

Visualize data with ggplot2



Cars and displacement

What is the relationship between a car's engine weight (displacement) and its mileage (miles per gallon)?

Your turn #1

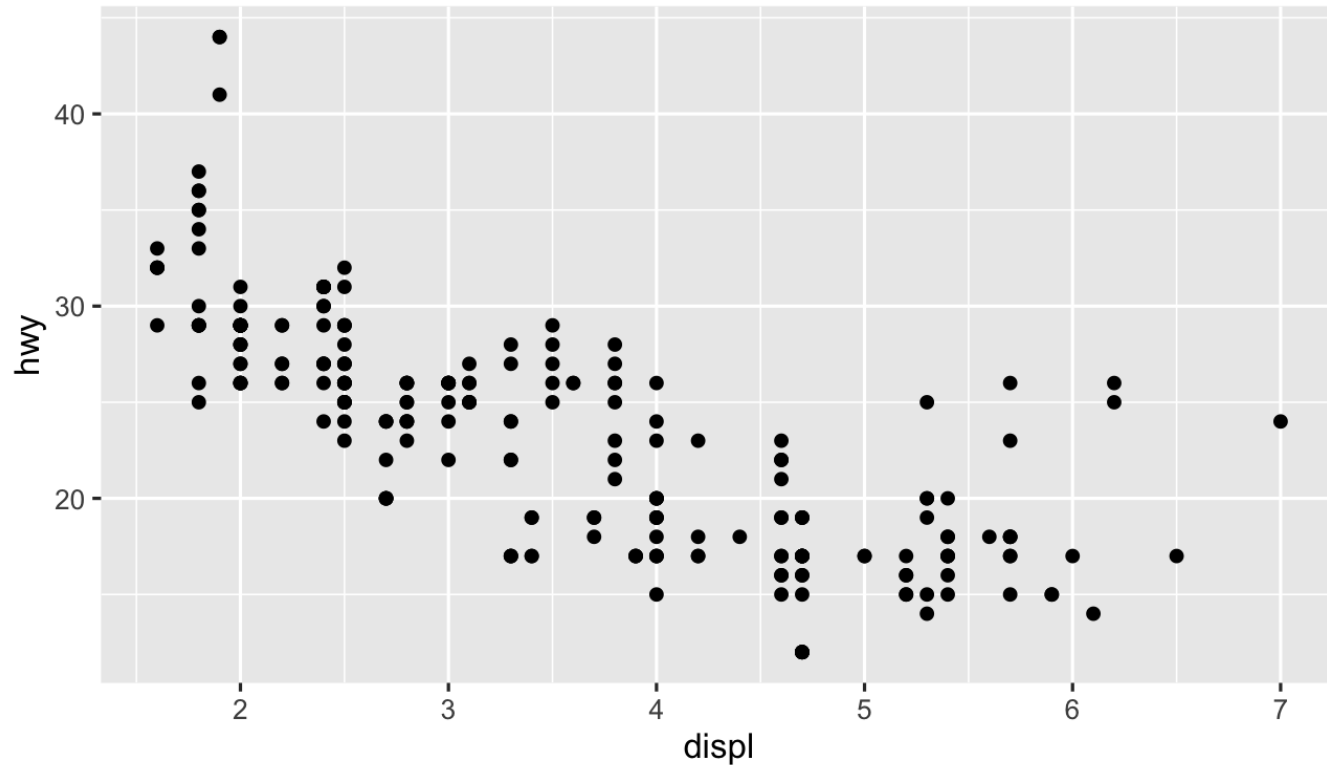
Run this code in your Quarto file to make a graph

Pay attention to spelling, capitalization, and parentheses!

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

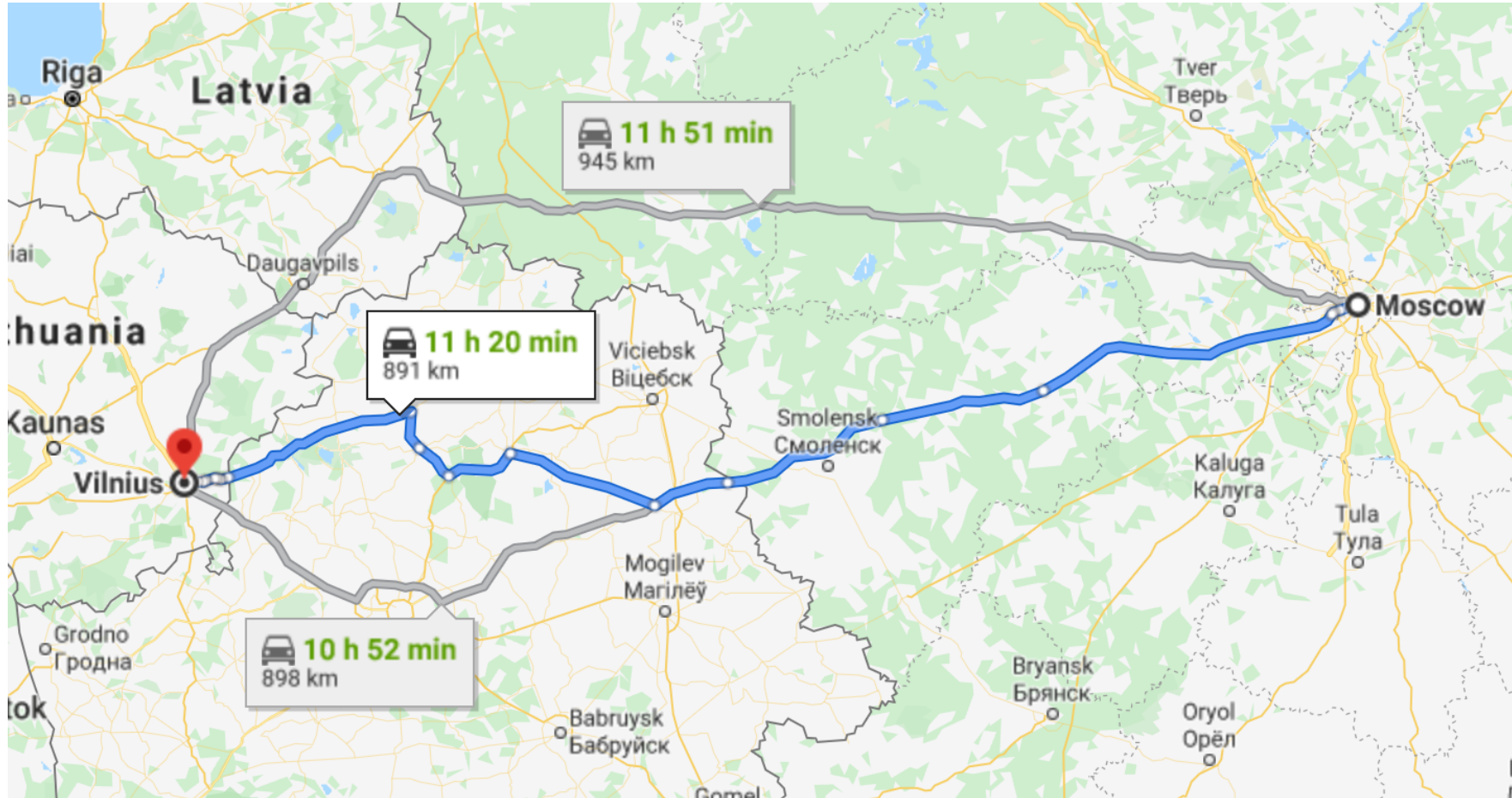
01:00

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



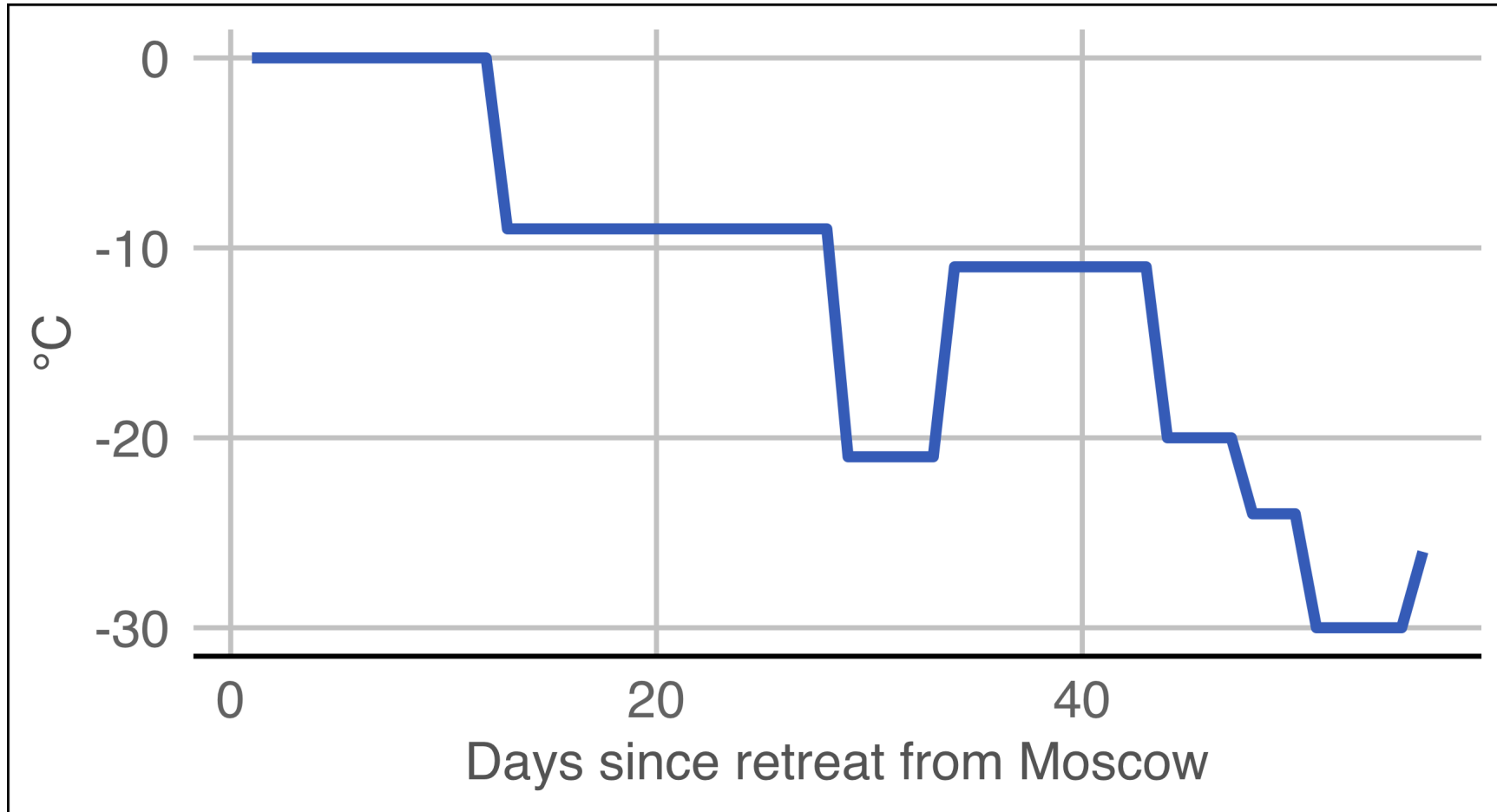


Long distance!



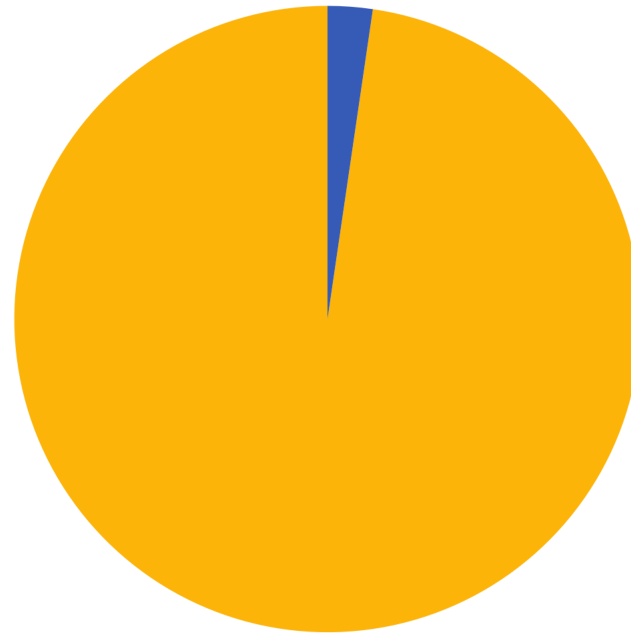
Moscow to Vilnius

Very cold!



Lots of people died!

Napoleon's Grande Armée



■ Died ■ Survived

Carte Figurative des pertes successives en hommes de l'Armée Française dans la campagne de Russie 1812-1813.

Dressée par M. Minard, Inspecteur Général des Ponts et Chaussées en retraite. Paris, le 20 Novembre 1869.

Les nombres d'hommes présents sont représentés par les largeurs des zones colorées à raison d'un millimètre pour dix mille hommes; ils sont de plus écrits en travers des zones. Le rouge désigne les hommes qui entrent en Russie, le noir ceux qui en sortent. — Les renseignements qui ont servi à dresser la carte ont été puisés dans les ouvrages de M. M. Chiers, de Ségur, de Fezensac, de Chambray et le journal inédit de Jacob, pharmacien de l'Armée depuis le 28 Octobre.

Pour mieux faire juger à l'œil la diminution de l'armée, j'ai supposé que les corps du Prince Jérôme et du Maréchal Davout qui avaient été détachés sur Minsk et Mohilow et ont rejoint vers Orscha et Witebsk, avaient toujours marché avec l'armée.

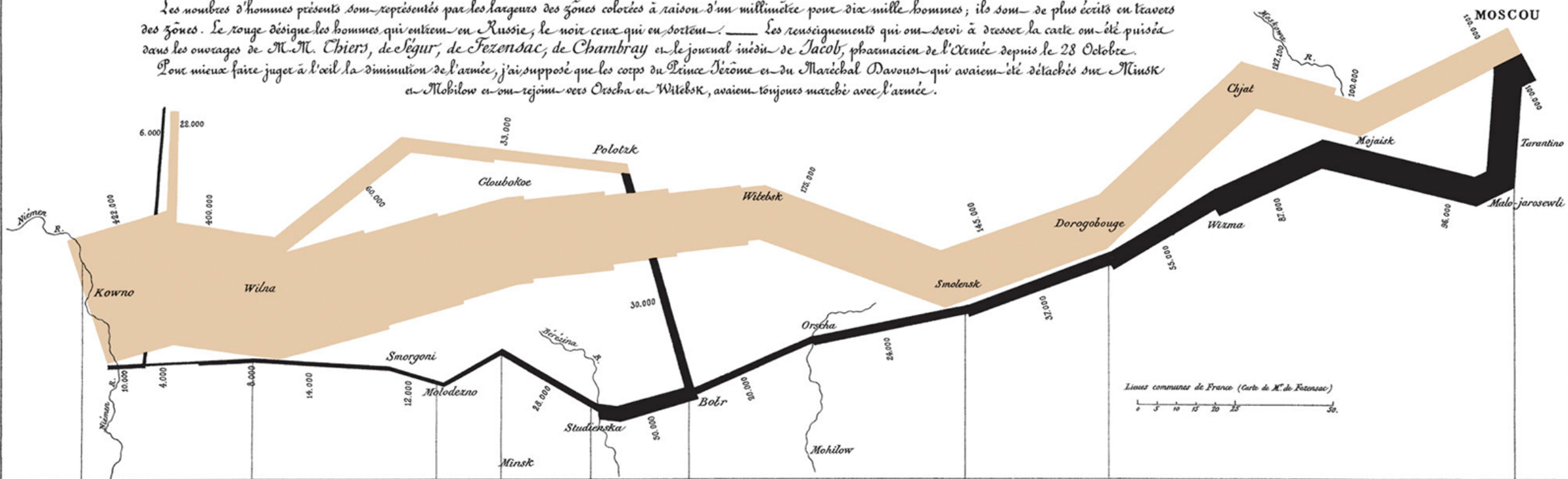
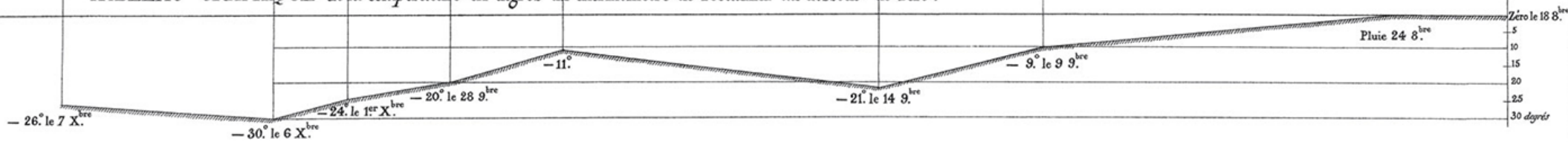


TABLEAU GRAPHIQUE de la température en degrés du thermomètre de Réaumur au dessous de zéro.

