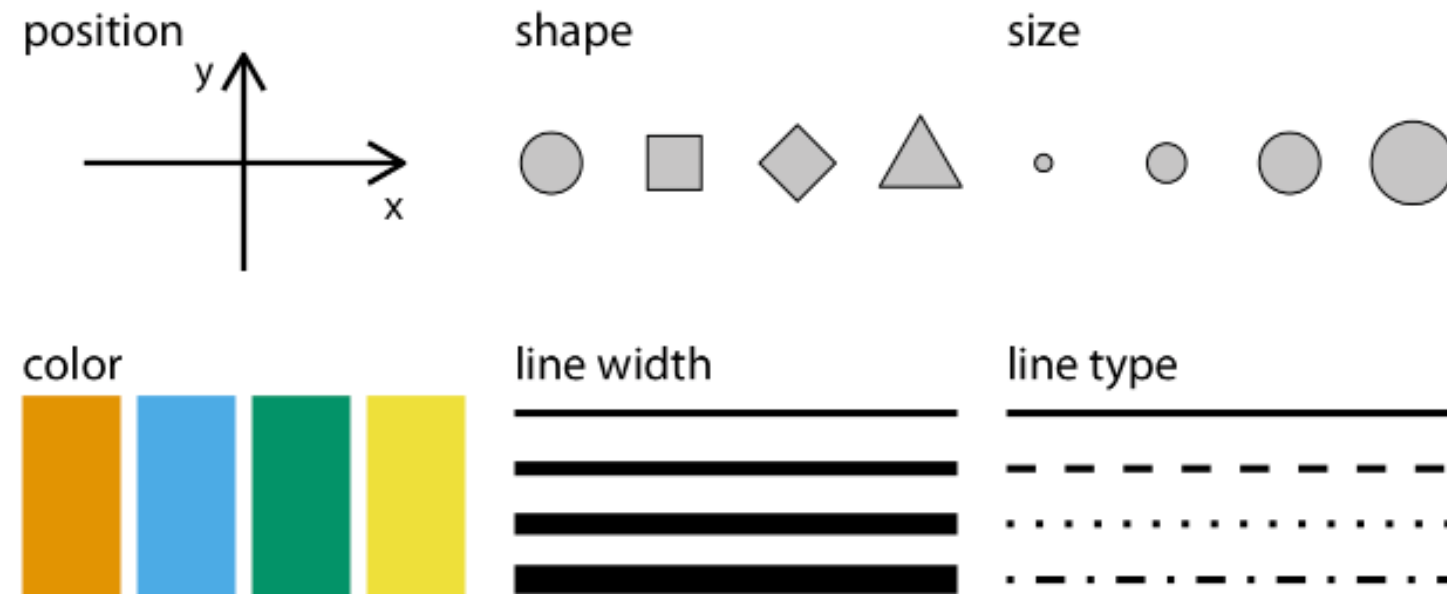
# gg: Mapping Aesthetics I

Data

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

Aesthetics

`+aes(...)`



# gg: Mapping Aesthetics II

**Data**

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

**Aesthetics**

`+aes(...)`

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

# gg: Mapping Aesthetics III

**Data**

**Aesthetics**

`+aes(...)`

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

- `displ` → **x**
- `hwy` → **y**
- `class` → **color**, (or **shape**, **size**, etc.)

# gg: Mapping Aesthetics IV

**Data**

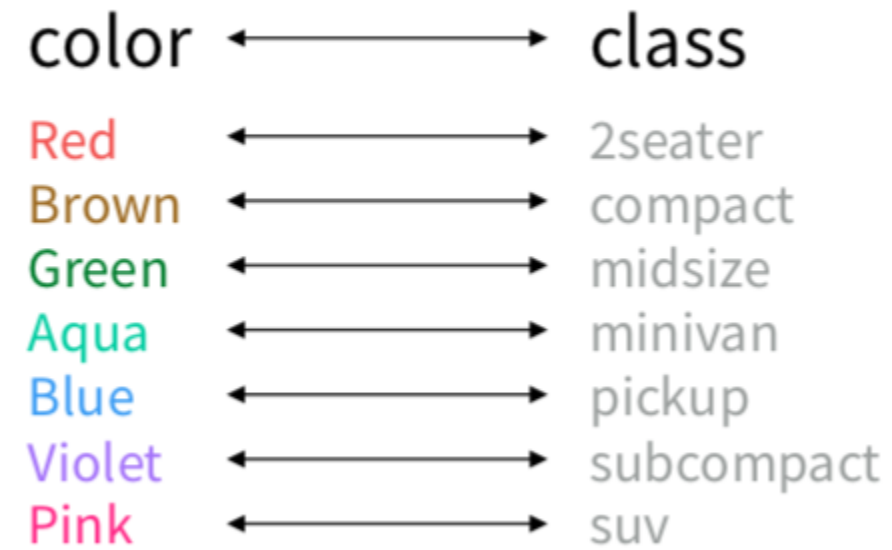
**Aesthetics** map data to visual elements or parameters

**Aesthetics**

`+aes(...)`

**Visual Space**

**Data Space**



# gg: Mapping Aesthetics V

**Data**

**Aesthetics**

+aes(...)

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

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

# gg: Geoms I

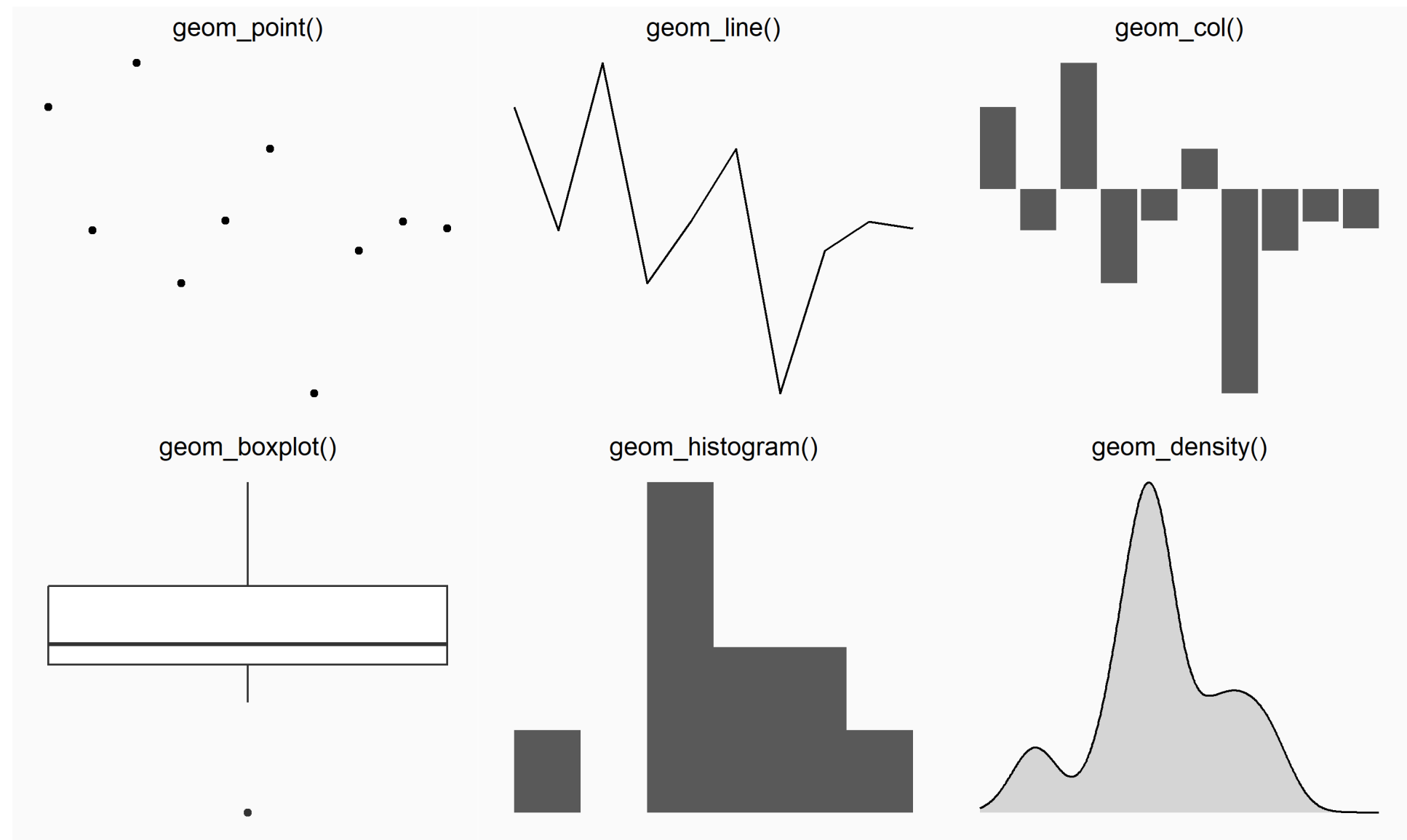
Data

Aesthetics

Geoms

`+geom_*(...)`

**Geometric objects** displayed on the plot





# gg: Geoms II

Data

Aesthetics

Geoms

+geom\_\*(...)

**Geometric objects** displayed on the plot

- What **geoms** you should use depends on what you want to show:

Type	geom
Point	geom_point()
Line	geom_line(), geom_path()
Bar	geom_bar(), geom_col()
Histogram	geom_histogram()
Regression	geom_smooth()
Boxplot	geom_boxplot()
Text	geom_text()
Density	geom_density()

# gg: Geoms III

Data

Aesthetics

Geoms

+geom\_\*(...)

**Geometric objects** displayed on the plot

1	##	[1]	"geom_abline"	"geom_area"	"geom_bar"	"geom_bin2d"
2	##	[5]	"geom_blank"	"geom_boxplot"	"geom_col"	"geom_contour"
3	##	[9]	"geom_count"	"geom_crossbar"	"geom_curve"	"geom_dens"
4	##	[13]	"geom_density_2d"	"geom_density2d"	"geom_dotplot"	"geom_errorbar"
5	##	[17]	"geom_errorbarh"	"geom_freqpoly"	"geom_hex"	"geom_histogram"
6	##	[21]	"geom_hline"	"geom_jitter"	"geom_label"	"geom_line"
7	##	[25]	"geom_linerange"	"geom_map"	"geom_path"	"geom_point"
8	##	[29]	"geom_pointrange"	"geom_polygon"	"geom_qq"	"geom_qq_line"
9	##	[33]	"geom_quantile"	"geom_raster"	"geom_rect"	"geom_ribbon"
10	##	[37]	"geom_rug"	"geom_segment"	"geom_sf"	"geom_sf_line"
11	##	[41]	"geom_sf_text"	"geom_smooth"	"geom_spoke"	"geom_step"
12	##	[45]	"geom_text"	"geom_tile"	"geom_violin"	"geom_vline"

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

# gg: Geoms IV

Data

Aesthetics

Geoms

+geom\_\*(...)

**Geometric objects** displayed on the plot

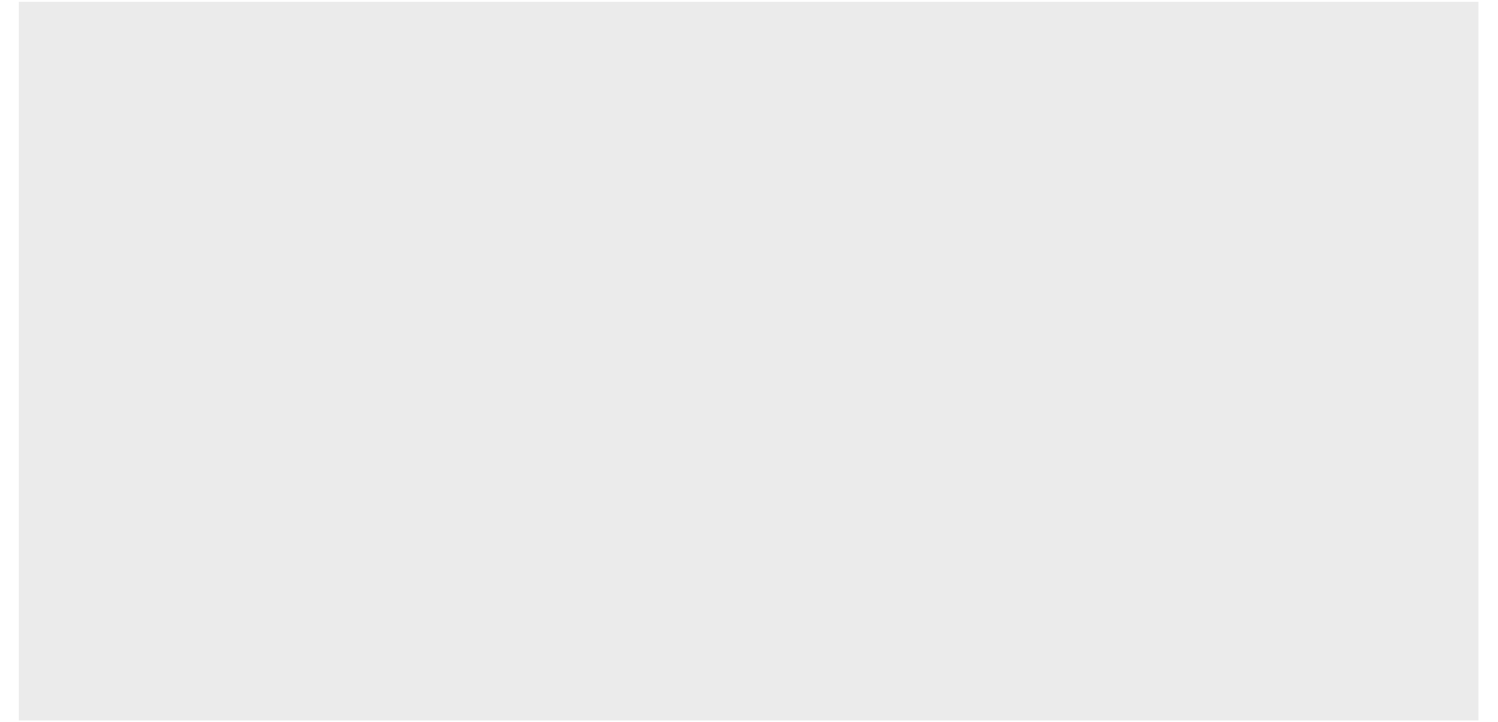
Or just start typing `geom_` in R Studio!

```
ggplot(df_geom) +  
  aes(x, y) +
```

```
|
```

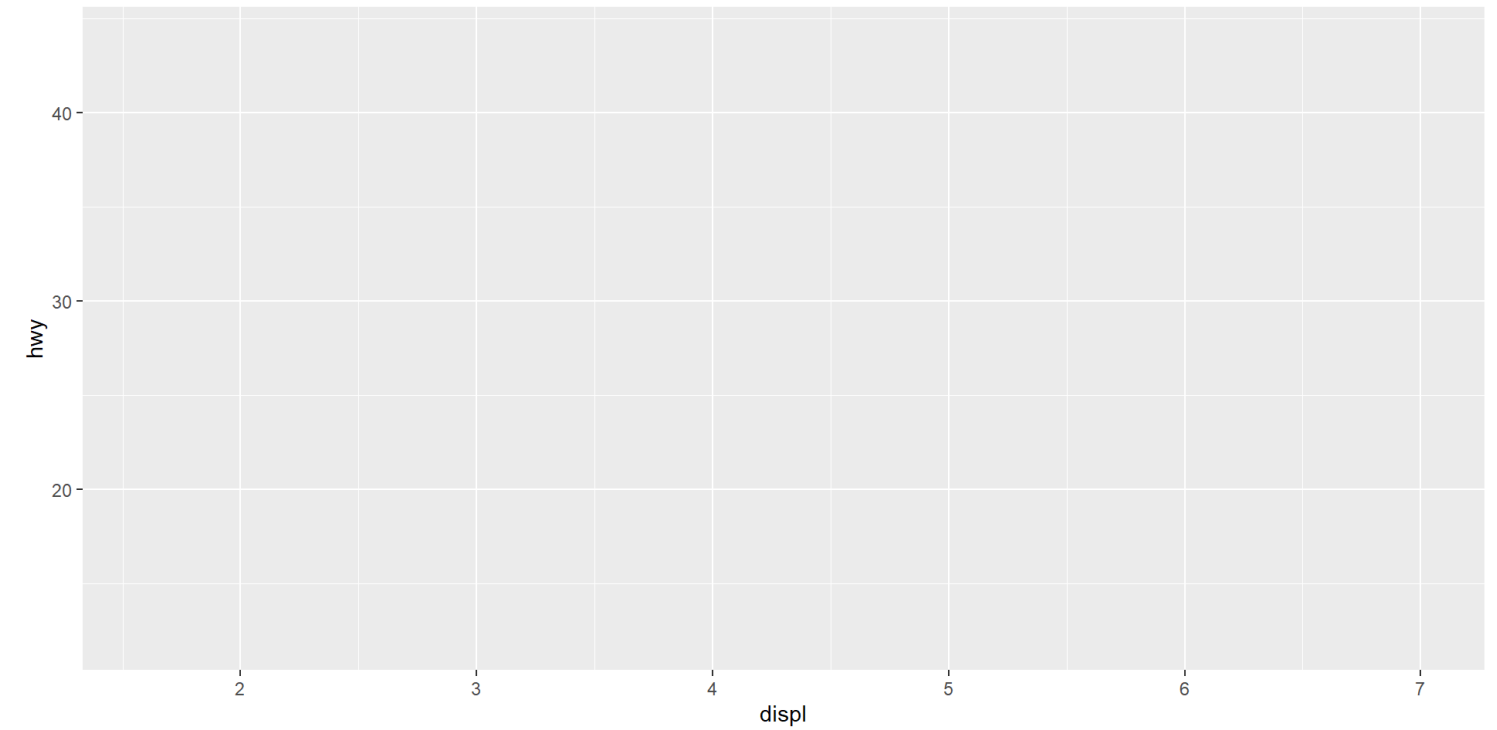
# Let's Make a Plot!

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



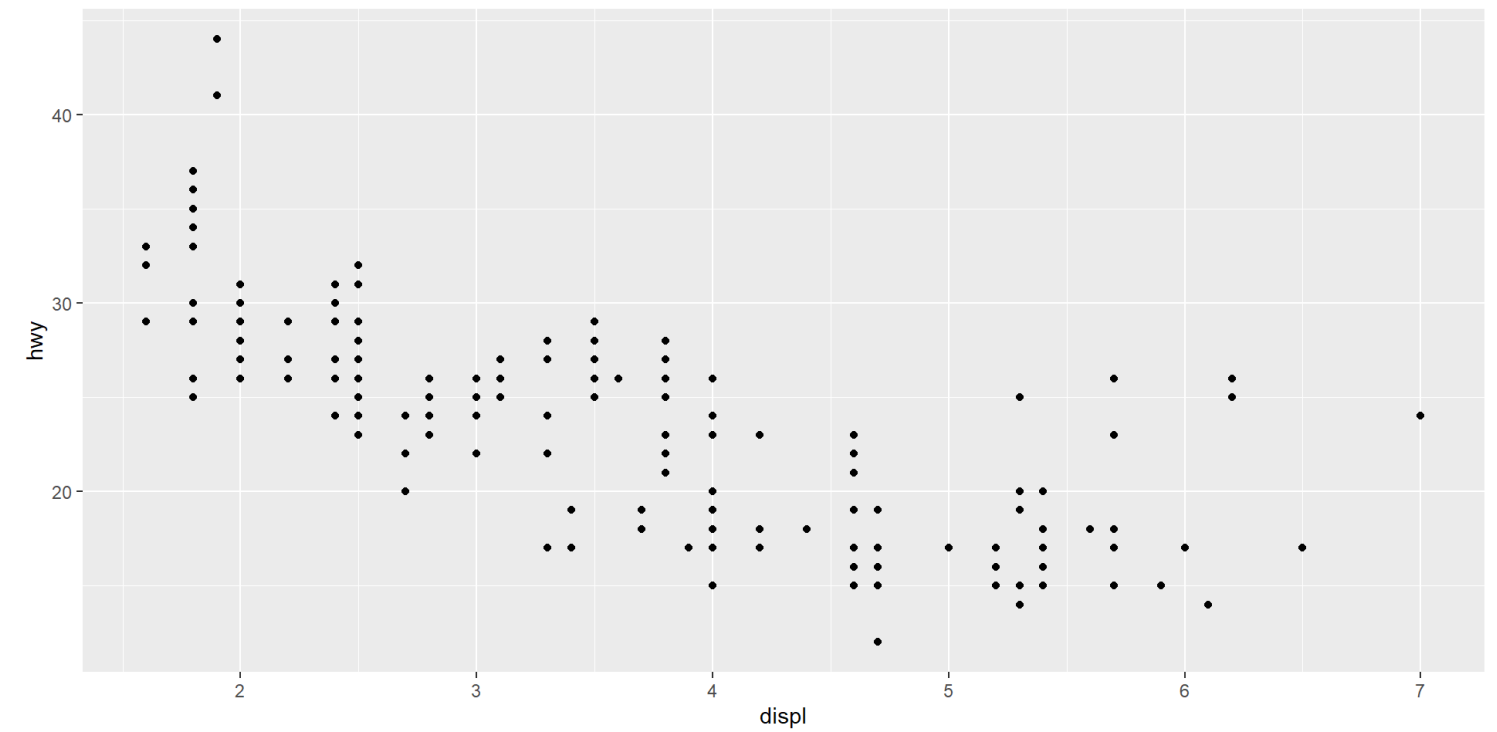
# Let's Make a Plot!

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



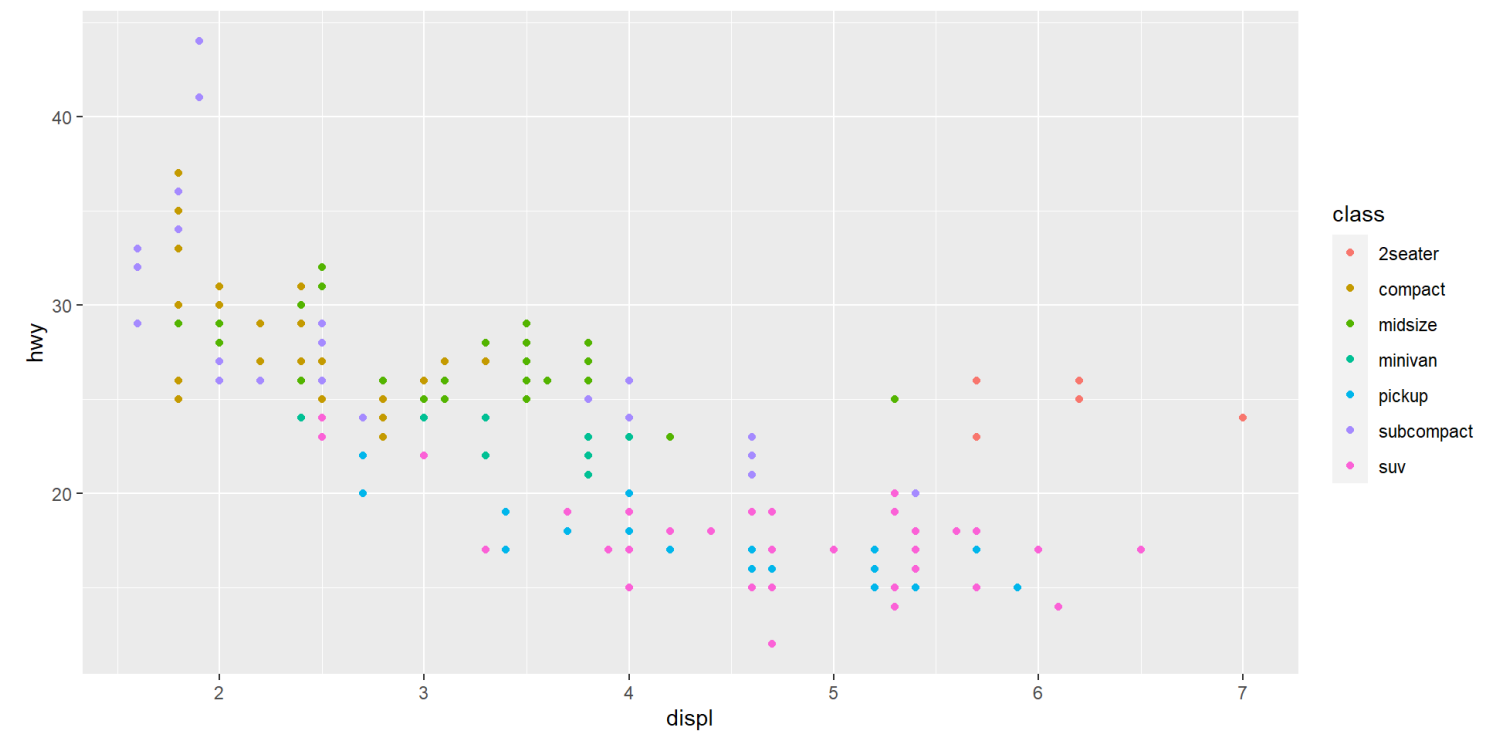
# Let's Make a Plot!

```
1 ggplot(data = mpg) +  
2   aes(x = displ,  
3       y = hwy) +  
4   geom_point()
```