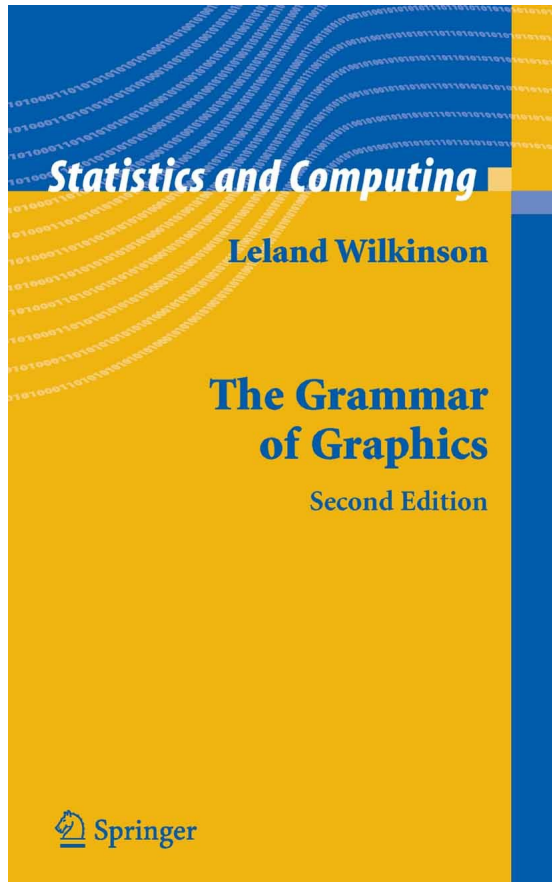


Les Cosaques passent au galop le Niemen gelé.

Autog. par Regnier, 8, Pav. S^{te} Marie S^t G^{ermain} à Paris.

Imp. Lith. Regnier et Douvret.

Mapping data to aesthetics



Aesthetic

Visual property of a graph

Position, shape, color, etc.

Data

A column in a dataset

Mapping data to aesthetics

Data	Aesthetic	Graphic/Geometry
Longitude	Position (x-axis)	Point
Latitude	Position (y-axis)	Point
Army size	Size	Path
Army direction	Color	Path
Date	Position (x-axis)	Line + text
Temperature	Position (y-axis)	Line + text

Mapping data to aesthetics

Data	aes()	geom
Longitude	x	geom_point()
Latitude	y	geom_point()
Army size	size	geom_path()
Army direction	color	geom_path()
Date	x	geom_line() + geom_text()
Temperature	y	geom_line() + geom_text()

ggplot() template

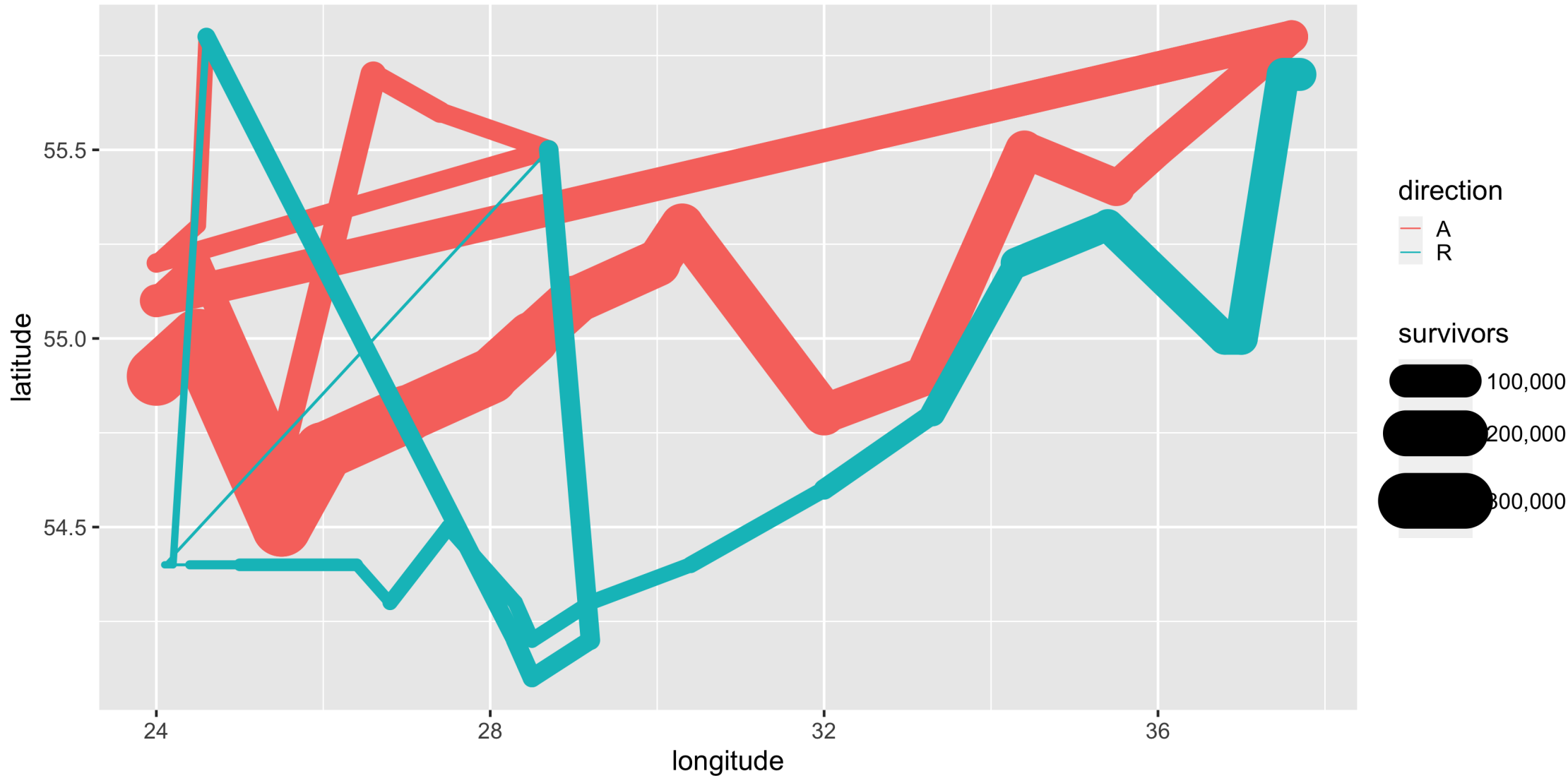
```
ggplot(data = DATA) +  
  GEOM_FUNCTION(mapping = aes(AESTHETIC MAPPINGS))
```

```
ggplot(data = troops) +  
  geom_path(mapping = aes(x = longitude,  
                           y = latitude,  
                           color = direction,  
                           size = survivors))
```

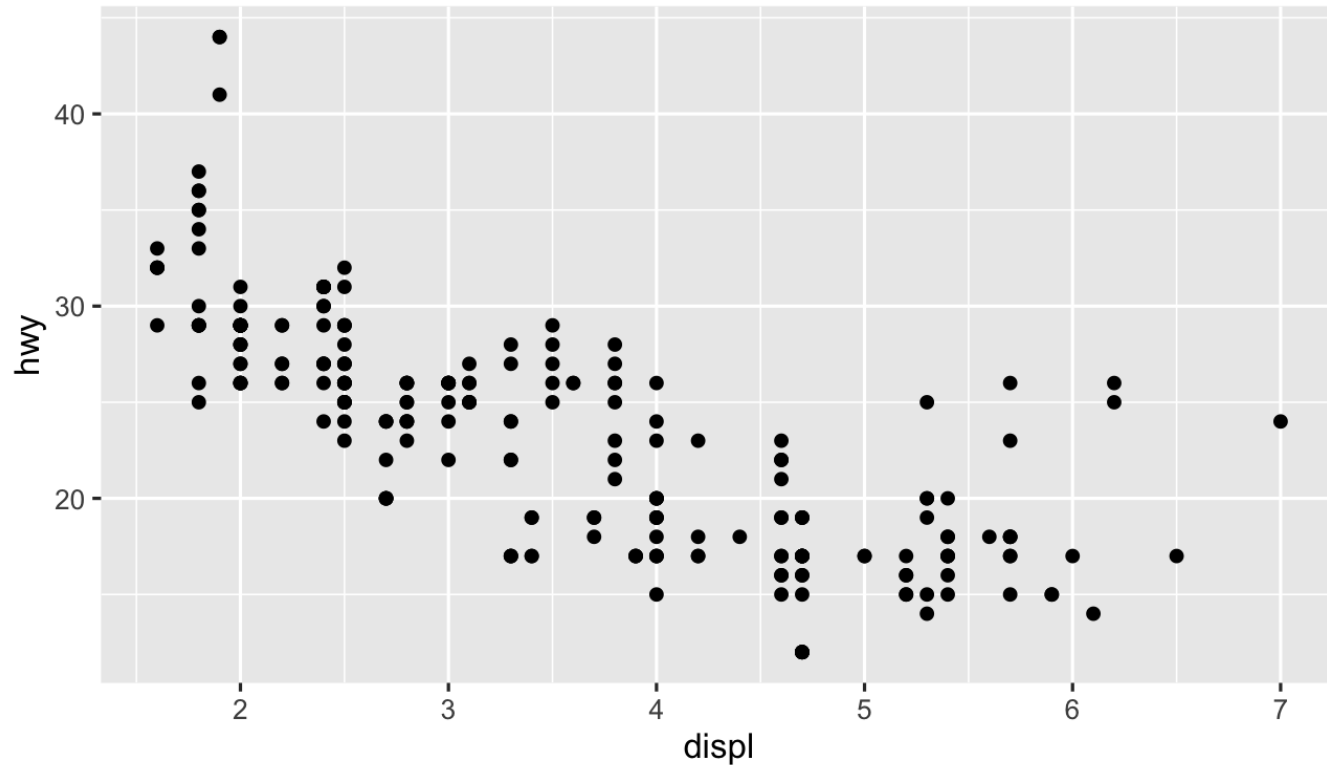
This is a dataset named `troops`:

longitude	latitude	direction	survivors
24	54.9	A	340000
24.5	55	A	340000
...

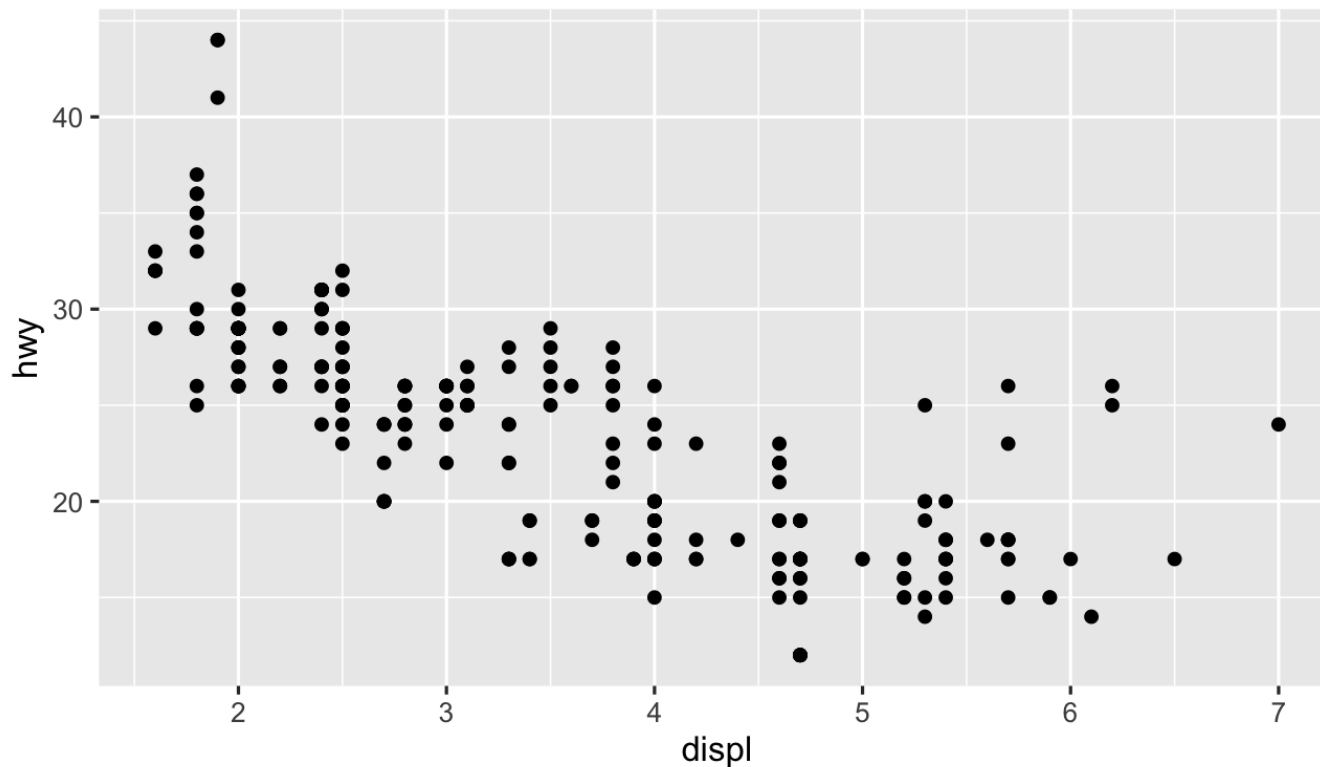
```
ggplot(data = troops) +  
  geom_path(mapping = aes(x = longitude,  
                           y = latitude,  
                           color = direction,  
                           size = survivors))
```



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

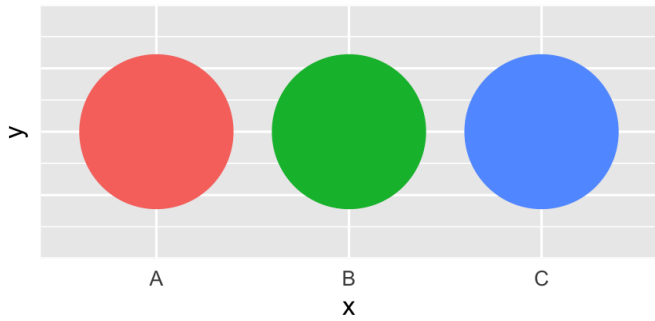


Heavy cars with better mileage?

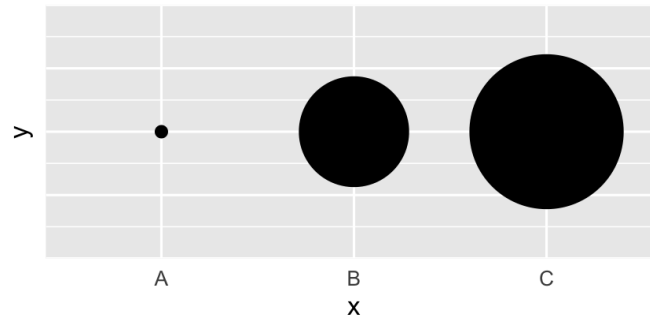


Aesthetics

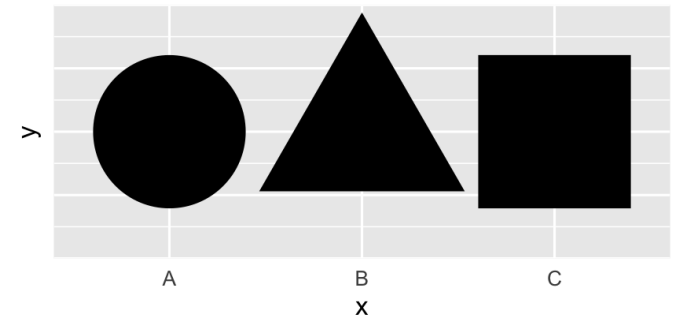
color (discrete)



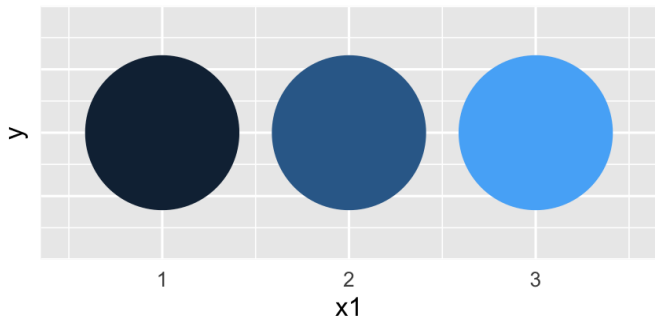
size



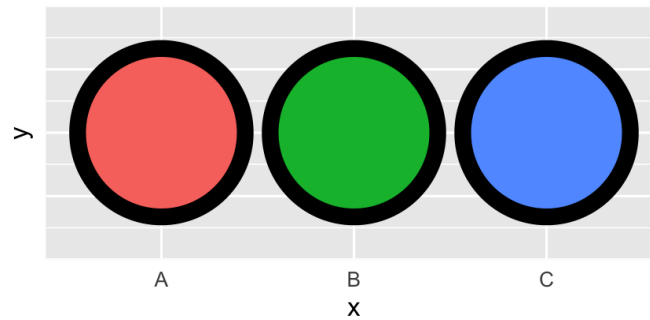
shape



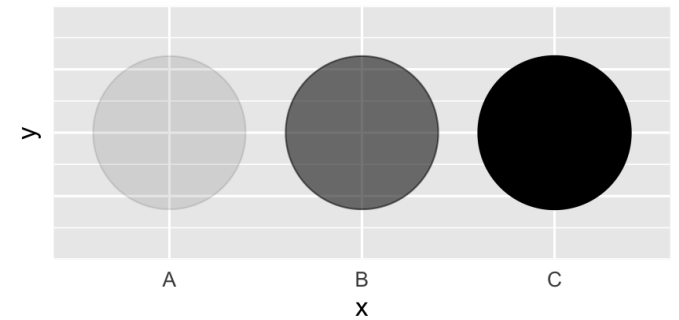
color (continuous)



fill



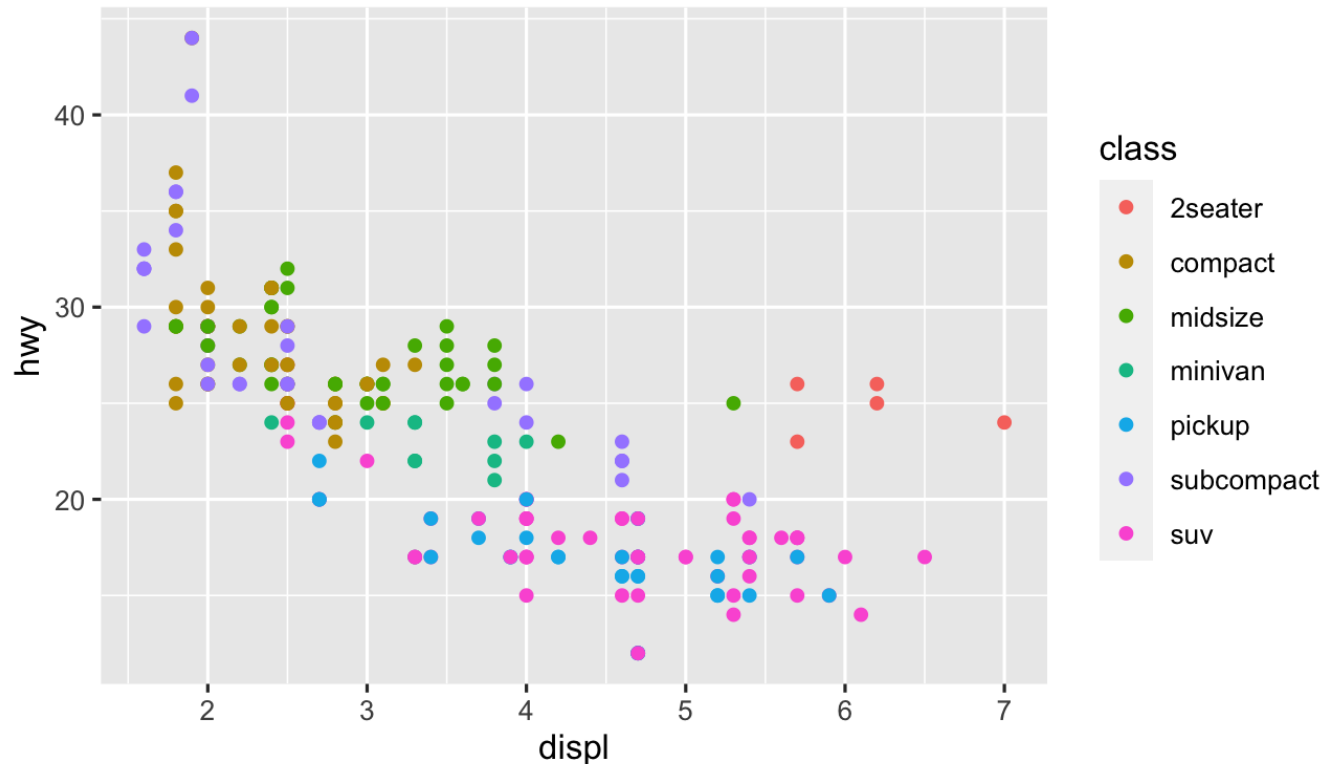
alpha



Mapping columns to aesthetics

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

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



Your turn #2

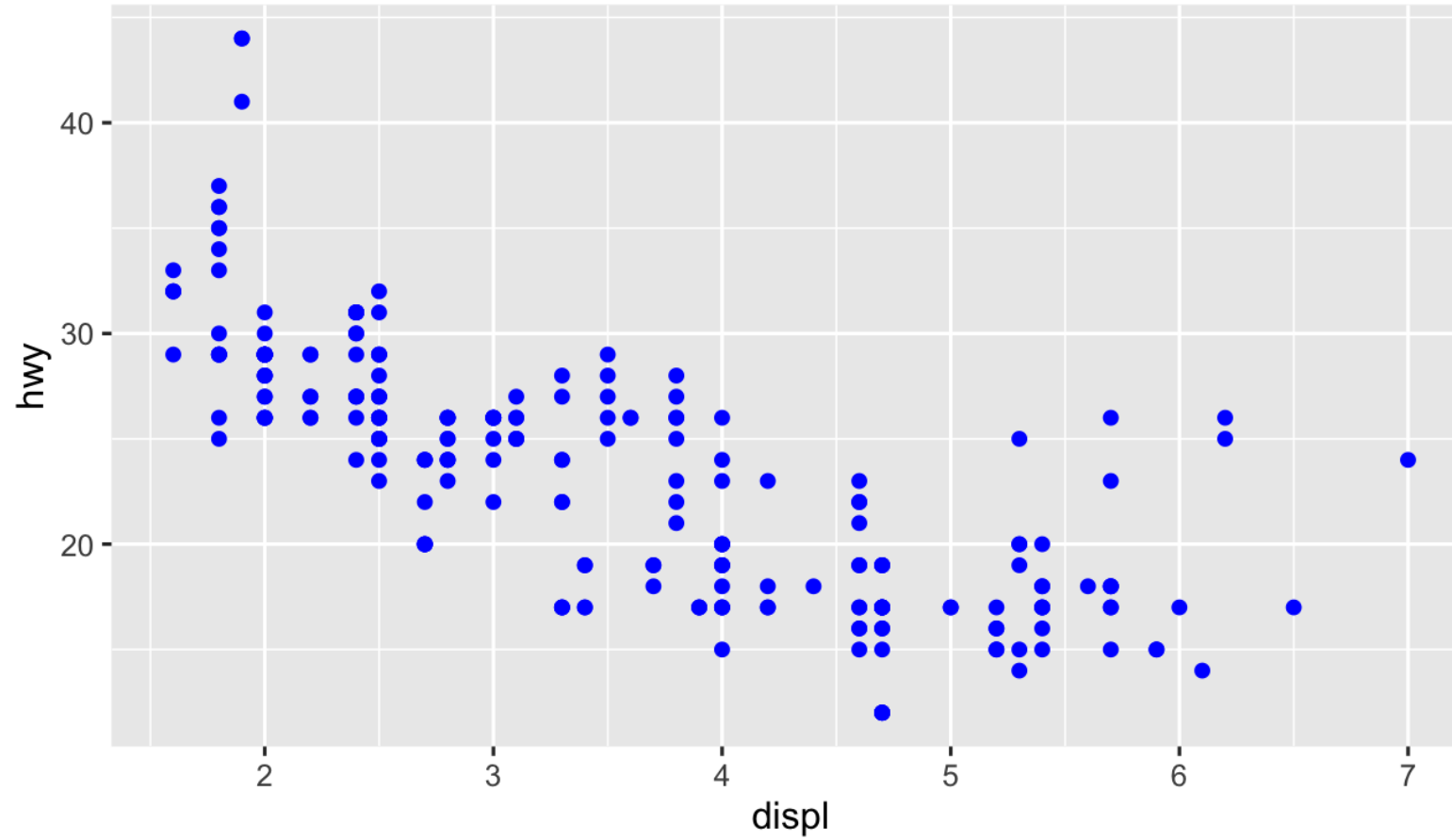
Add color, size, alpha, and shape aesthetics to your graph.

Experiment!

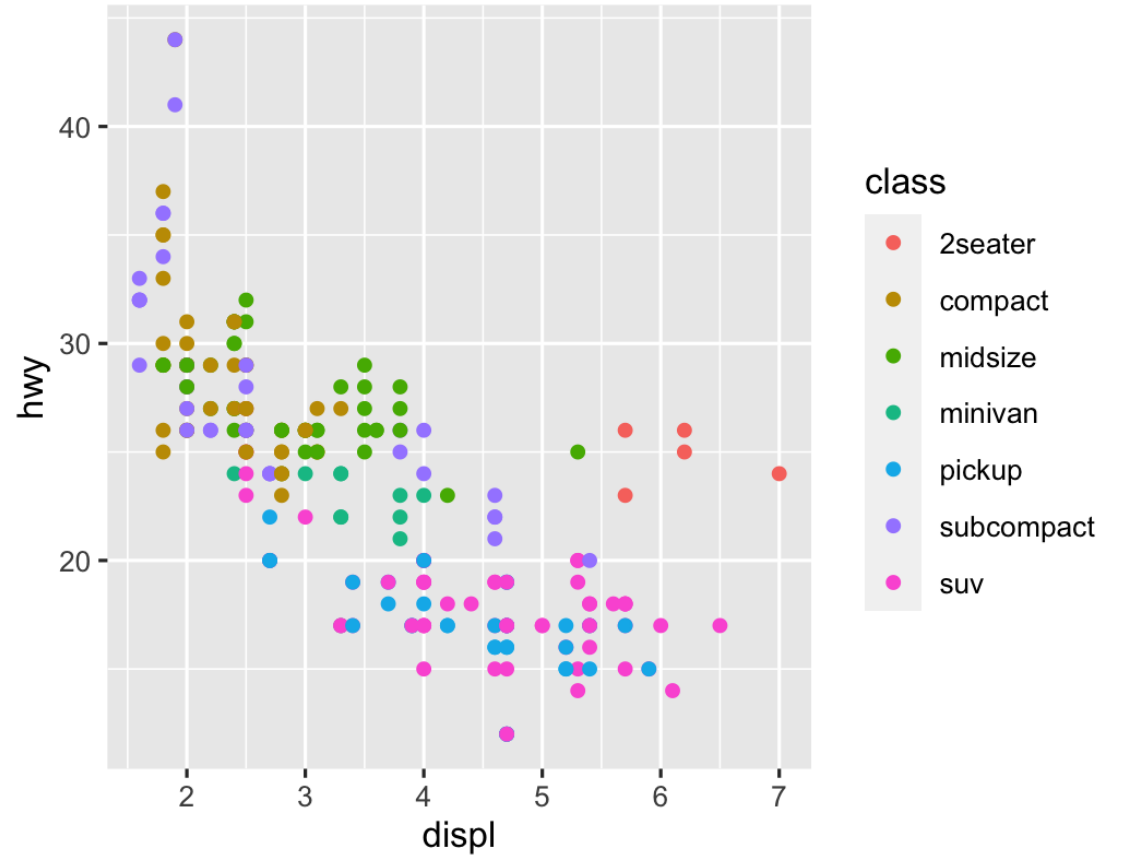
Do different things happen when you map aesthetics to discrete and continuous variables?

What happens when you use more than one aesthetic?

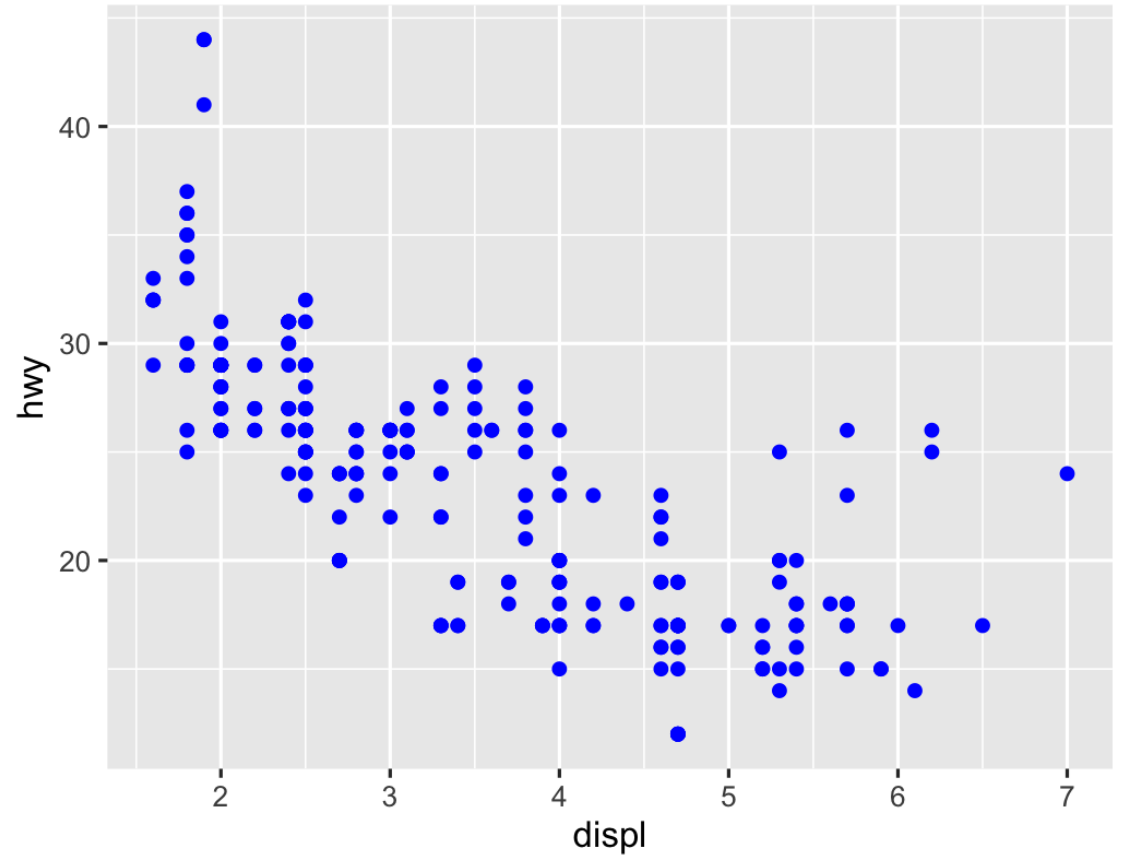
How would you make this plot?