# Let's Make a Plot!

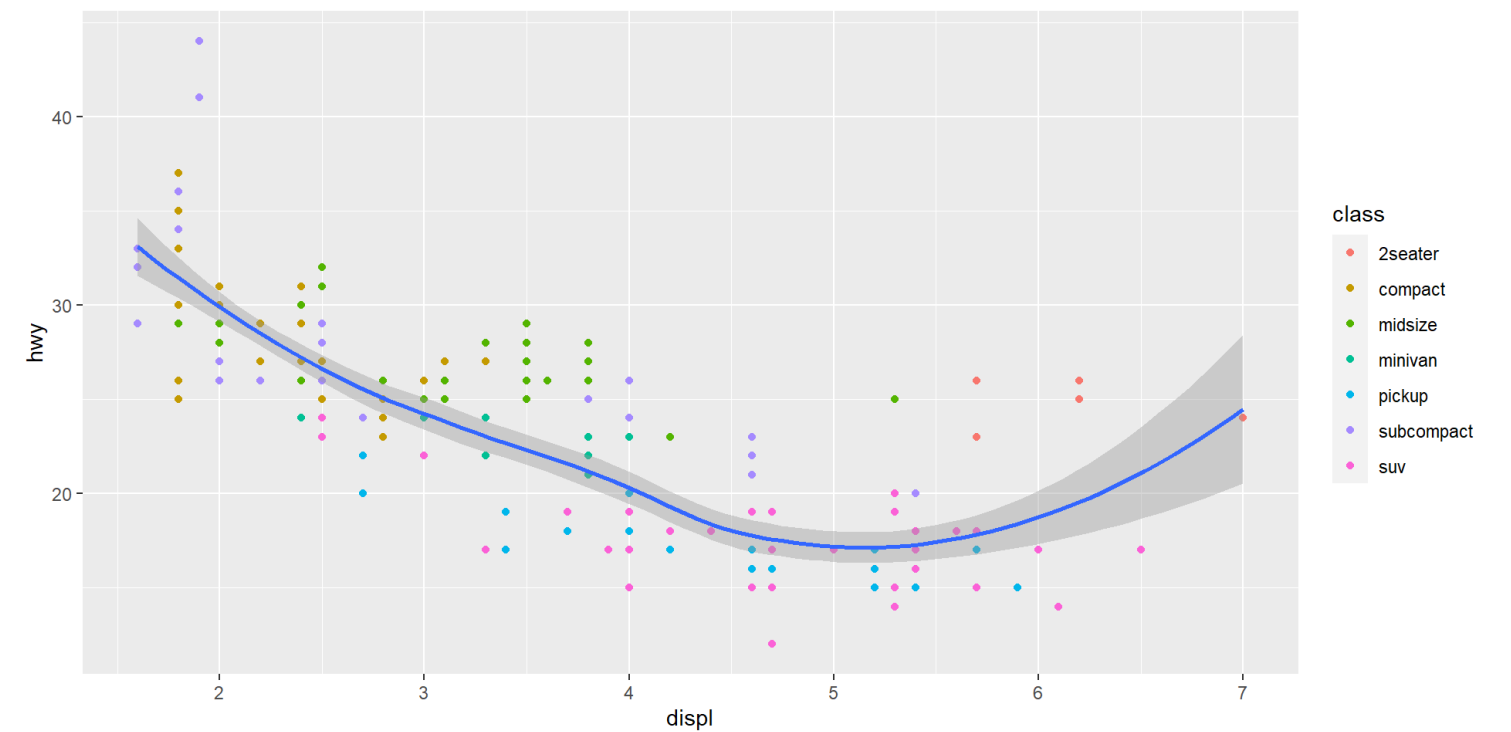
```
1 ggplot(data = mpg) +  
2   aes(x = displ,  
3       y = hwy) +  
4   geom_point(aes(color = class))
```





# Let's Make a Plot!

```
1 ggplot(data = mpg)+  
2   aes(x = displ,  
3       y = hwy)+  
4   geom_point(aes(color = class))+  
5   geom_smooth()
```



# More Geoms

## Data

## Aesthetics

## Geoms

`+geom_*(...)`

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

- `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

# More Geoms II

Data

Aesthetics

Geoms

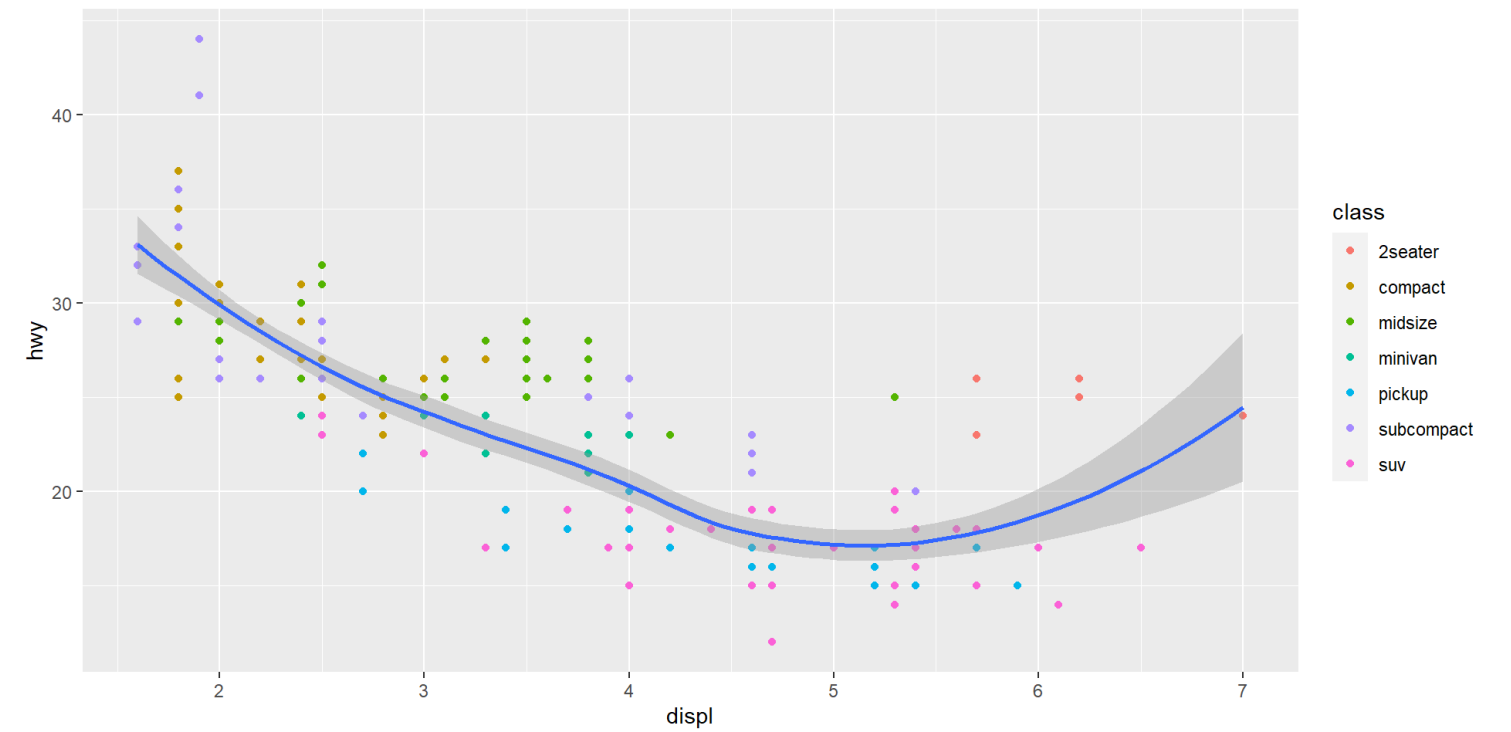
+geom\_\*(...)

geom\_\*(aes, data, stat, position)

- **stat**: some geoms statistically transform data
  - `geom_histogram()` uses `stat_bin()` to group observations into bins
- **position**: some adjust location of objects
  - `dodge`, `stack`, `jitter`

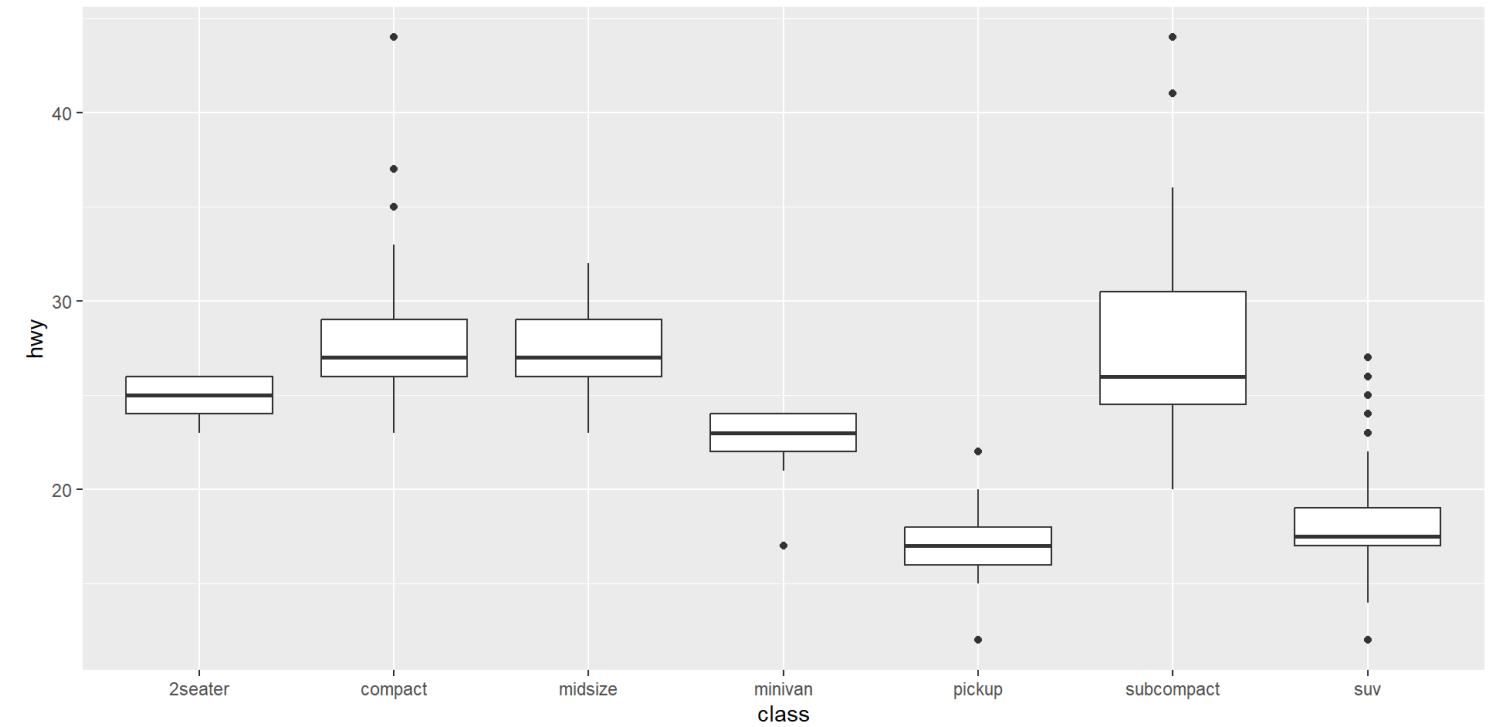
# Our Plot

```
1 ggplot(data = mpg)+  
2   aes(x = displ,  
3       y = hwy)+  
4   geom_point(aes(color = class))+  
5   geom_smooth()
```