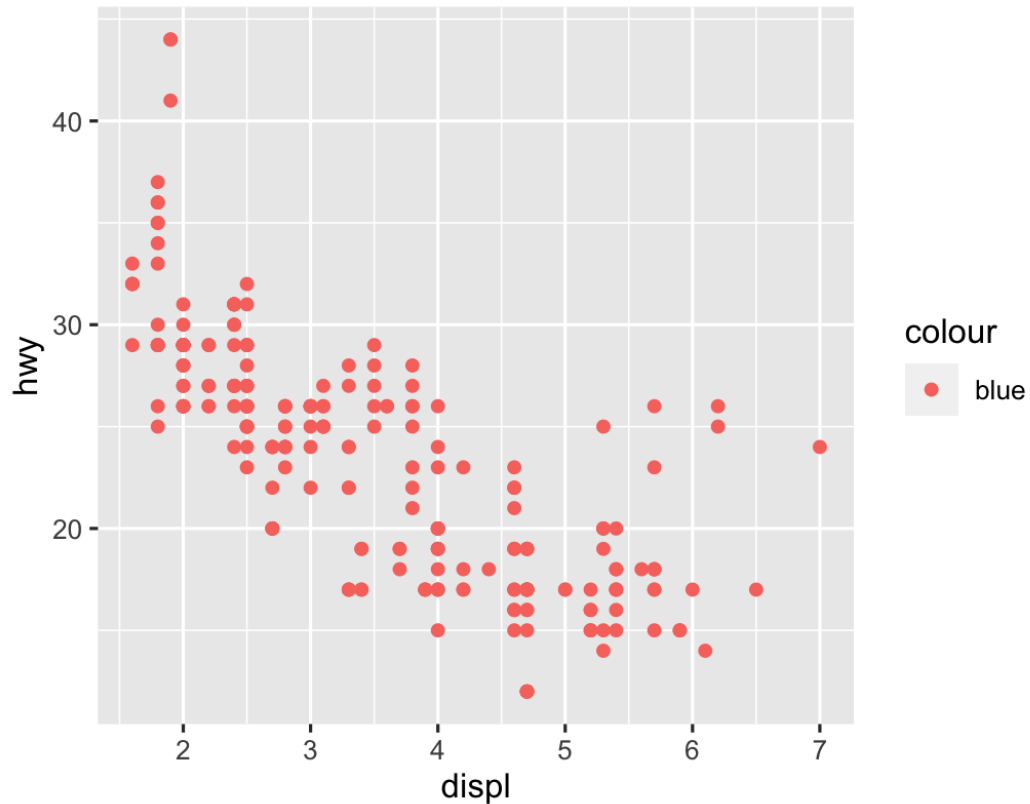
```
ggplot(mpg) +  
  geom_point(aes(x = displ, y = hwy,  
                 color = class))
```



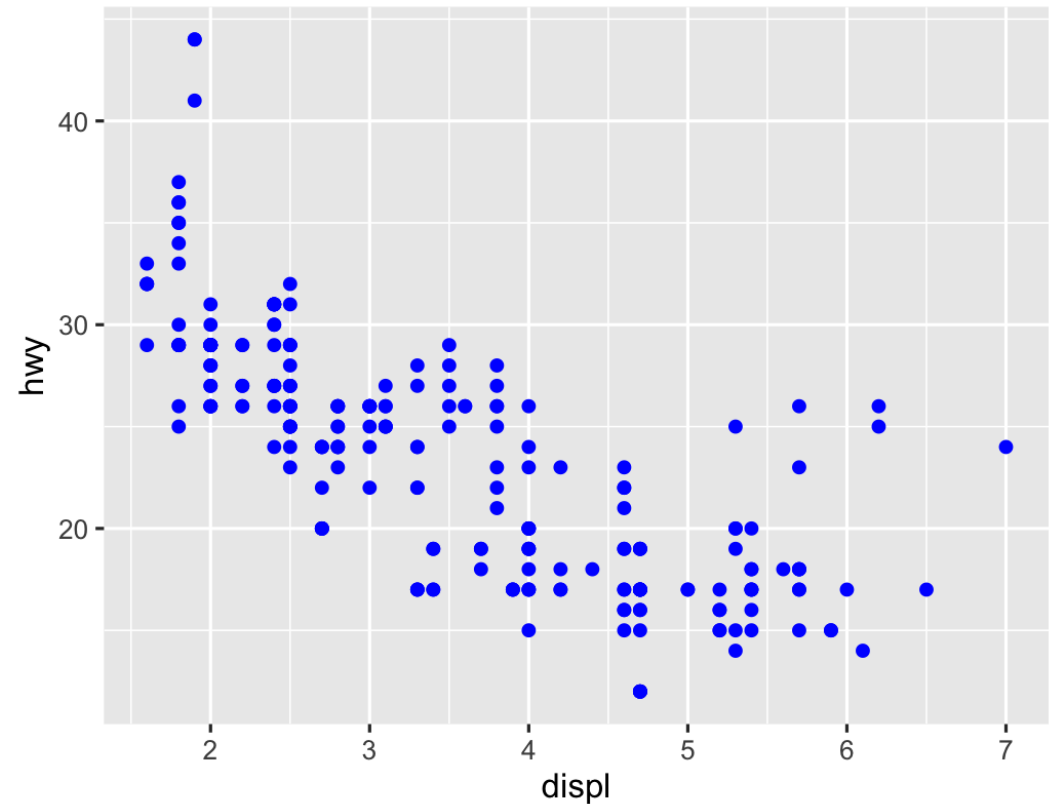
```
ggplot(mpg) +  
  geom_point(aes(x = displ, y = hwy),  
             color = "blue")
```



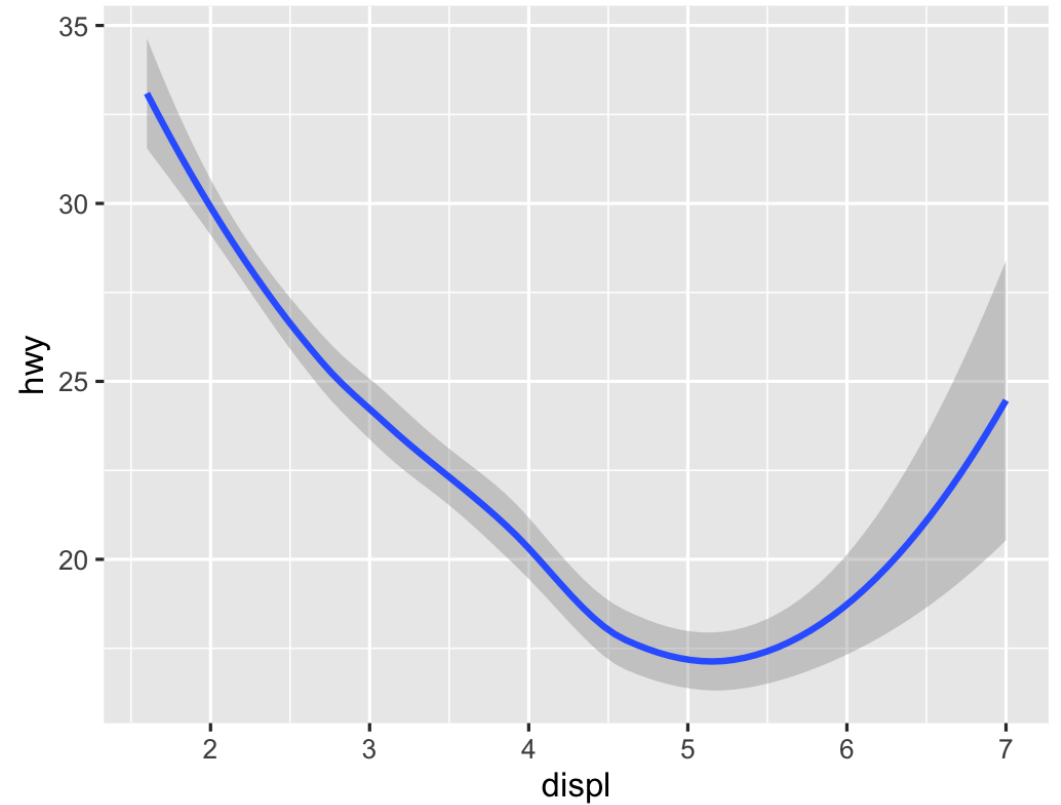
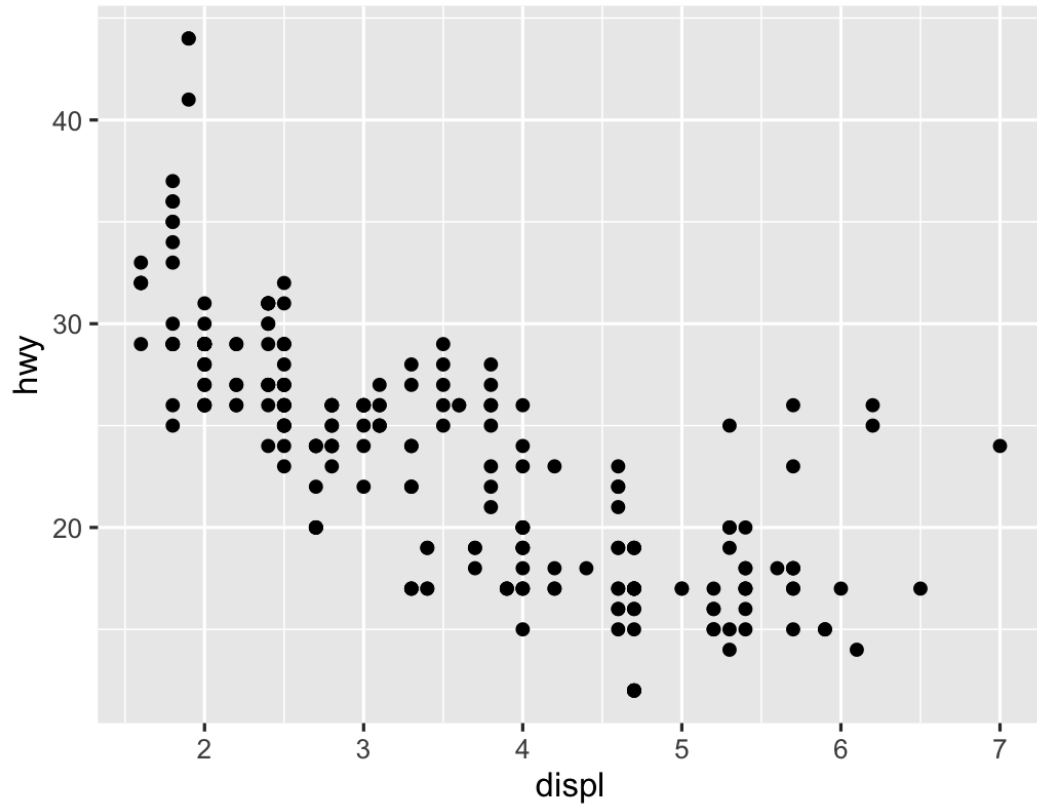
```
ggplot(mpg) +  
  geom_point(aes(x = displ, y = hwy,  
                 color = "blue"))
```



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







What's the same? What's different?



Geoms

```
ggplot(data = DATA) +  
  GEOM_FUNCTION(mapping = aes(AESTHETIC MAPPINGS))
```

Possible geoms

	Example geom	What it makes
	<code>geom_col()</code>	Bar charts
<i>text</i>	<code>geom_text()</code>	Text
	<code>geom_point()</code>	Points
	<code>geom_boxplot()</code>	Boxplots
	<code>geom_sf()</code>	Maps

Possible geoms

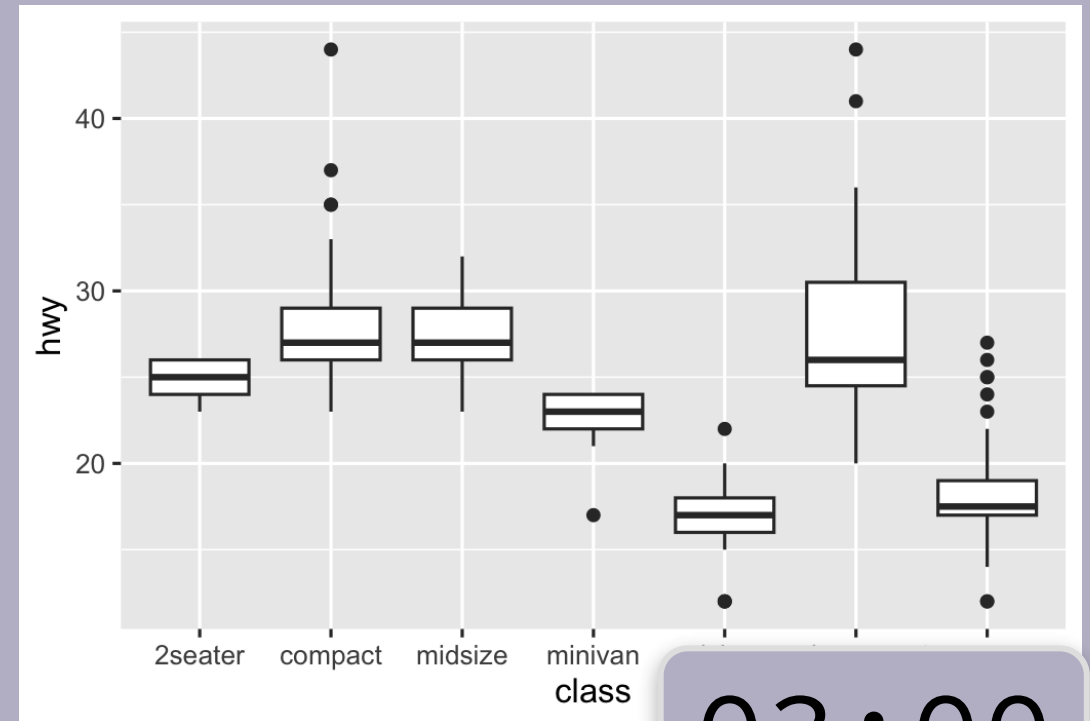
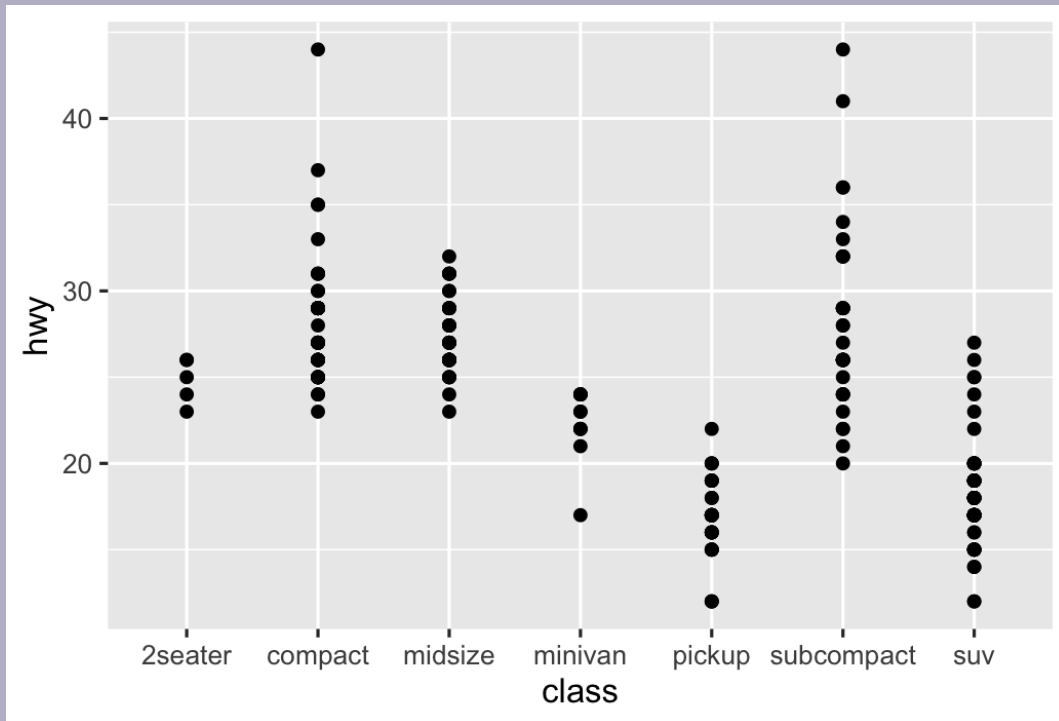
There are dozens of possible geoms!

See **the ggplot2 documentation** for complete examples of all the different geom layers

Also see the ggplot cheatsheet

Your turn #3

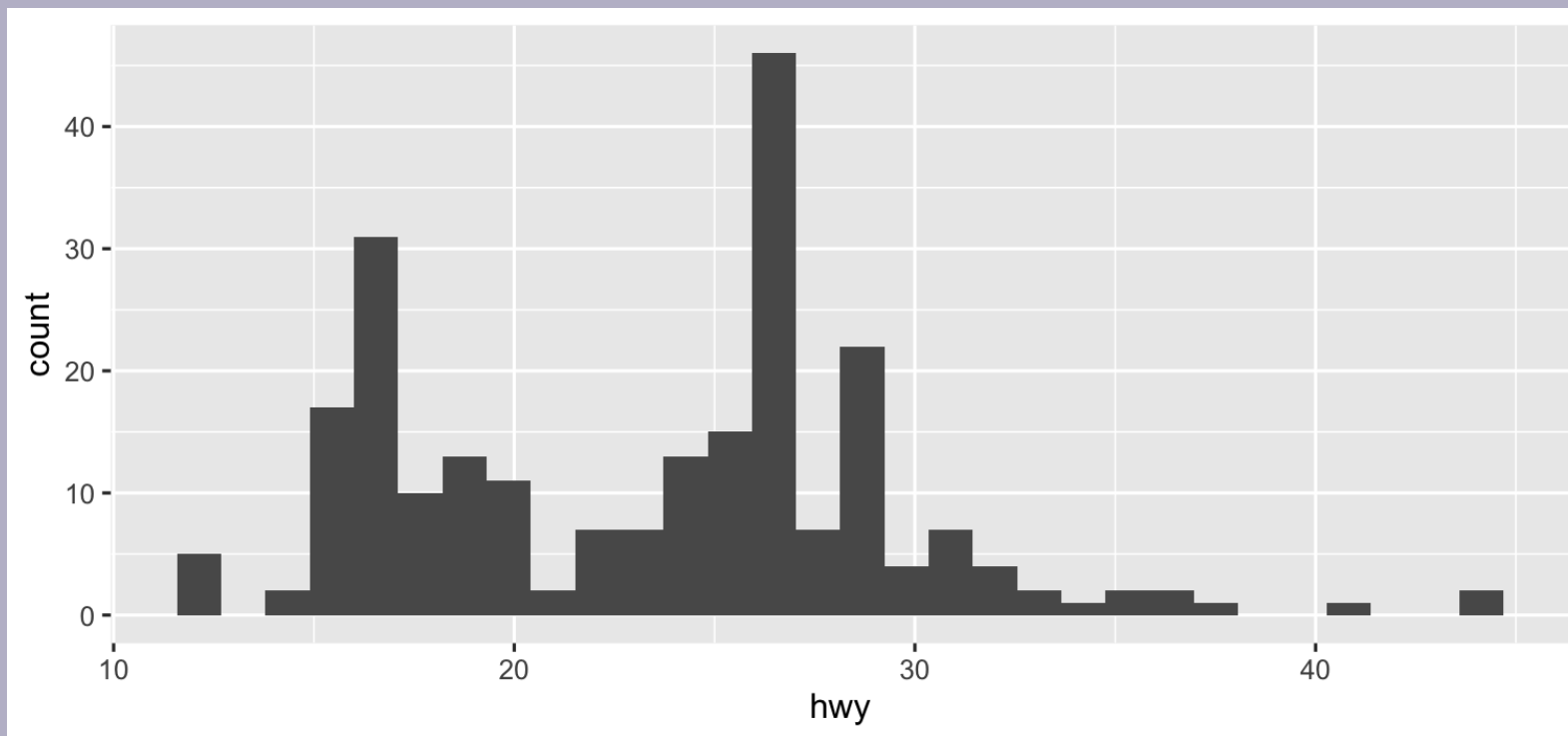
Replace this scatterplot with boxplots. Use the cheatsheet.



03:00

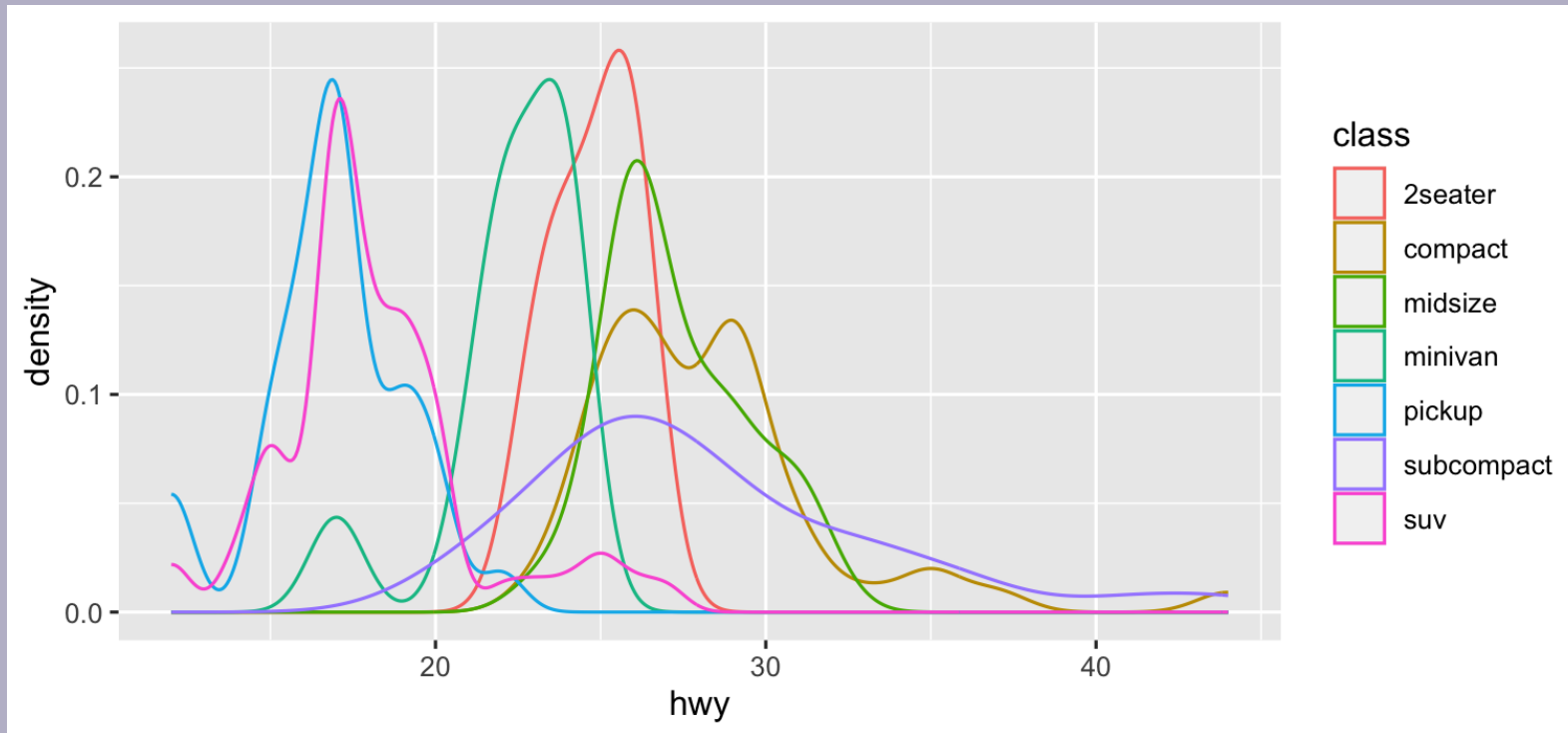
Your turn #4

**Make a histogram of `hwy`. Use the cheatsheet.
Hint: don't supply a `y` variable.**

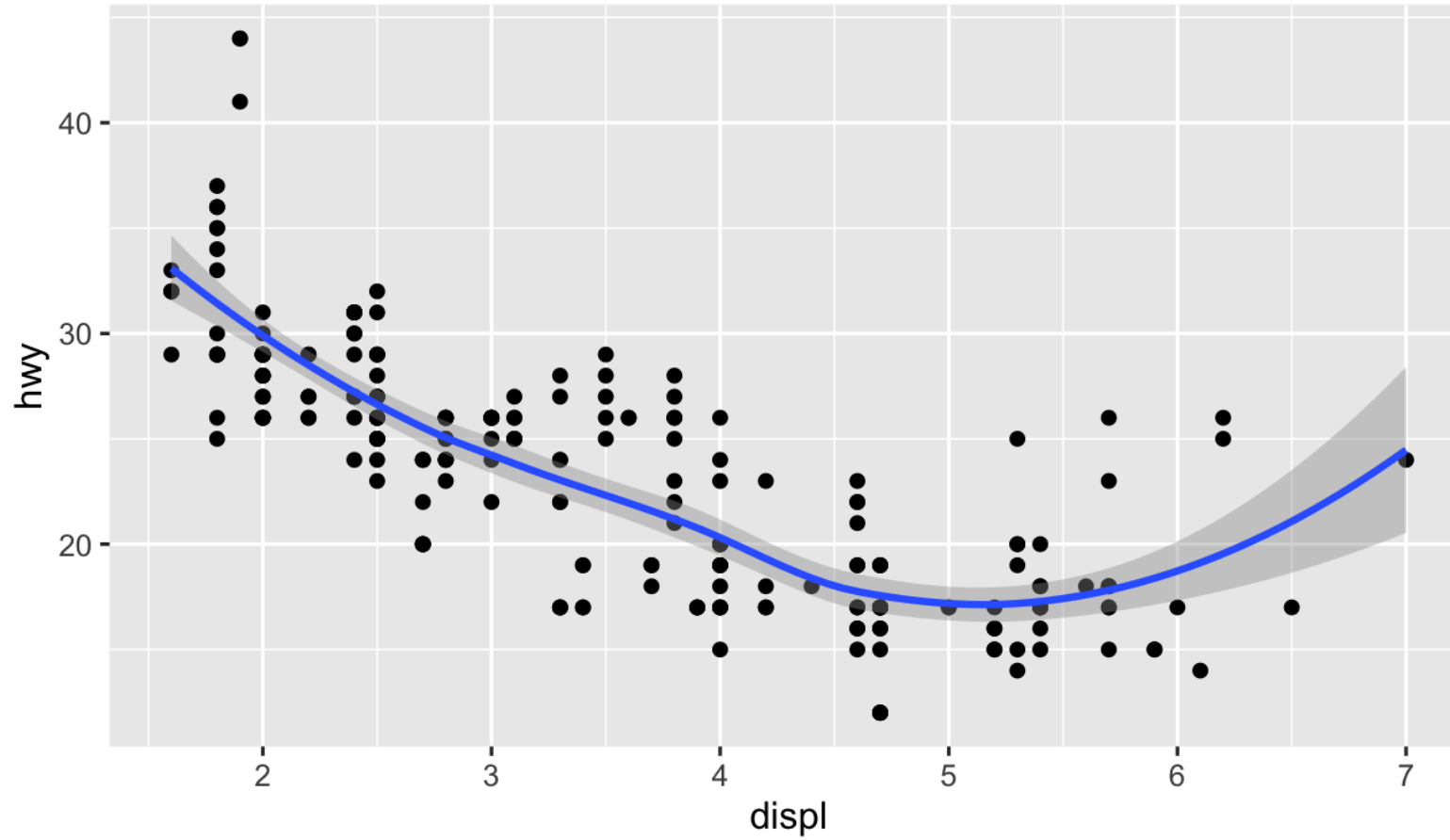


Your turn #5

Make this density plot of `hwy` colored by `class`.
Use the cheatsheet. Hint: don't supply a `y` variable.



Complex graphs!



Your turn #6

Predict what this code will do. Then run it.

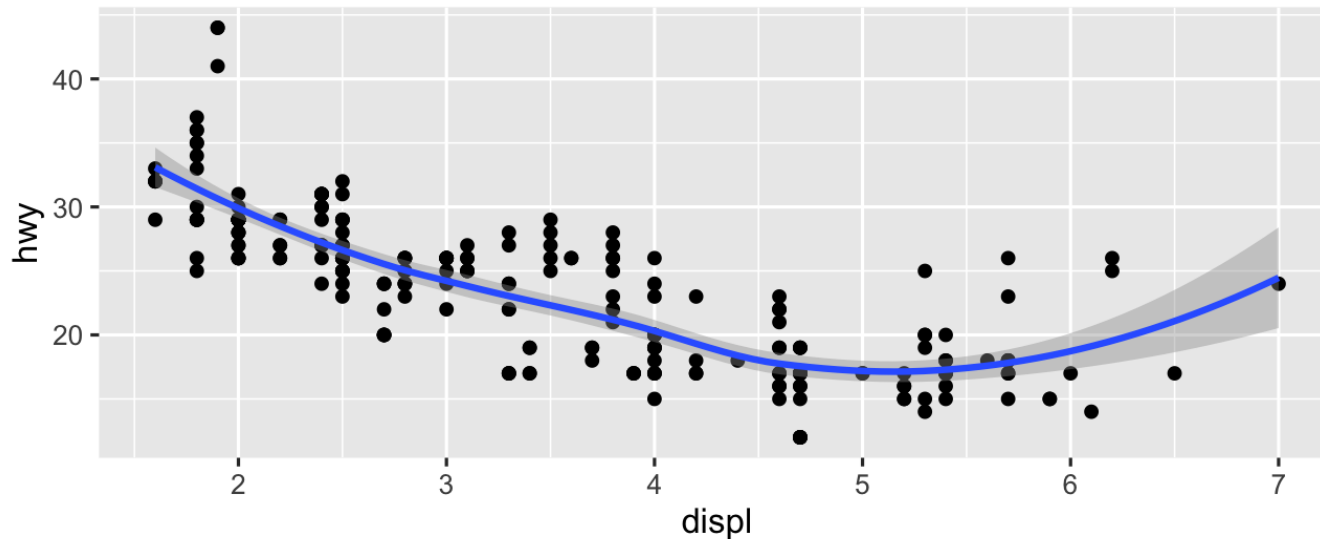
```
ggplot(data = mpg) +  
  geom_point(mapping = aes(x = displ, y = hwy)) +  
  geom_smooth(mapping = aes(x = displ, y = hwy))
```

02:00

Global vs. local

Any aesthetics in `ggplot()` will show up in all `geom_` layers

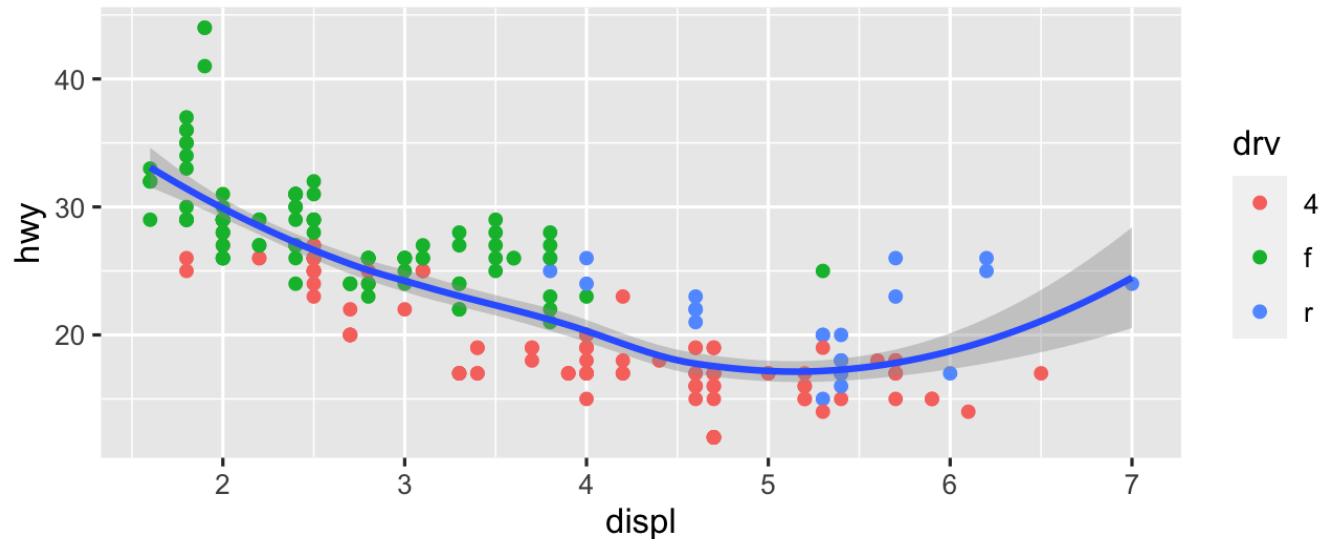
```
ggplot(data = mpg, mapping = aes(x = displ, y = hwy)) +  
  geom_point() +  
  geom_smooth()
```