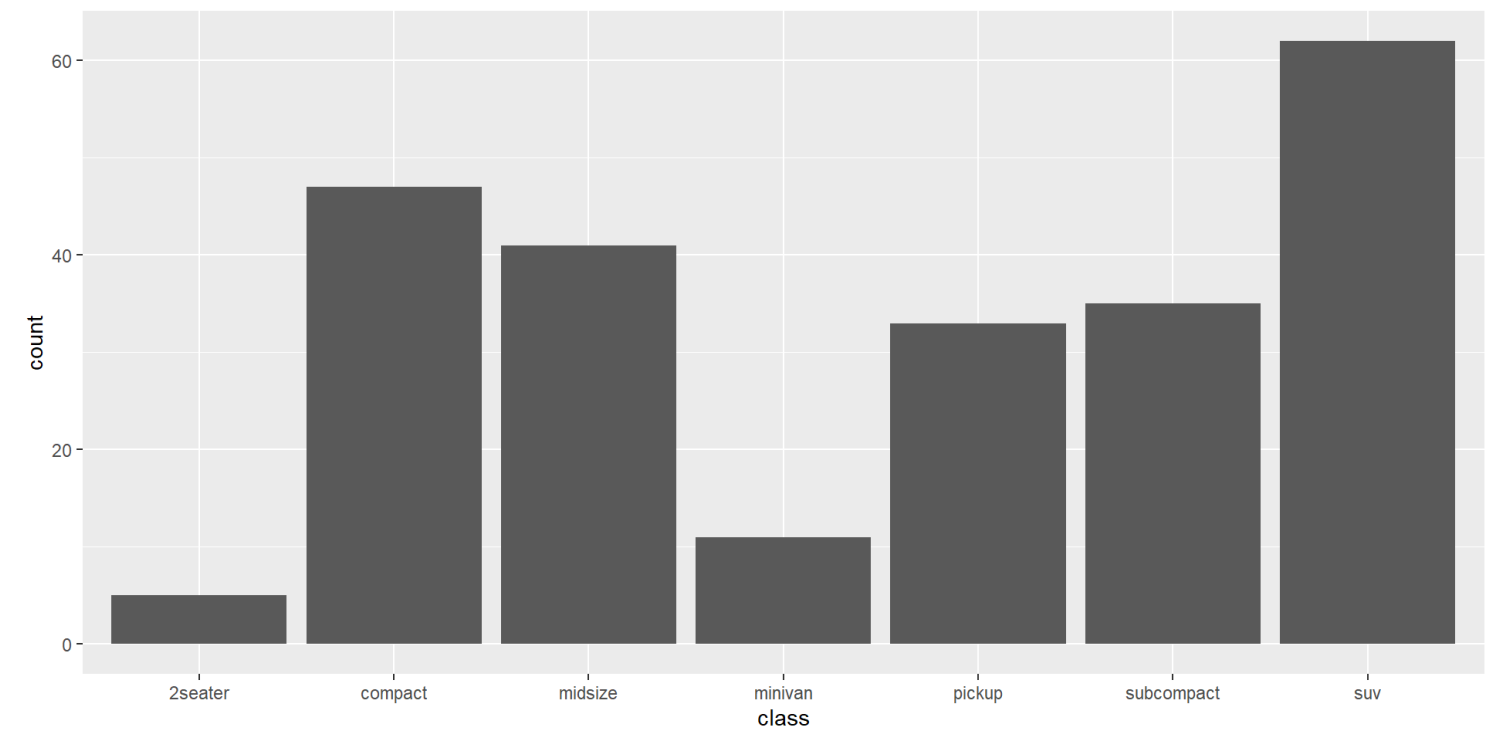
# Change Our Plot

```
1 ggplot(data = mpg) +  
2   aes(x = class,  
3       y = hwy) +  
4   geom_boxplot()
```



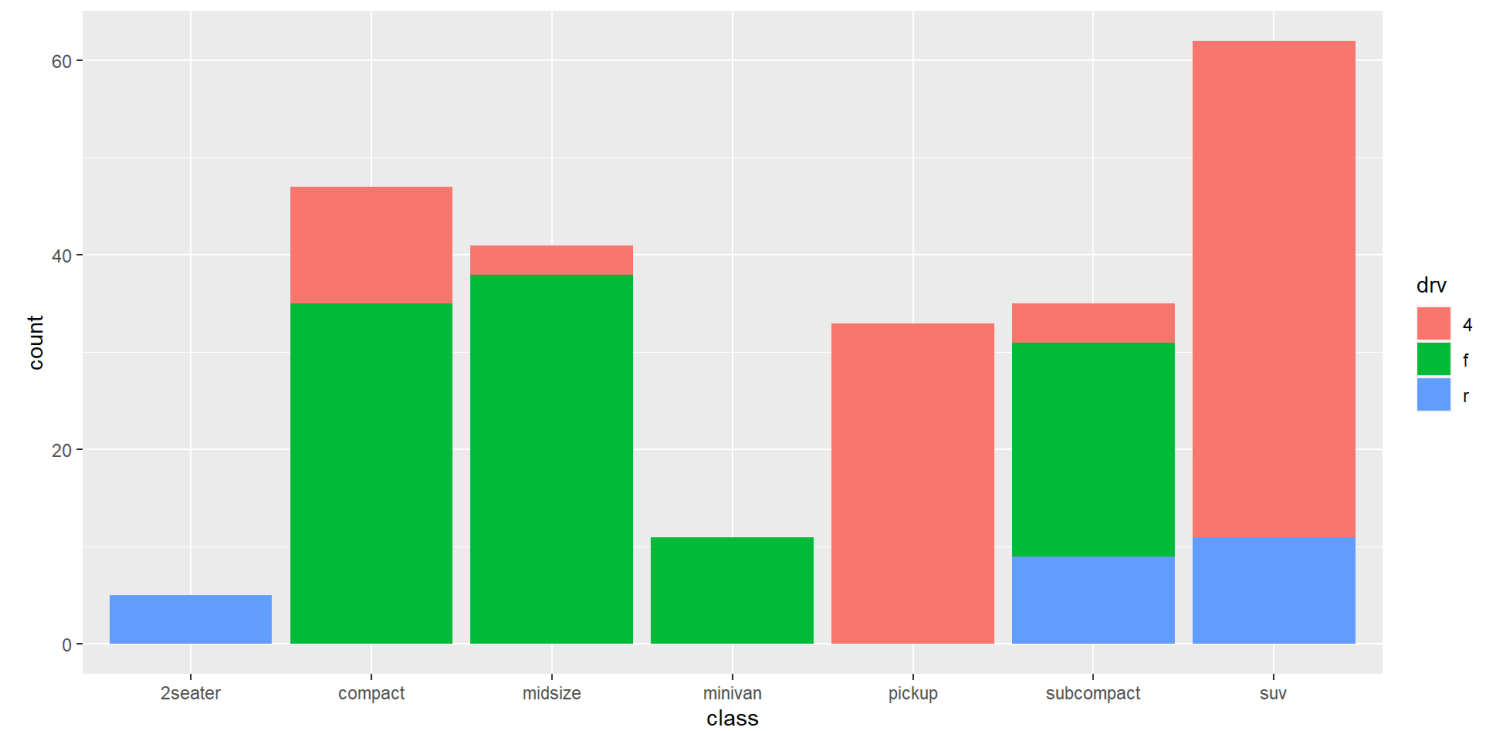
# Change Our Plot

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



# Change Our Plot

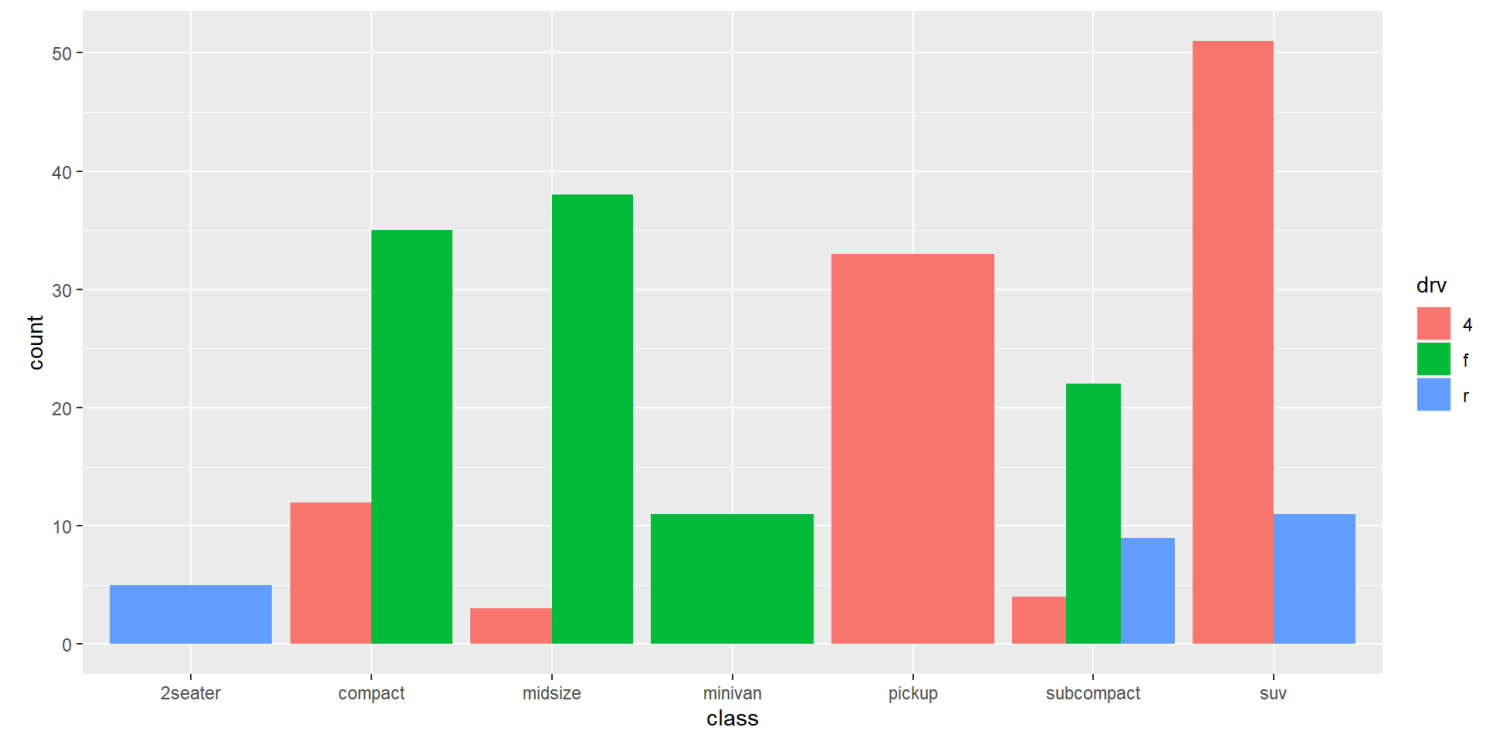
```
1 ggplot(data = mpg) +  
2   aes(x = class,  
3       fill = drv) +  
4   geom_bar()
```