Global vs. local

Any aesthetics in `geom_` layers only apply to that layer

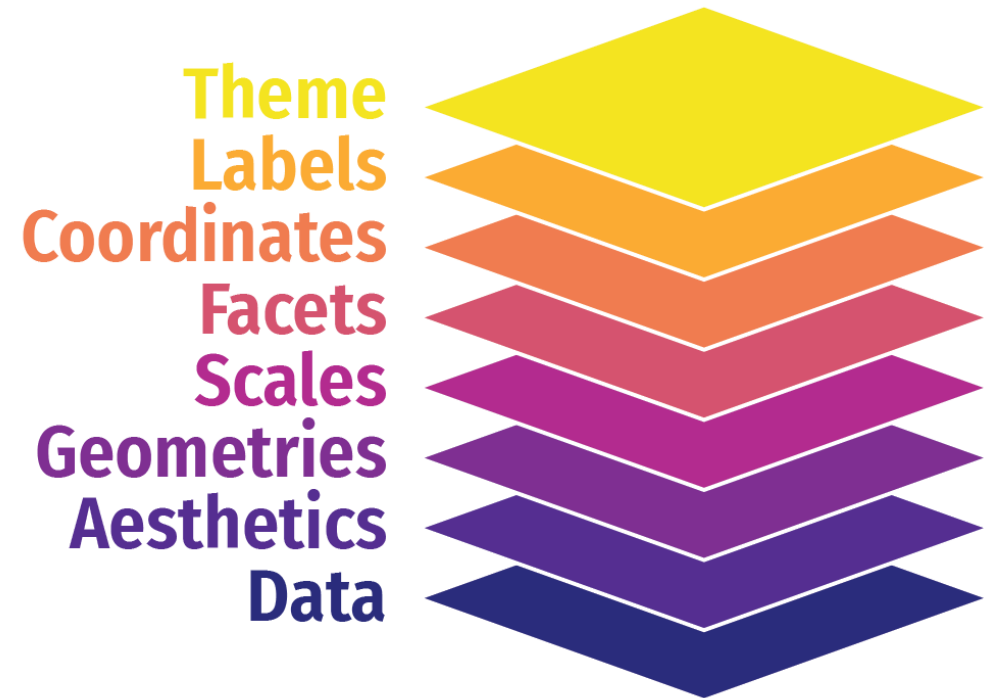
```
ggplot(data = mpg, mapping = aes(x = displ, y = hwy)) +  
  geom_point(mapping = aes(color = drv)) +  
  geom_smooth()
```



So much more!

There are many other layers we can use to make and enhance graphs!

We sequentially add layers onto the foundational `ggplot()` plot to create complex figures

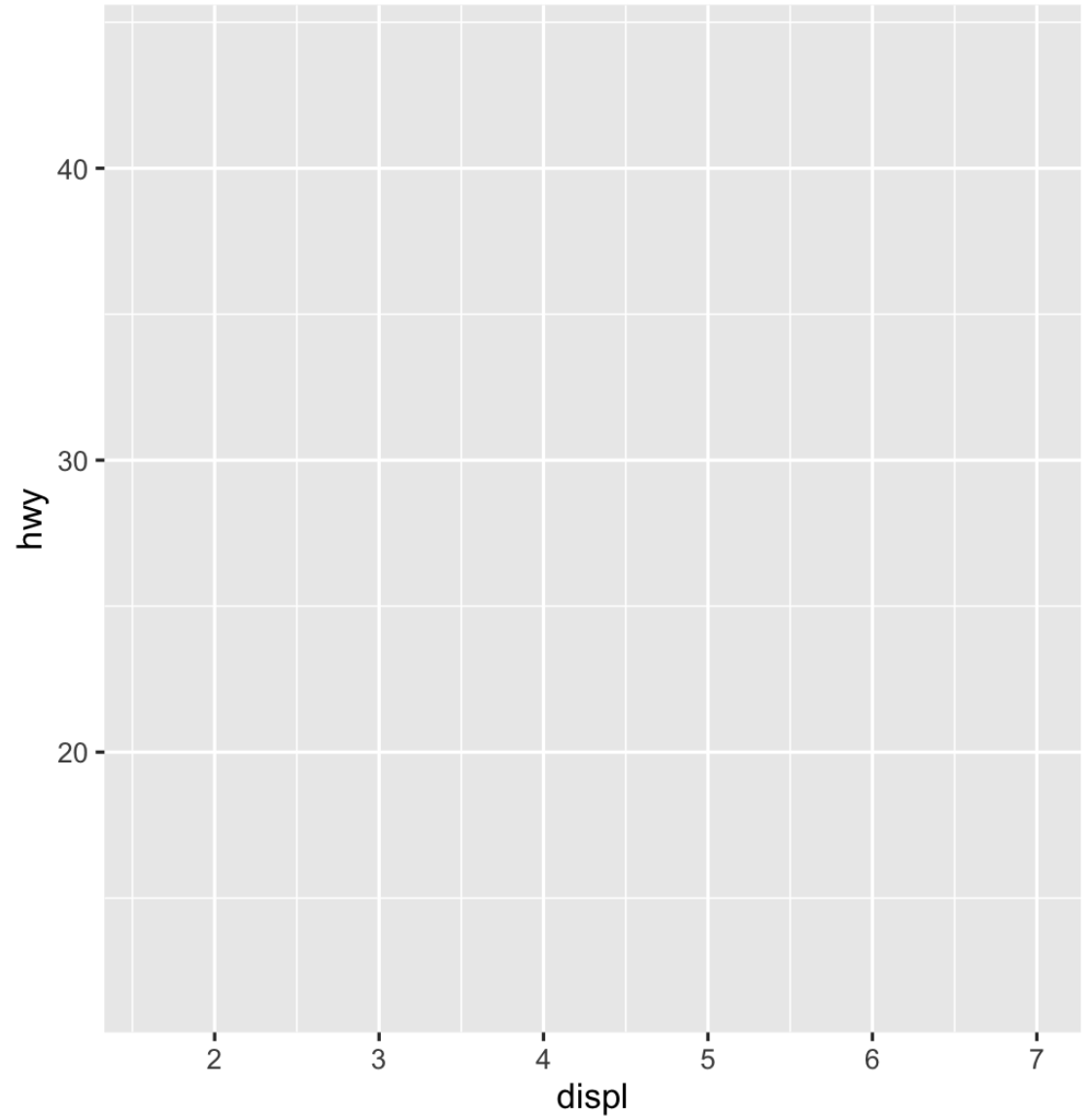


Putting it all together

**We can build a plot sequentially
to see how each grammatical layer
changes the appearance**

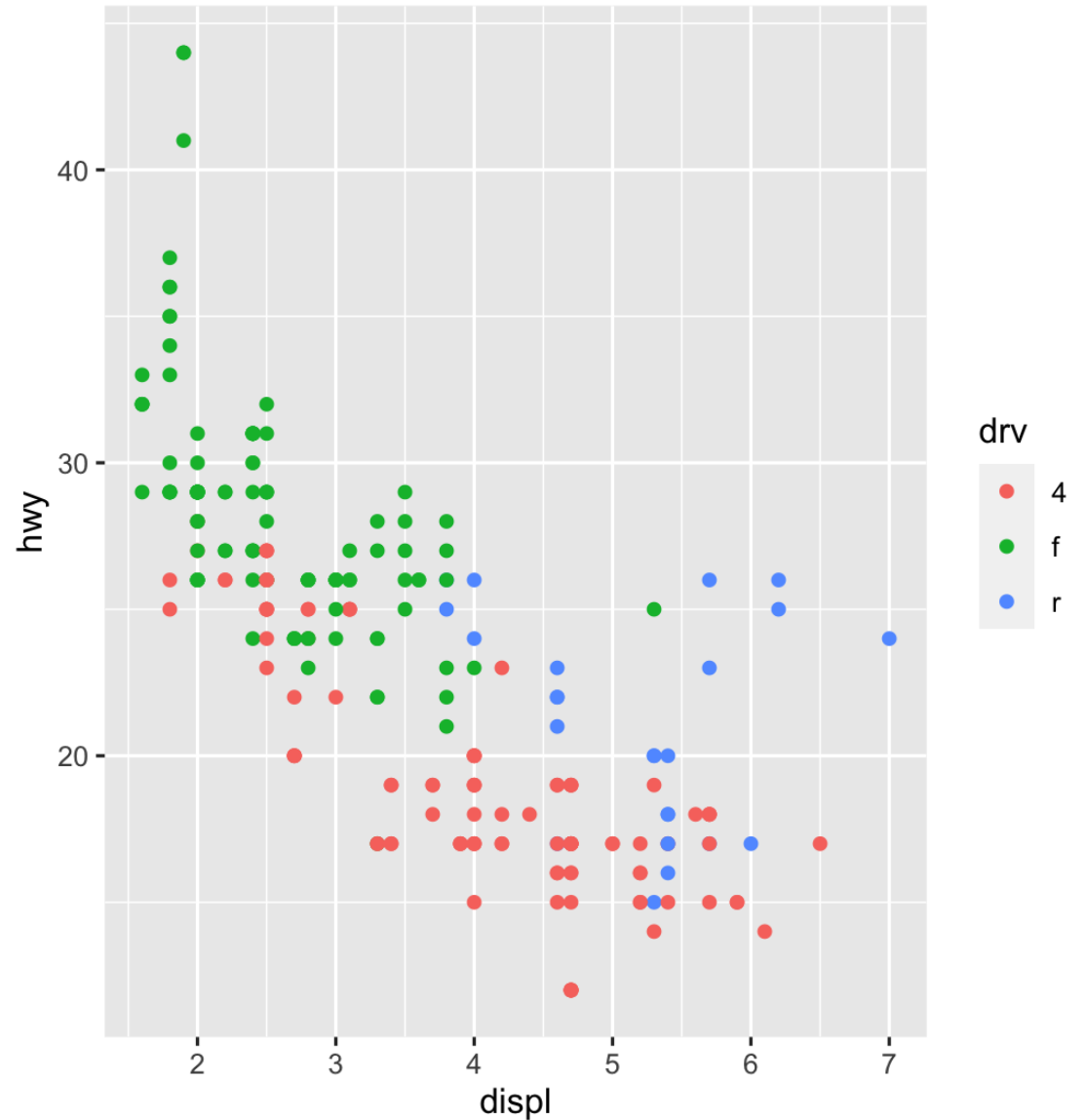
Start with data and aesthetics

```
ggplot(data = mpg,  
       mapping = aes(x = displ,  
                     y = hwy,  
                     color = drv))
```



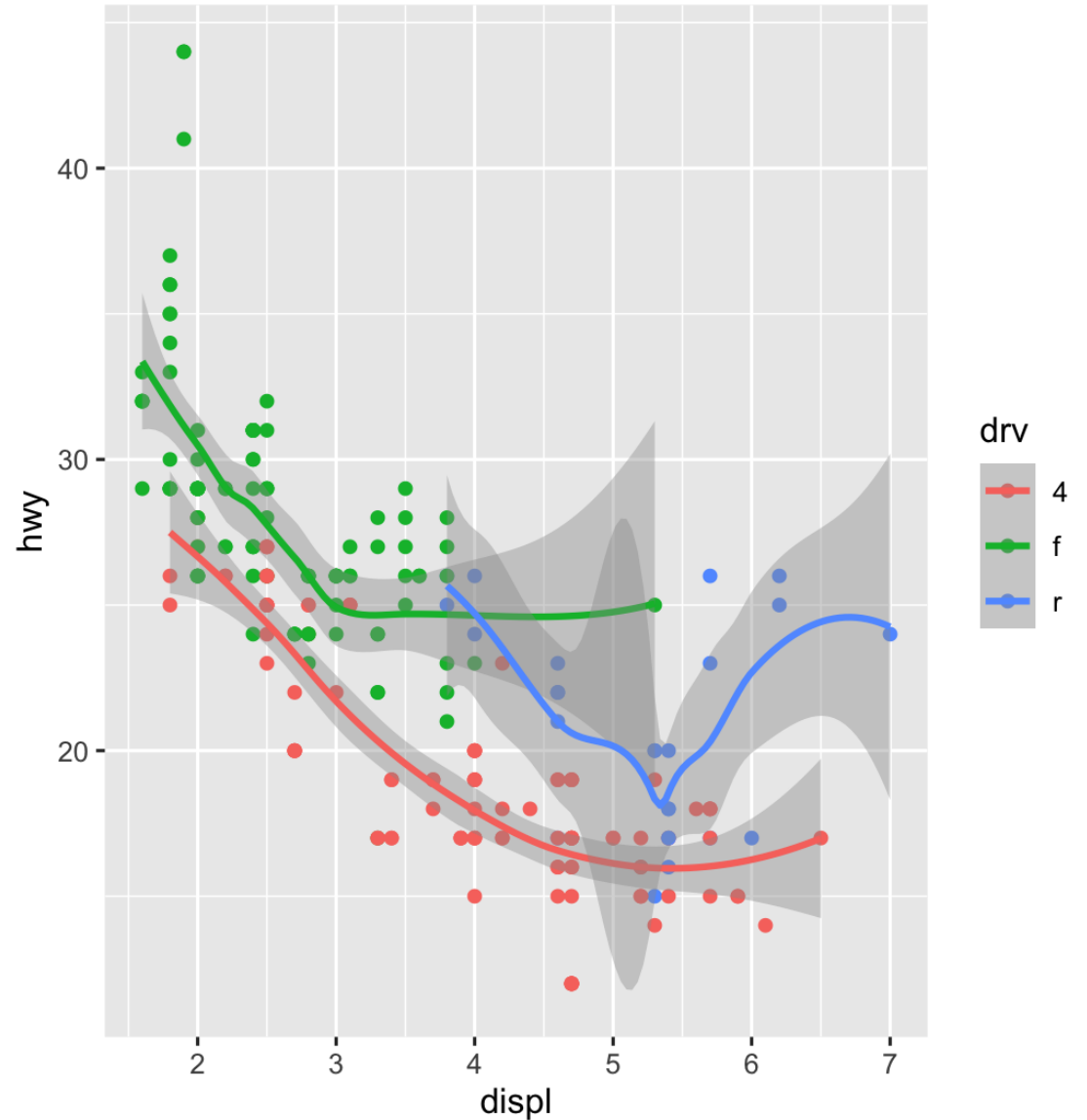
Add a point geom

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



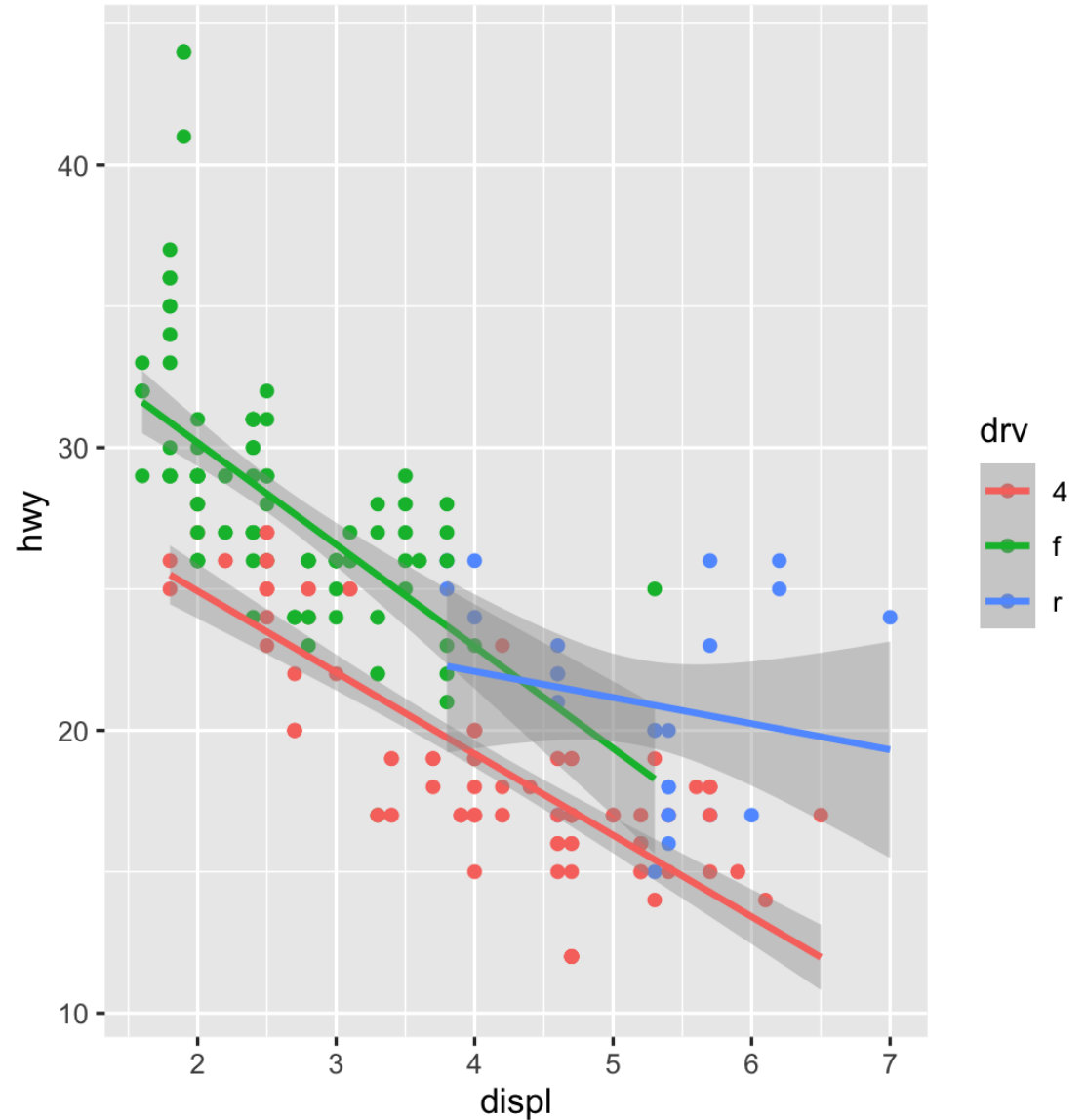
Add a smooth geom

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



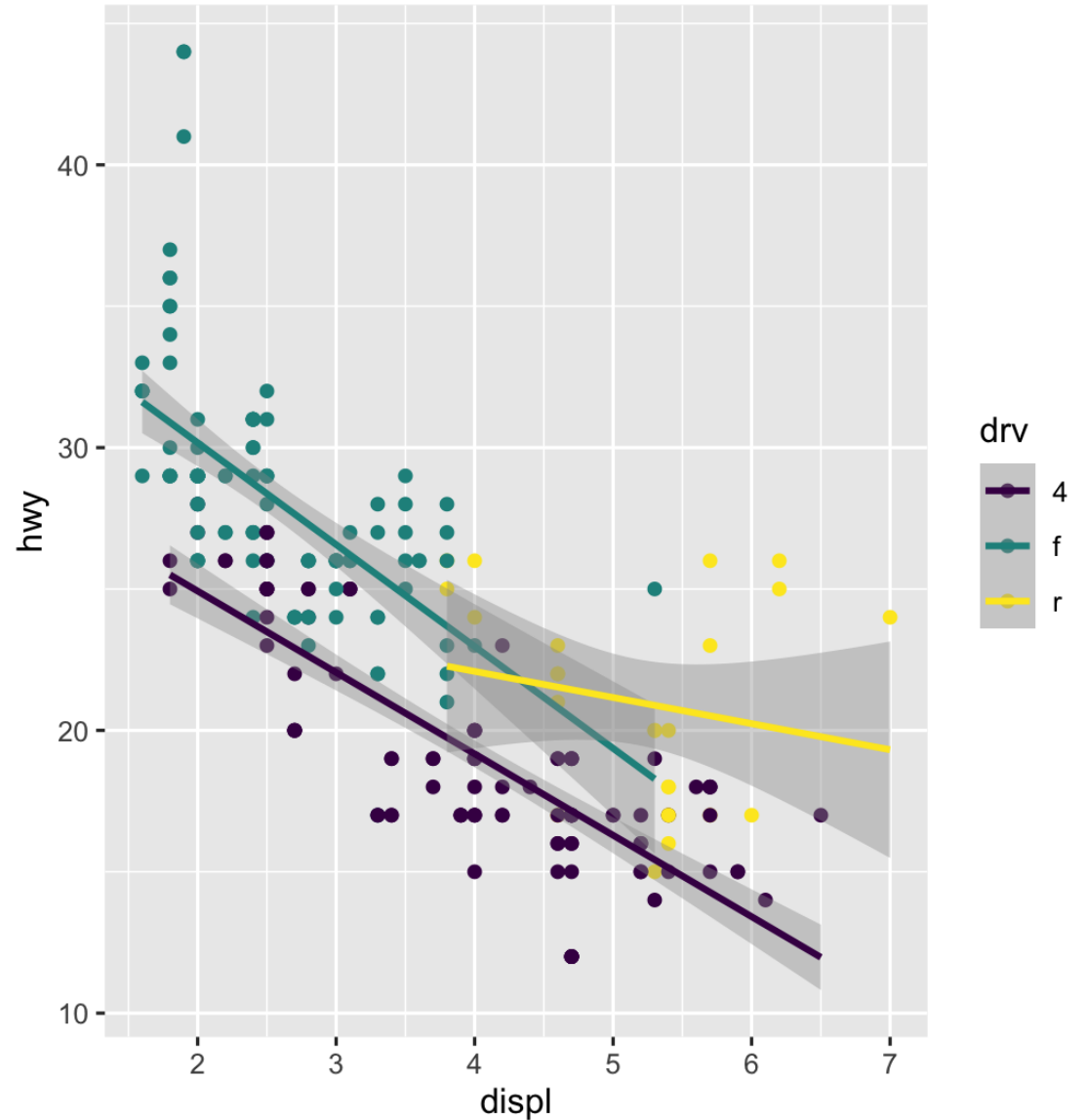
Make it straight

```
ggplot(data = mpg,  
       mapping = aes(x = displ,  
                     y = hwy,  
                     color = drv)) +  
  geom_point() +  
  geom_smooth(method = "lm")
```



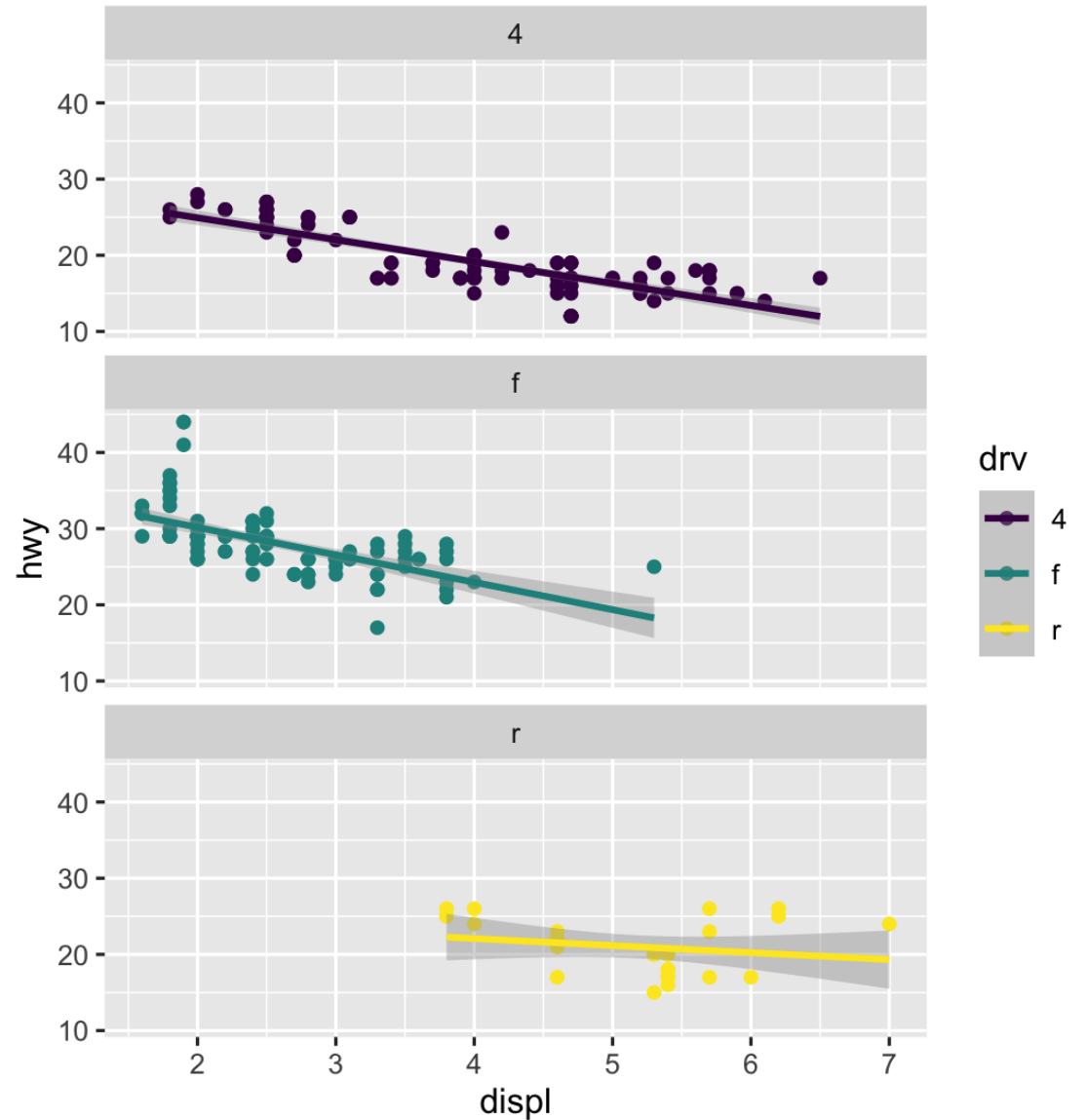
Use a viridis color scale

```
ggplot(data = mpg,  
       mapping = aes(x = displ,  
                     y = hwy,  
                     color = drv)) +  
  geom_point() +  
  geom_smooth(method = "lm") +  
  scale_color_viridis_d()
```



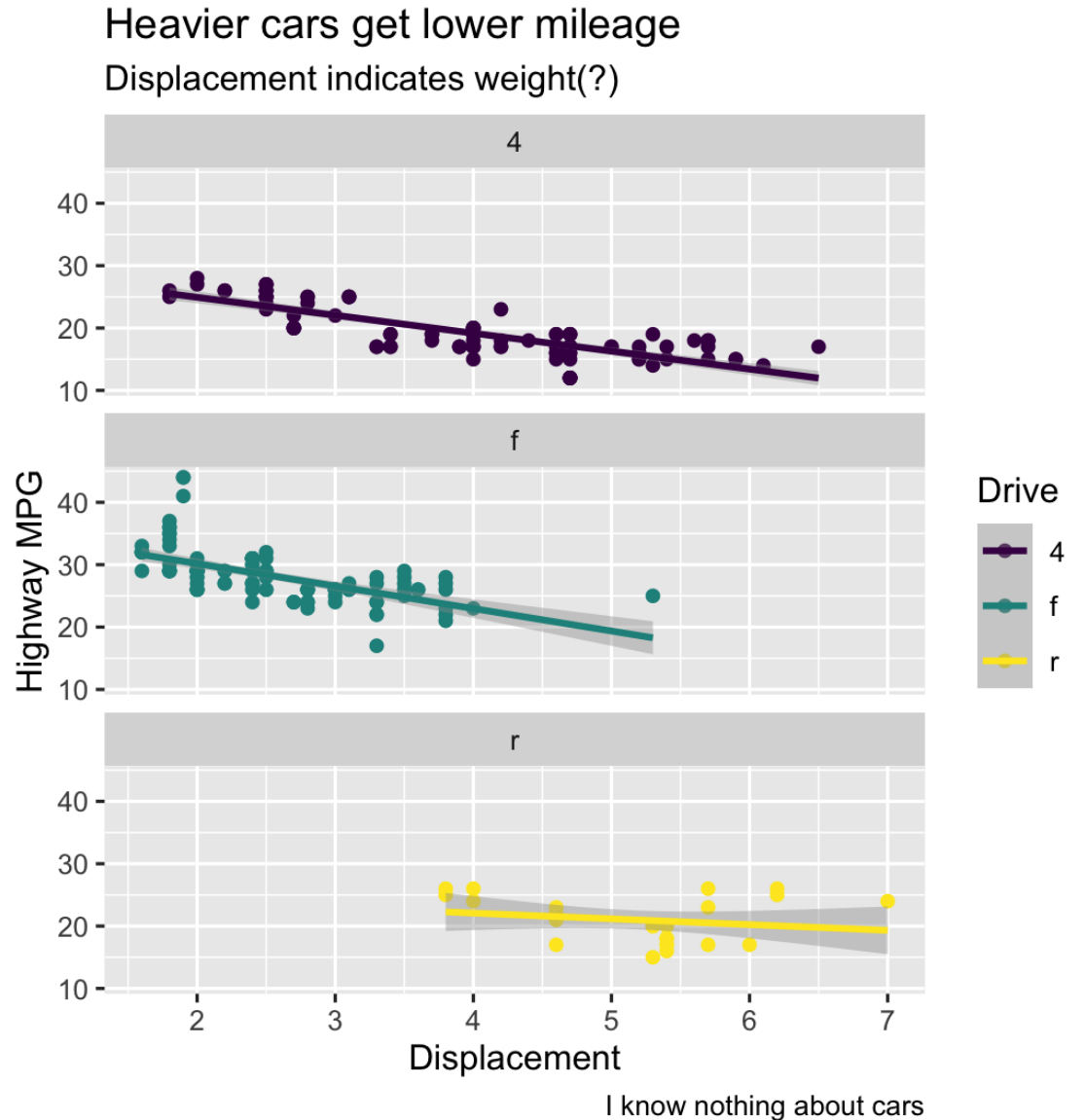
Facet by drive

```
ggplot(data = mpg,  
       mapping = aes(x = displ,  
                     y = hwy,  
                     color = drv)) +  
  geom_point() +  
  geom_smooth(method = "lm") +  
  scale_color_viridis_d() +  
  facet_wrap(vars(drv), ncol = 1)
```