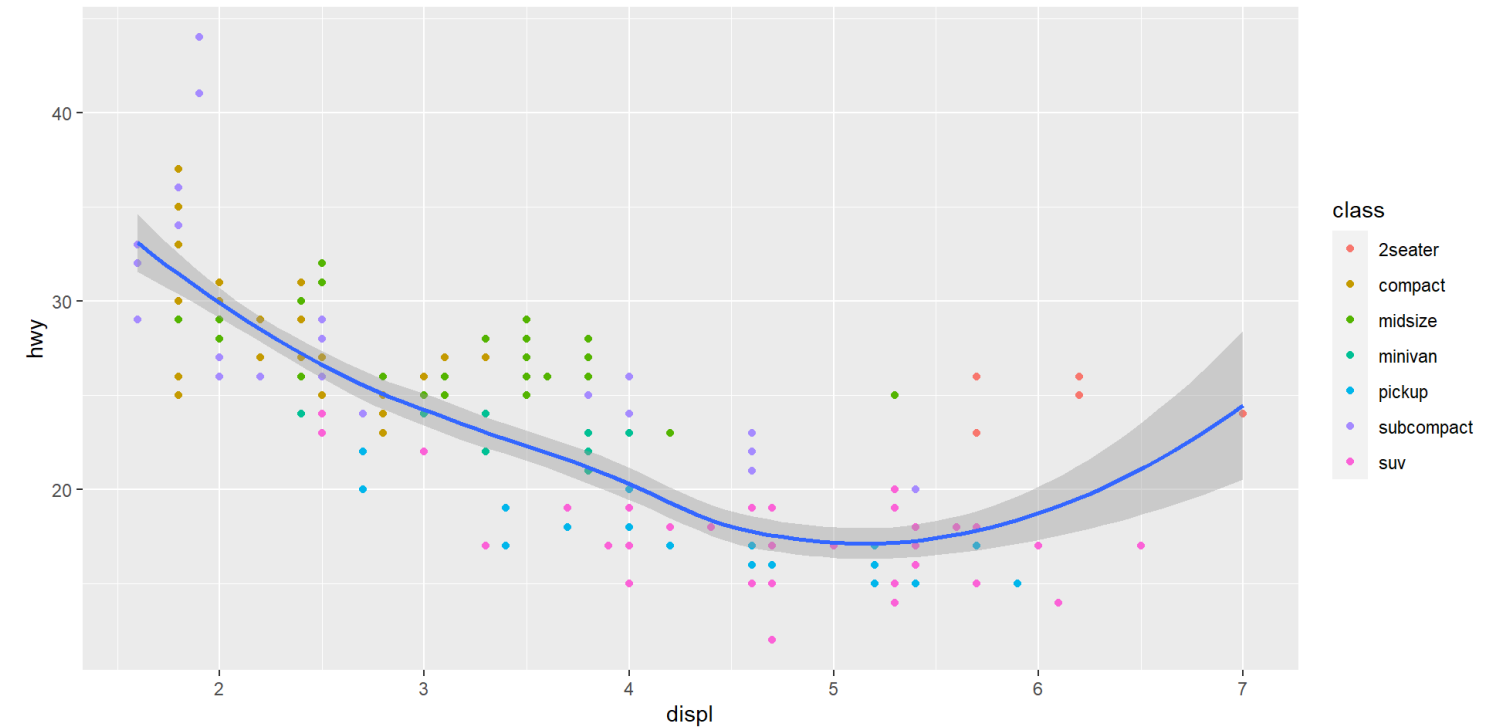
# Change Our Plot

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



# Back to the Original (and Saving It)

```
1 # save plot as p
2 p <- ggplot(data = mpg) +
3   aes(x = displ,
4       y = hwy) +
5   geom_point(aes(color = class)) +
6   geom_smooth()
7
8 p # show plot
```



# gg: Facets I

Data

Aesthetics

Geoms

Facets

+ facet\_wrap()

+ facet\_grid()



# gg: Facets II

Data

Aesthetics

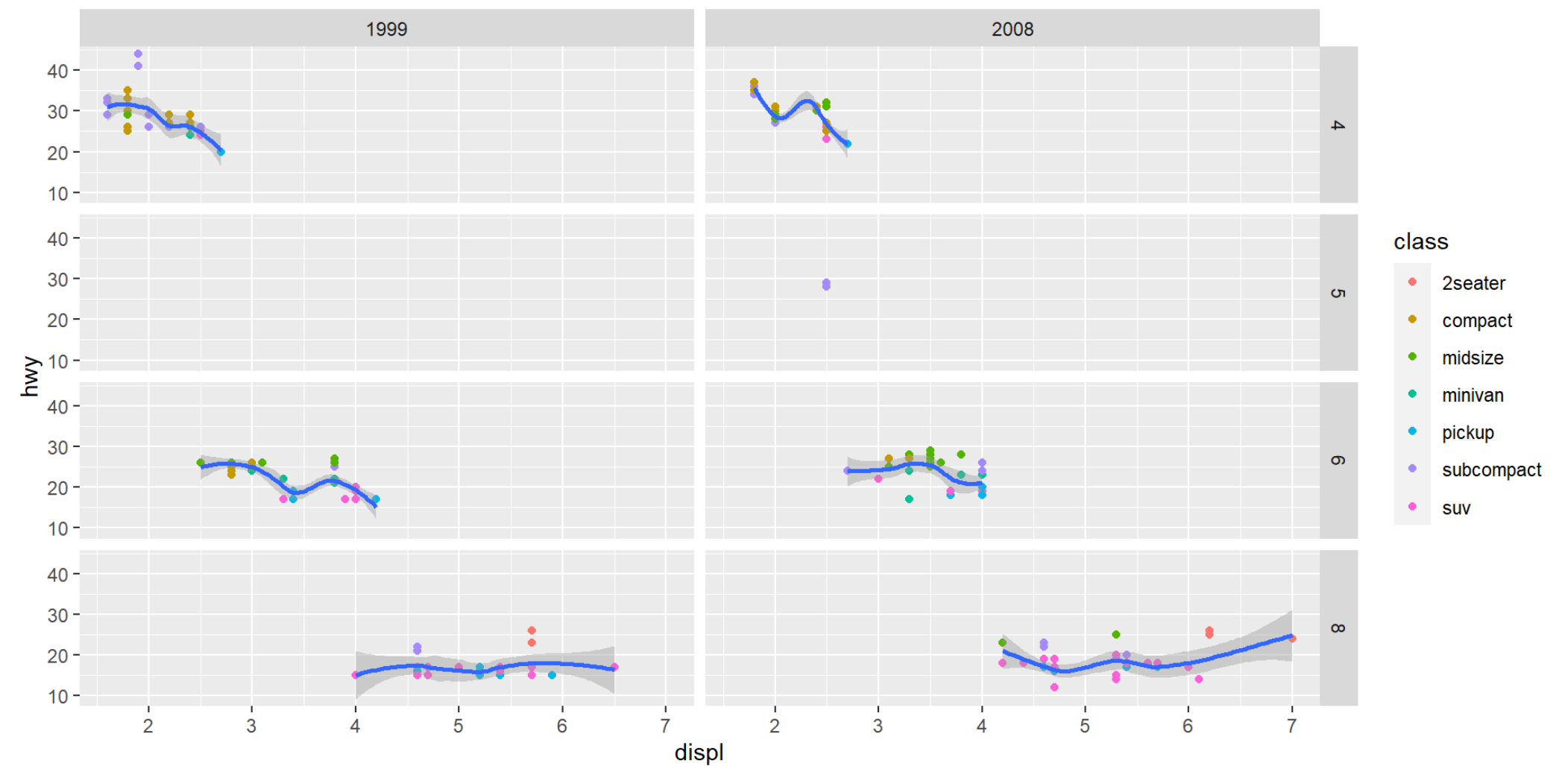
Geoms

Facets

+ facet\_wrap()

+ facet\_grid()

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



# gg: Labels

## Data

## Aesthetics

## Geoms

## Facets

+ `labs()`

```
1 (p <- p + facet_wrap(~year)+  
2   labs(x = "Engine Displacement (Liters)",  
3         y = "Highway MPG",  
4         title = "Car Mileage and Displacement",  
5         subtitle = "More Displacement Lowers Highway MPG",  
6         caption = "Source: EPA",  
7         color = "Vehicle Class"))
```



# gg: Scales I

Data

Aesthetics

Geoms

Facets

Scales

+ `scale_*_*()`

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

- `<aes>`: parameter to adjust
- `<type>`: type of parameter
- Discrete x-axis: `scale_x_discrete()`
- Continuous y-axis: `scale_y_continuous()`
- Rescale x-axis to log: `scale_x_log10()`
- Use different color palette: `scale_fill_discrete()`,  
`scale_color_manual()`

# gg: Scales II

Data

Aesthetics

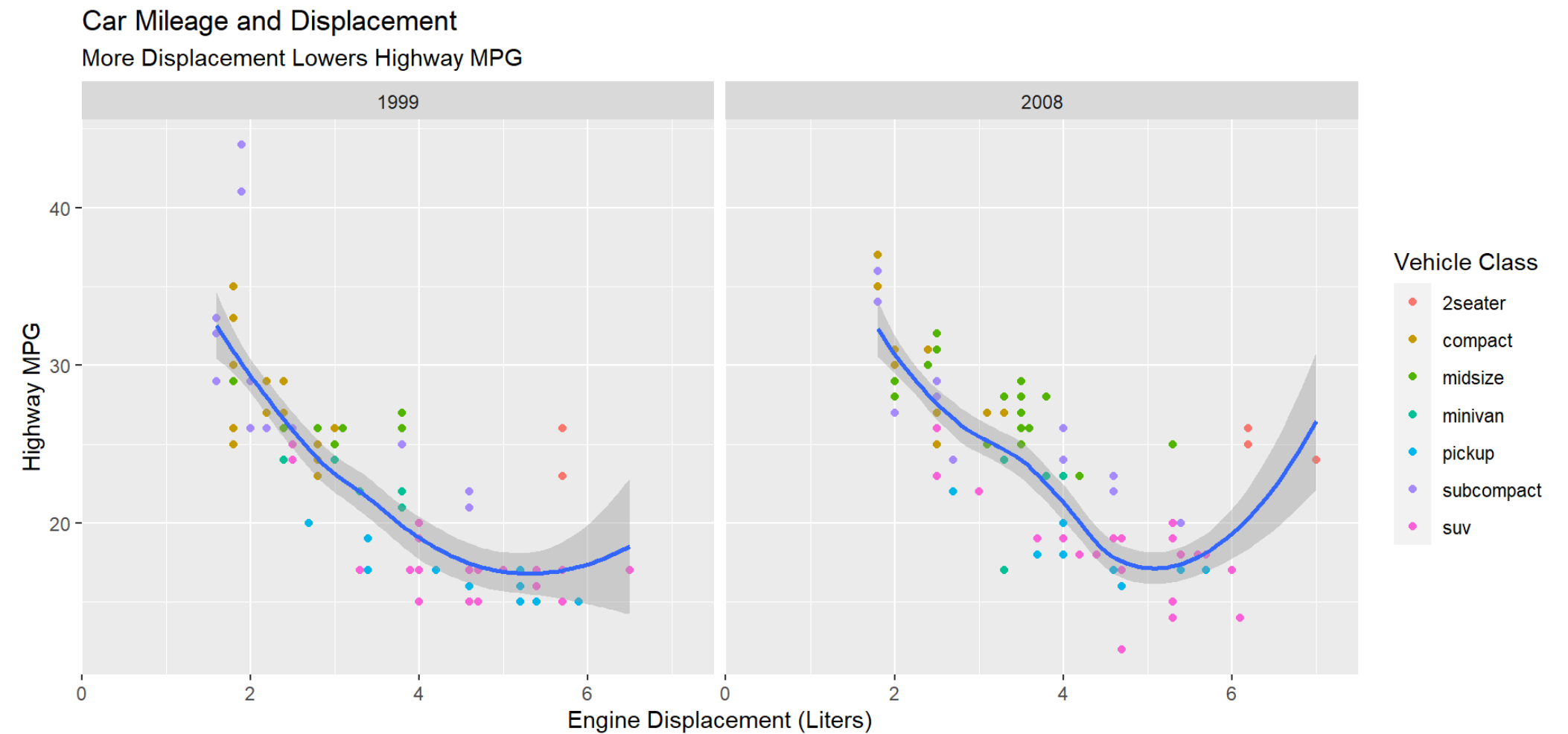
Geoms

Facets

Scales

+ scale\_\*\_\*()

```
1 p + scale_x_continuous(breaks = seq(0, 10, 2),  
2                           limits = c(0, 7.5),  
3                           expand = c(0, 0)  
4 )
```



Source: EPA

# gg: Scales II

Data

Aesthetics

Geoms

Facets

Scales

+ scale\_\*\_\*()

```
1 p + scale_x_continuous(breaks = seq(0, 10, 2),  
2                          limits = c(0, 7.5),  
3                          expand = c(0, 0)  
4                          ) +  
5 scale_color_viridis_d()
```



Source: EPA



# gg: Themes I

**Data**

**Aesthetics**

**Geoms**

**Facets**

**Scales**

**Themes**

+ theme\_\*

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

- Some themes that come with `ggplot2`:
  - + `theme_bw()`
  - + `theme_dark()`
  - + `theme_gray()`
  - + `theme_minimal()`
  - + `theme_light()`
  - + `theme_classic()`

# gg: Themes II

**Data**

**Aesthetics**

**Geoms**

**Facets**

**Scales**

**Themes**

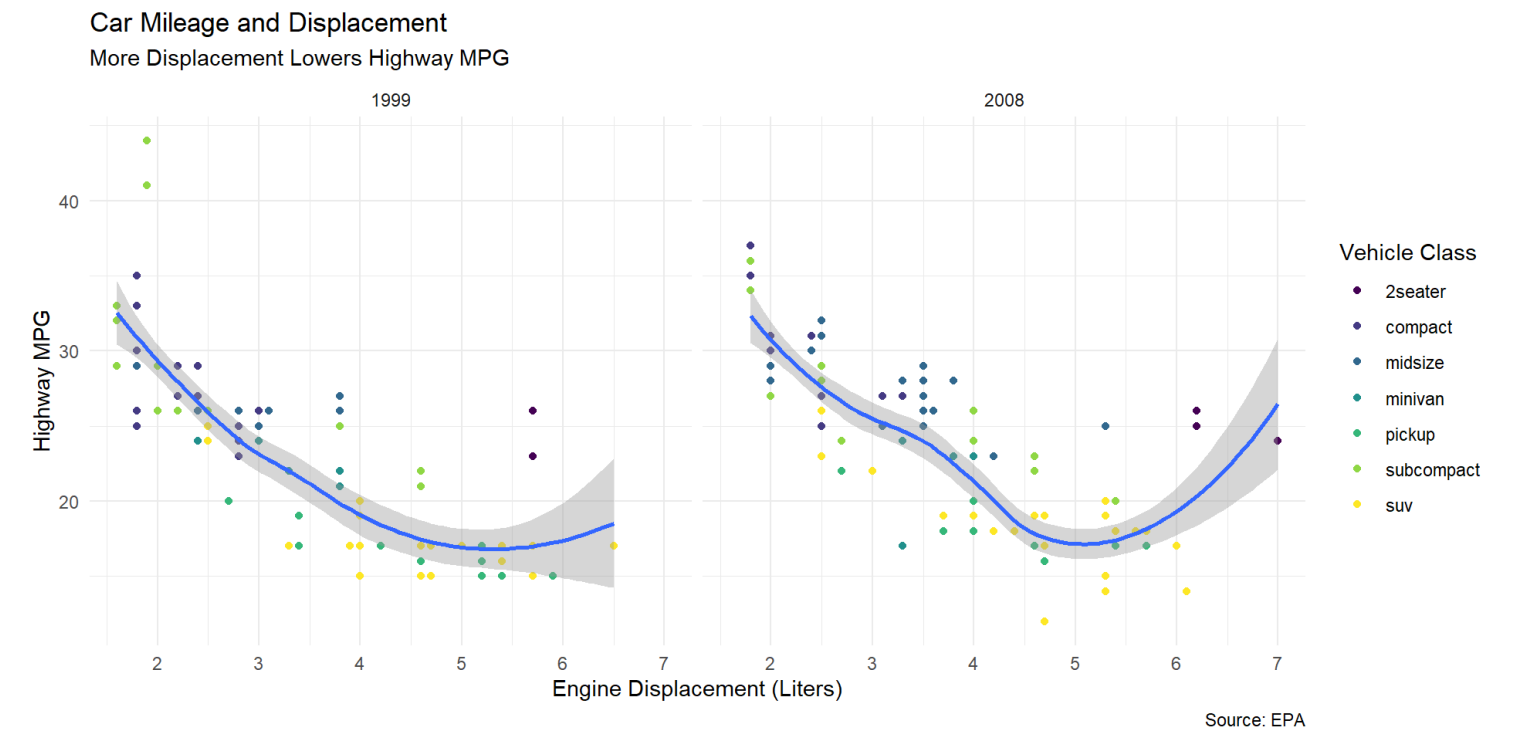
+ `theme_*`()

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

- Many parameters we could customize
- Global options: `line`, `rect`, `text`, `title`
- `axis`: x-, y-, or other axis title, ticks, lines
- `legend`: plot legends for fill or color
- `panel`: actual plot area
- `plot`: whole image
- `strip`: facet labels

# gg: Themes III

```
1 ggplot(data = mpg)+
2   aes(x = displ,
3       y = hwy)+
4   geom_point(aes(color = class))+
5   geom_smooth()+
6   facet_wrap(~year)+
7   labs(x = "Engine Displacement (Liters)",
8        y = "Highway MPG",
9        title = "Car Mileage and Displacement",
10       subtitle = "More Displacement Lowers Highway MPG",
11       caption = "Source: EPA",
12       color = "Vehicle Class")+
13   scale_color_viridis_d()+
14   theme_minimal()
```



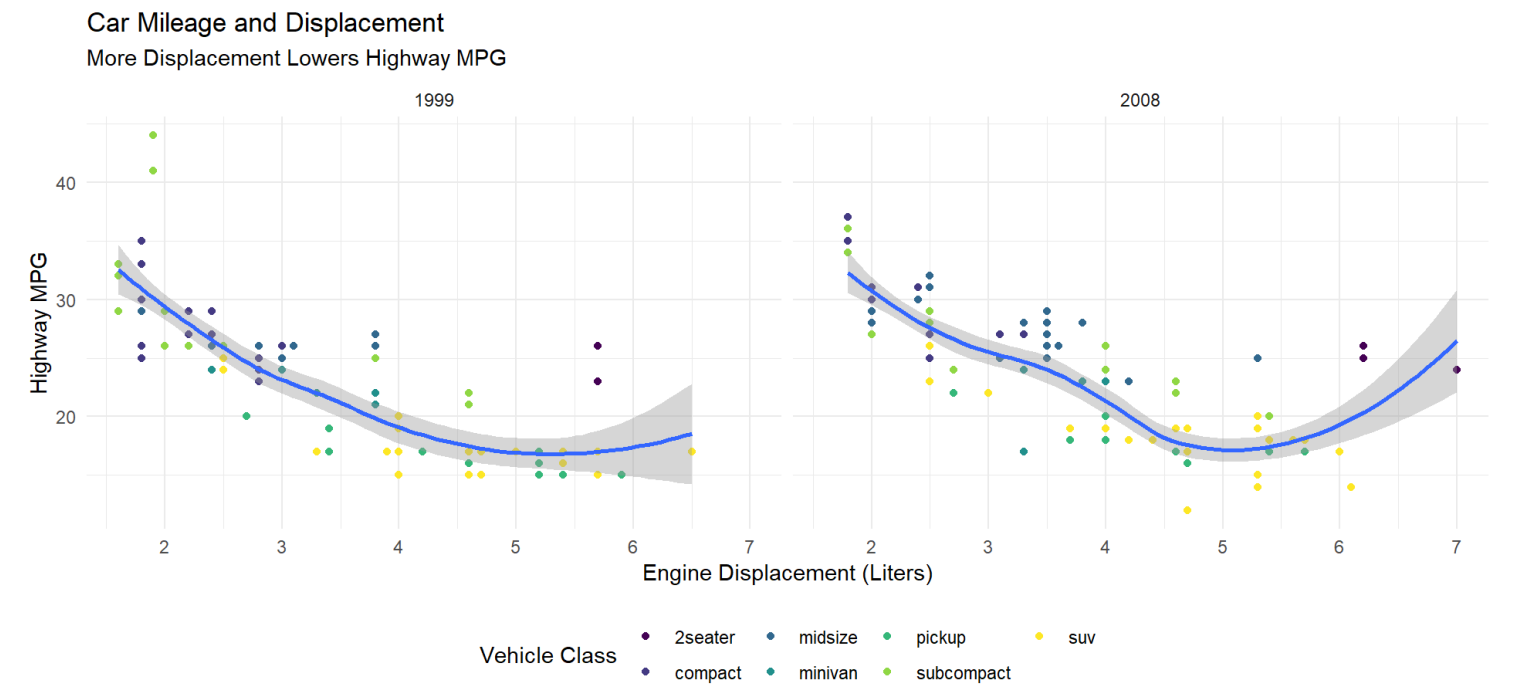
# gg: Themes IV

```
1 ggplot(data = mpg)+
2   aes(x = displ,
3       y = hwy)+
4   geom_point(aes(color = class))+
5   geom_smooth()+
6   facet_wrap(~year)+
7   labs(x = "Engine Displacement (Liters)",
8        y = "Highway MPG",
9        title = "Car Mileage and Displacement",
10       subtitle = "More Displacement Lowers Highway Highwa",
11       caption = "Source: EPA",
12       color = "Vehicle Class")+
13   scale_color_viridis_d()+
14   theme_minimal()+
15   theme(text = element_text(family = "Fira Sans"))
```



# gg: Themes V

```
1 ggplot(data = mpg)+
2   aes(x = displ,
3       y = hwy)+
4   geom_point(aes(color = class))+
5   geom_smooth()+
6   facet_wrap(~year)+
7   labs(x = "Engine Displacement (Liters)",
8        y = "Highway MPG",
9        title = "Car Mileage and Displacement",
10       subtitle = "More Displacement Lowers Highway MPG",
11       caption = "Source: EPA",
12       color = "Vehicle Class")+
13   scale_color_viridis_d()+
14   theme_minimal()+
15   theme(text = element_text(family = "Fira Sans"),
16         legend.position = "bottom")
```



Source: EPA

# gg: Themes VI

Data

Aesthetics

Geoms

Facets

Scales

Themes

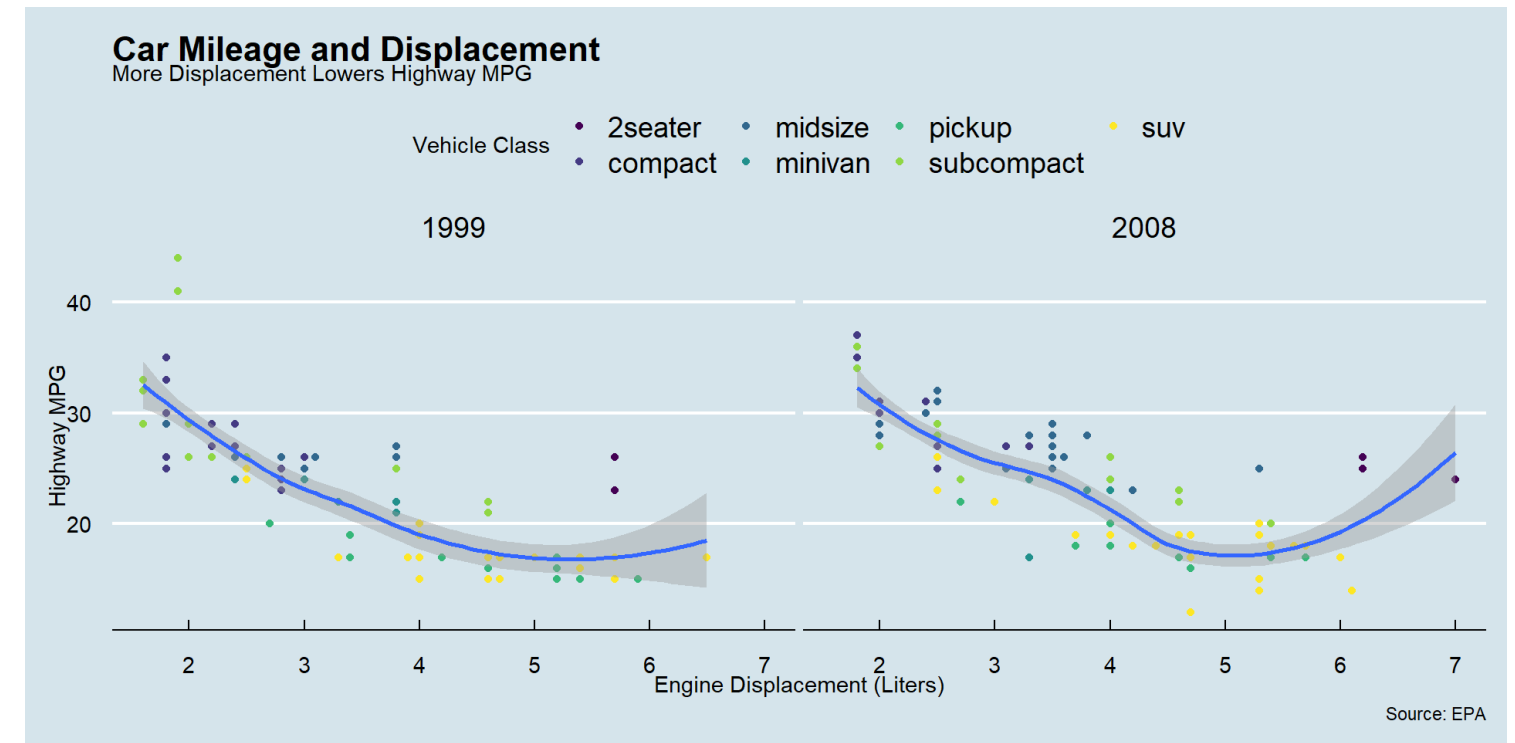
+ theme\_\*()

- `ggthemes` package adds some other nice themes

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

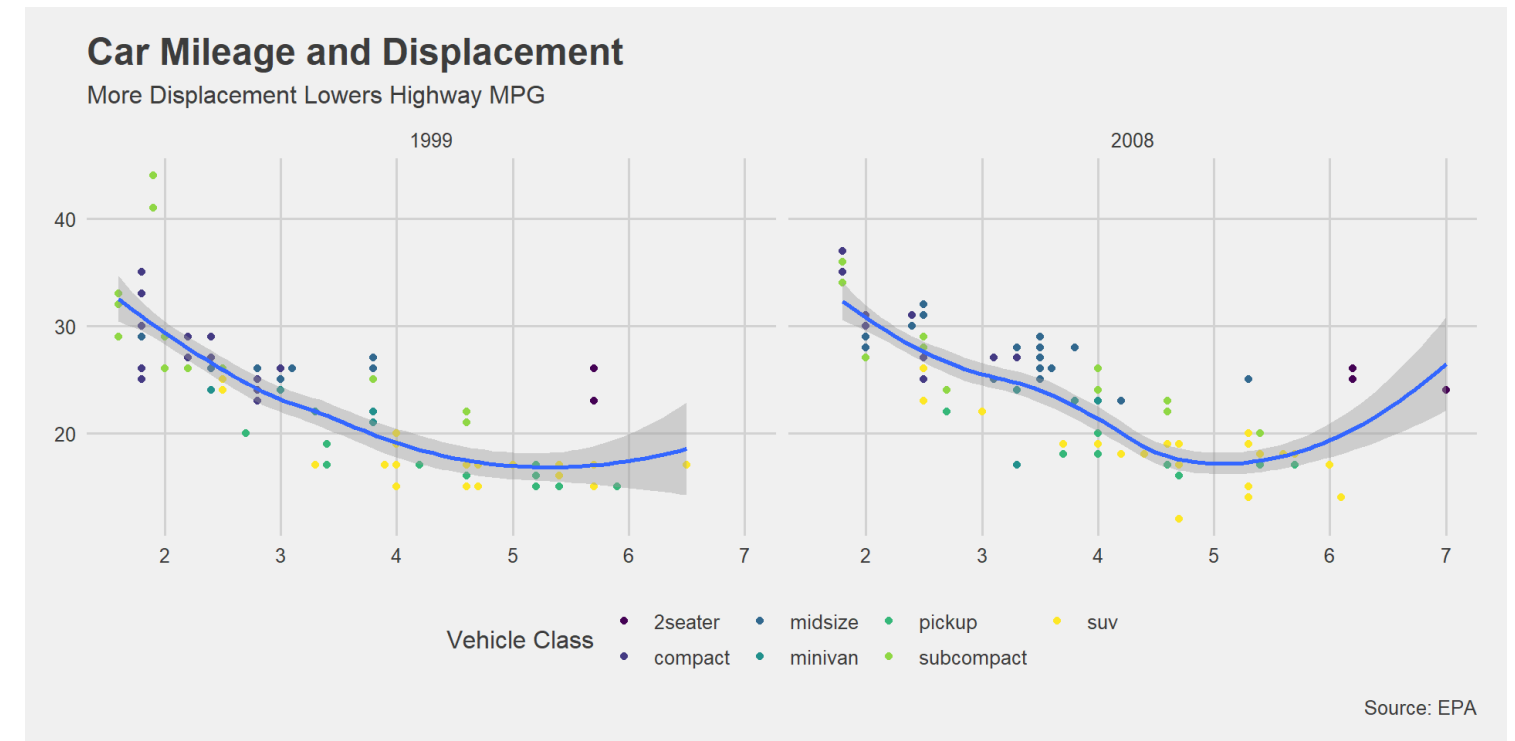
# gg: Themes VII

```
1 library(ggthemes)
2 ggplot(data = mpg)+
3   aes(x = displ,
4       y = hwy)+
5   geom_point(aes(color = class))+
6   geom_smooth()+
7   facet_wrap(~year)+
8   labs(x = "Engine Displacement (Liters)",
9        y = "Highway MPG",
10        title = "Car Mileage and Displacement",
11        subtitle = "More Displacement Lowers Highway",
12        caption = "Source: EPA",
13        color = "Vehicle Class")+
14   scale_color_viridis_d()+
15   theme_economist()+
16   theme(text = element_text(family = "Fira Sans"))
```



# gg: Themes VIII

```
1 library(ggthemes)
2 ggplot(data = mpg)+
3   aes(x = displ,
4       y = hwy)+
5   geom_point(aes(color = class))+
6   geom_smooth()+
7   facet_wrap(~year)+
8   labs(x = "Engine Displacement (Liters)",
9        y = "Highway MPG",
10        title = "Car Mileage and Displacement",
11        subtitle = "More Displacement Lowers Highway MPG",
12        caption = "Source: EPA",
13        color = "Vehicle Class")+
14   scale_color_viridis_d()+
15   theme_fivethirtyeight()+
16   theme(text = element_text(family = "Fira Sans"))
```



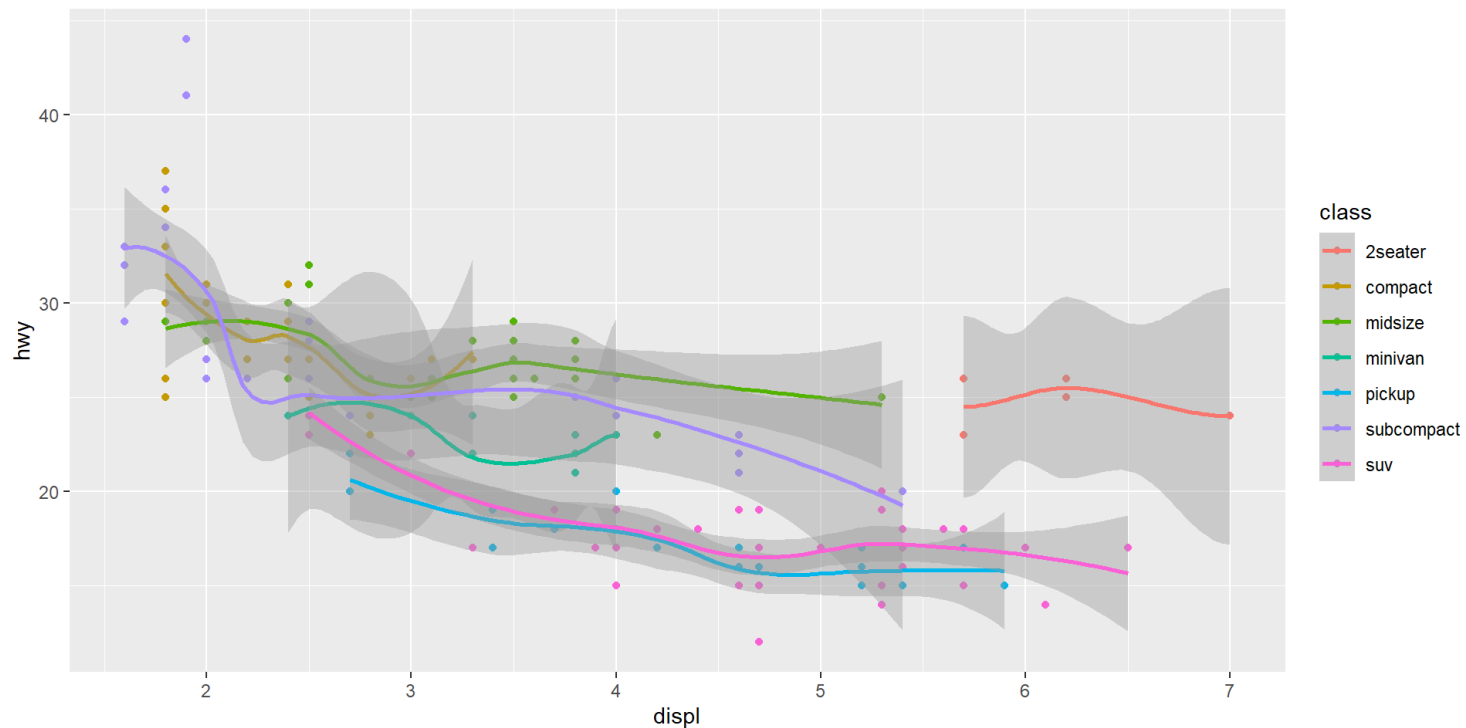


# Some Troubleshooting

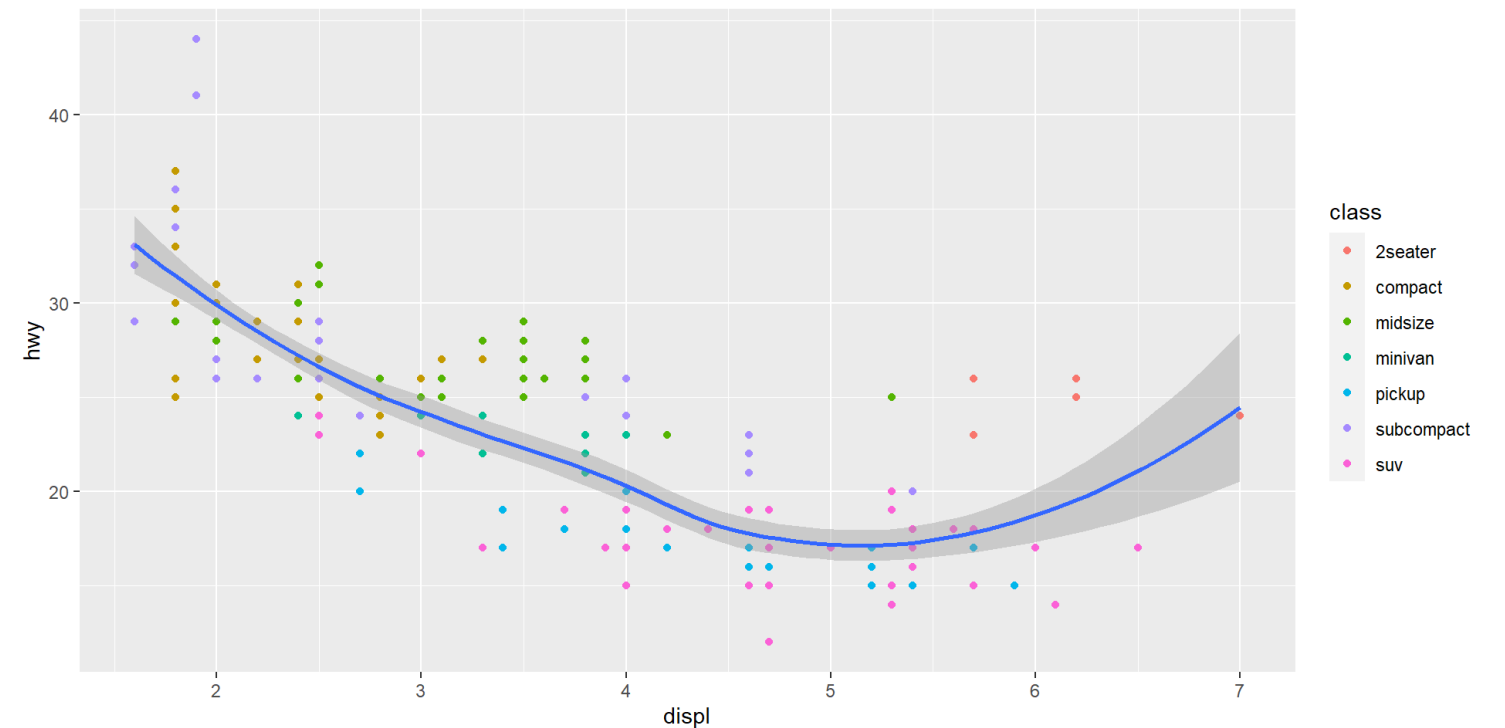
# Global vs. Local Aesthetic Mappings

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

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



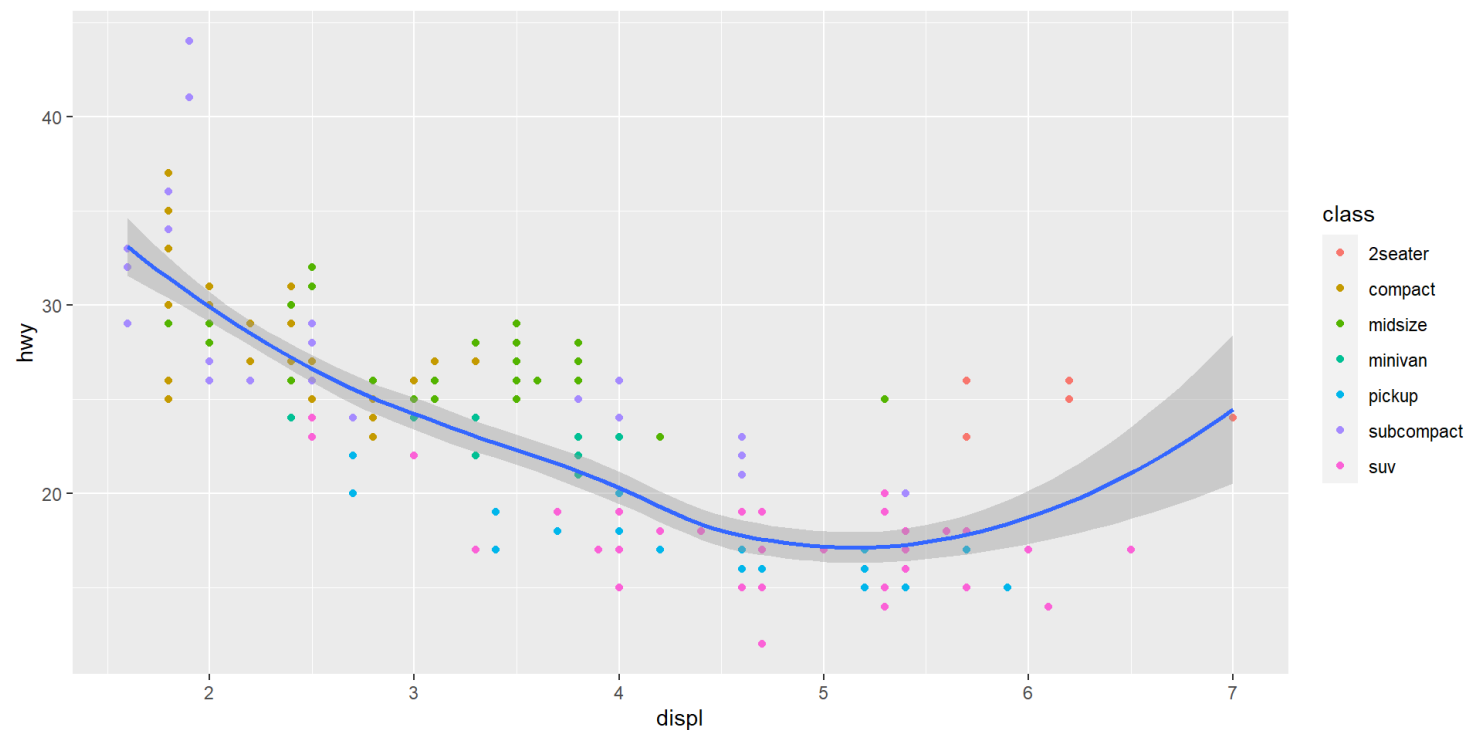
```
1 # ONLY points will map data to colors
2 ggplot(data = mpg, aes(x = displ,
3                         y = hwy)) +
4   geom_point(aes(color = class)) +
5   geom_smooth()
```



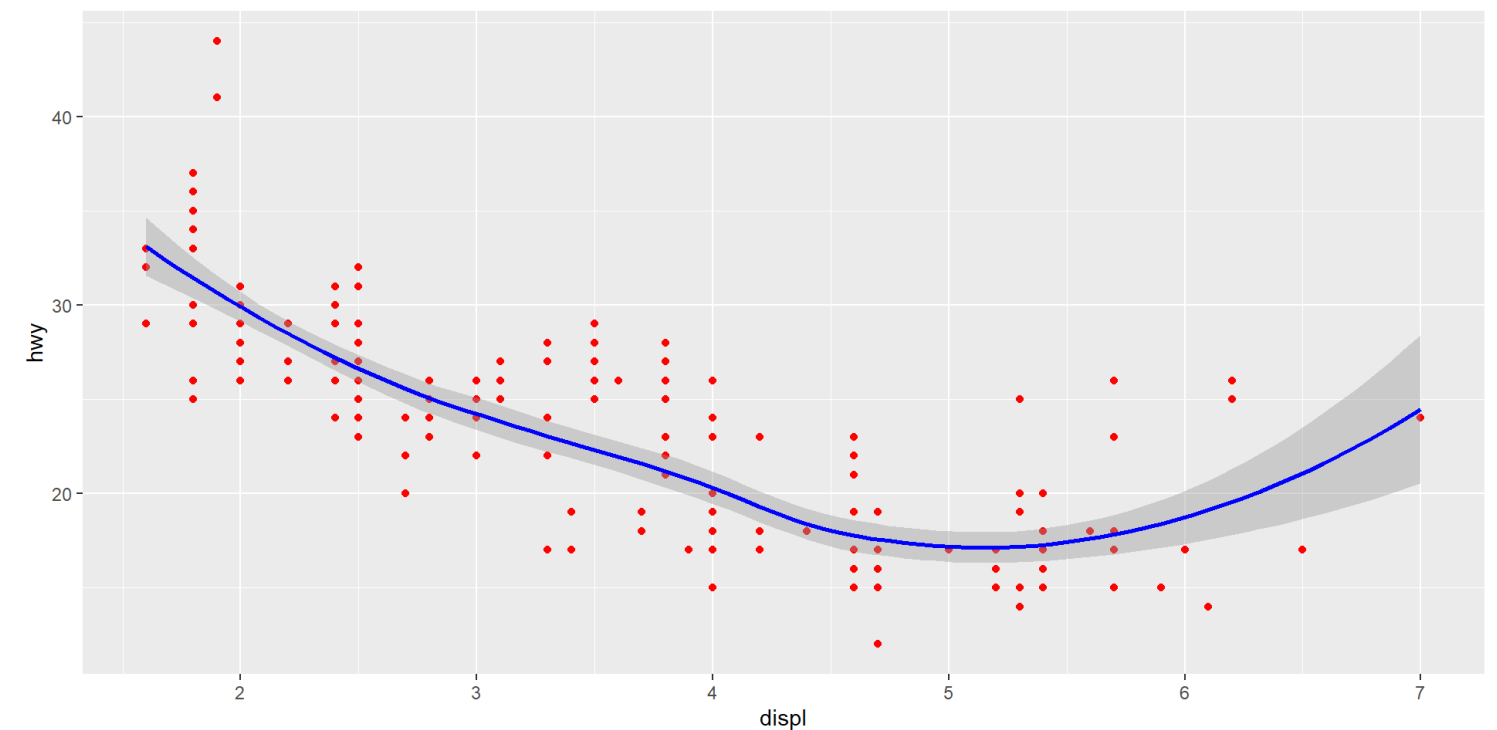
# Mapped vs. Set Aesthetics

- 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()`

```
1 # Point colors are mapped from class data
2 ggplot(data = mpg, aes(x = displ,
3                         y = hwy)) +
4   geom_point(aes(color = class)) +
5   geom_smooth()
```



```
1 # Point colors are all set to blue
2 ggplot(data = mpg, aes(x = displ,
3                         y = hwy)) +
4   geom_point(aes(), color = "red") +
5   geom_smooth(aes(), color = "blue")
```

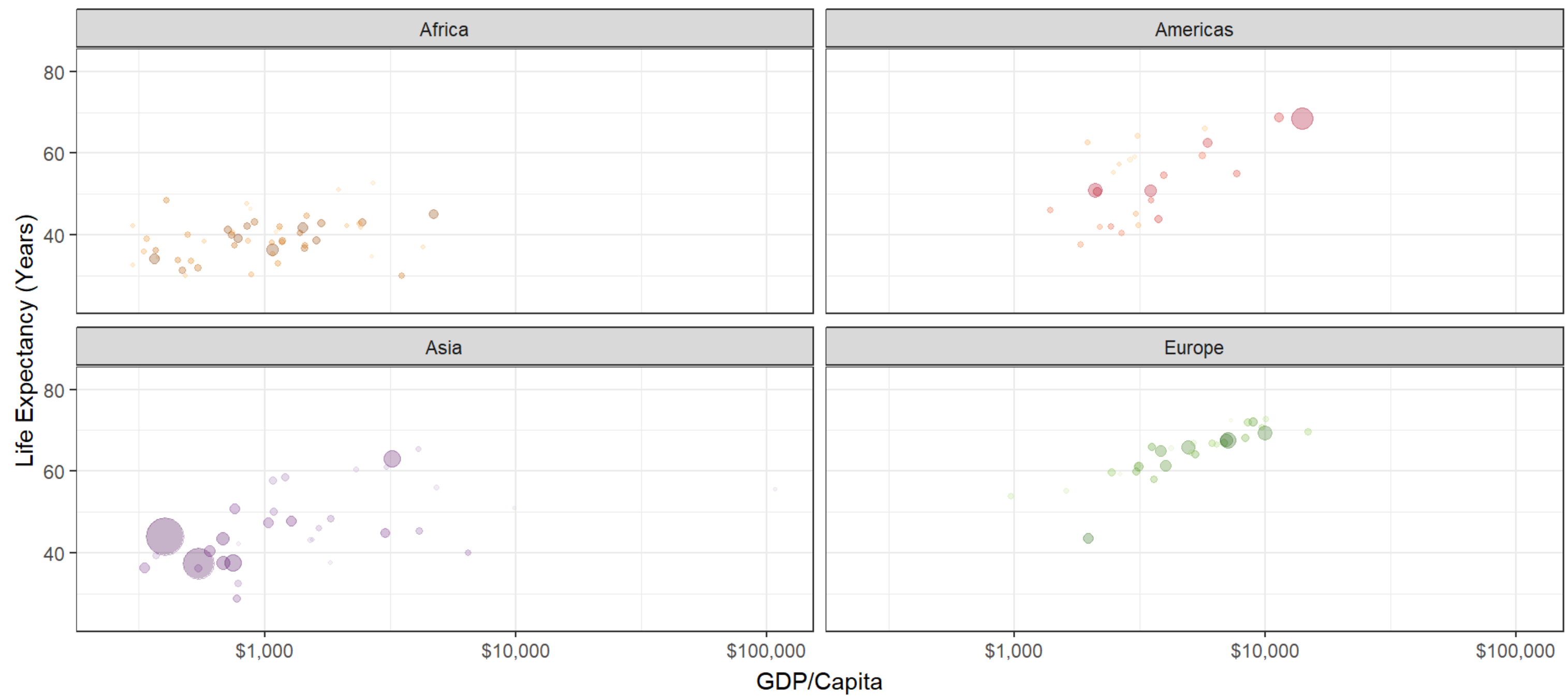


# Go Crazy II

Output	Code
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---

# Income & Life Expectancy - 1952



Source: Hans Rosling's gapminder.org

# 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 to. Add one geom function per layer.

**qplot**(x = cty, y = hwy, data = mpg, geom = "point") 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, unemployment))  
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, xend, y, yend, alpha, angle, color, curvature, linetype, size

**a + geom\_path**(lineend="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, lineheight, 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, lineheight, 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(cut, color))

**g + geom\_count**() x, y, alpha, color, fill, shape, size, stroke

#### THREE VARIABLES

seals\$z <- with(seals, sqrt(delta\_long^2 + delta\_lat^2)) l <- ggplot(seals, aes(long, lat))

##### l + geom\_contour(aes(z = z))

x, y, z, alpha, colour, group, linetype, size, weight

##### 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, colour, group, linetype, size

**h + geom\_hex**()  
x, y, alpha, colour, fill, size

##### continuous function

i <- ggplot(economics, aes(date, unemployment))

**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**(fatten = 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\_linerange**()  
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

**l + geom\_raster**(aes(fill = z), hjust=0.5, vjust=0.5, interpolate=FALSE)  
x, y, alpha, fill

**l + geom\_tile**(aes(fill = z)) x, y, alpha, color, fill, linetype, size, width



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# 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