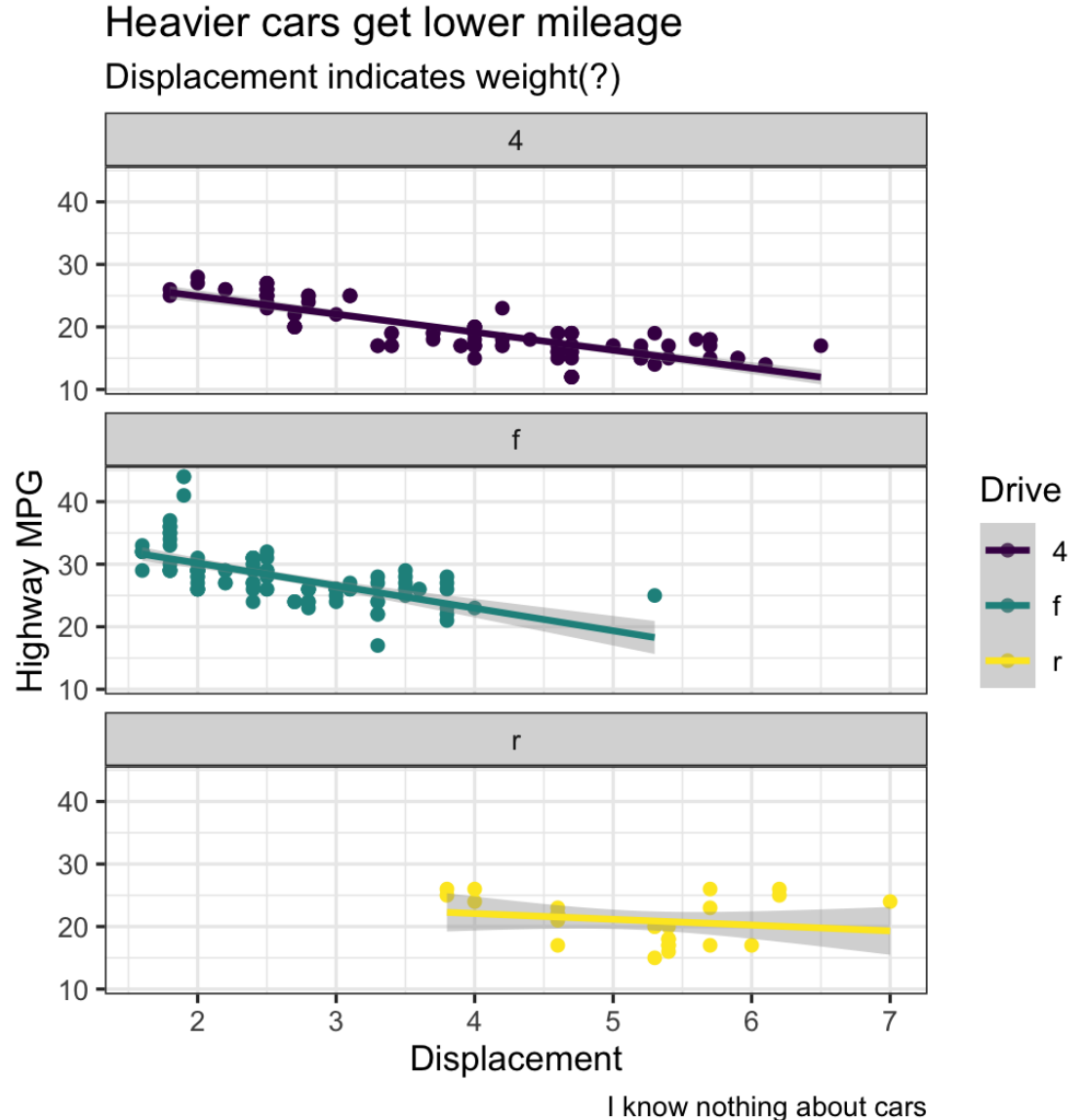
Add labels

```
ggplot(data = mpg,
       mapping = aes(x = displ,
                     y = hwy,
                     color = drv)) +
  geom_point() +
  geom_smooth(method = "lm") +
  scale_color_viridis_d() +
  facet_wrap(vars(drv), ncol = 1) +
  labs(x = "Displacement", y = "Highway MPG",
       color = "Drive",
       title = "Heavier cars get lower mileage",
       subtitle = "Displacement indicates weight",
       caption = "I know nothing about cars")
```



Add a theme

```
ggplot(data = mpg,
       mapping = aes(x = displ,
                     y = hwy,
                     color = drv)) +
  geom_point() +
  geom_smooth(method = "lm") +
  scale_color_viridis_d() +
  facet_wrap(vars(drv), ncol = 1) +
  labs(x = "Displacement", y = "Highway MPG",
       color = "Drive",
       title = "Heavier cars get lower mileage",
       subtitle = "Displacement indicates weight",
       caption = "I know nothing about cars")
  theme_bw()
```

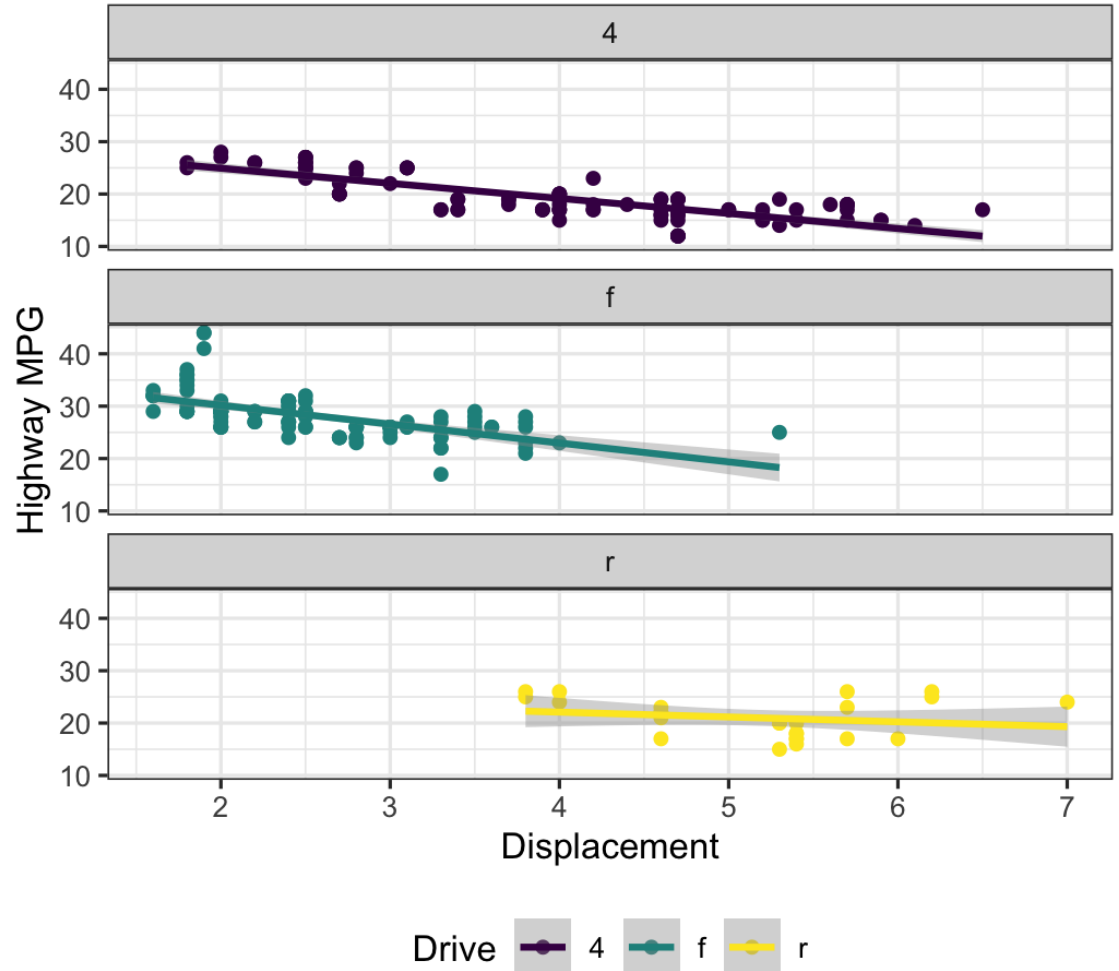


Modify the theme

```
ggplot(data = mpg,
       mapping = aes(x = displ,
                     y = hwy,
                     color = drv)) +
  geom_point() +
  geom_smooth(method = "lm") +
  scale_color_viridis_d() +
  facet_wrap(vars(drv), ncol = 1) +
  labs(x = "Displacement", y = "Highway MPG",
       color = "Drive",
       title = "Heavier cars get lower mileage",
       subtitle = "Displacement indicates weight",
       caption = "I know nothing about cars") +
  theme_bw() +
  theme(legend.position = "bottom",
       plot.title = element_text(face = "bold"))
```

Heavier cars get lower mileage

Displacement indicates weight(?)



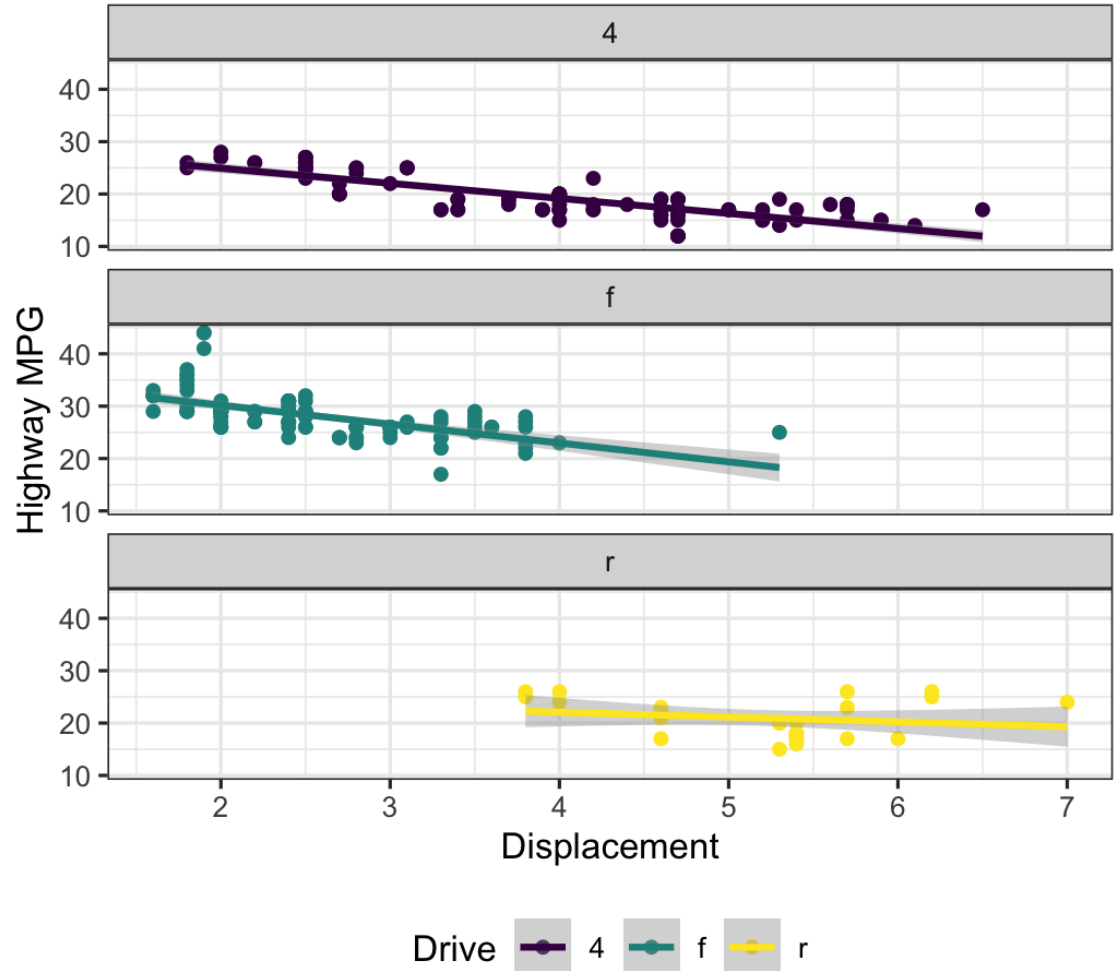
I know nothing about cars

Finished!

```
ggplot(data = mpg,
       mapping = aes(x = displ,
                     y = hwy,
                     color = drv)) +
  geom_point() +
  geom_smooth(method = "lm") +
  scale_color_viridis_d() +
  facet_wrap(vars(drv), ncol = 1) +
  labs(x = "Displacement", y = "Highway MPG",
       color = "Drive",
       title = "Heavier cars get lower mileage",
       subtitle = "Displacement indicates weight",
       caption = "I know nothing about cars") +
  theme_bw() +
  theme(legend.position = "bottom",
       plot.title = element_text(face = "bold"))
```

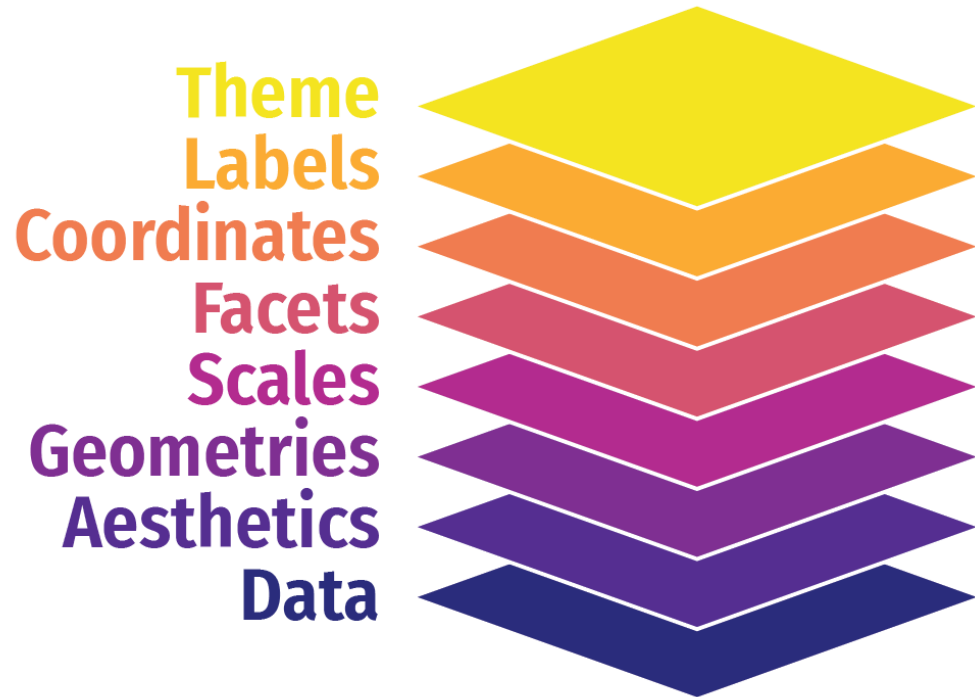
Heavier cars get lower mileage

Displacement indicates weight(?)



I know nothing about cars

So many possibilities!



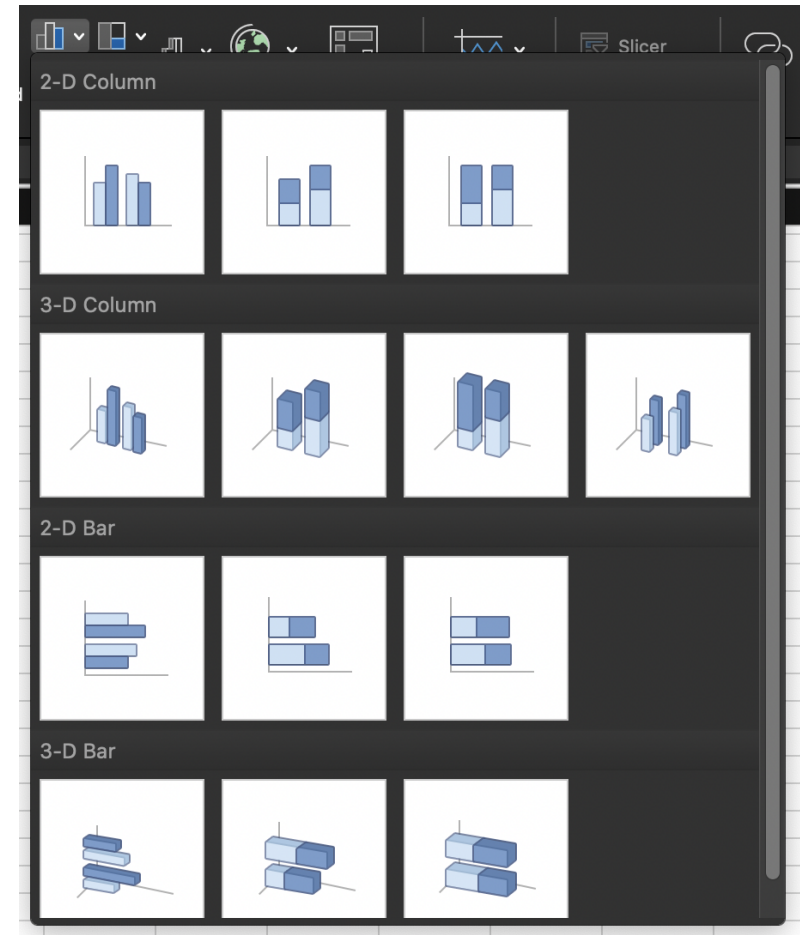
These were just a few examples of layers!

See **the ggplot2 documentation** for complete examples of everything you can do

A true grammar

With the grammar of graphics,
we don't talk about specific
chart *types*

Hunt through Excel menus for a
stacked bar chart and manually
reshape your data to work with it

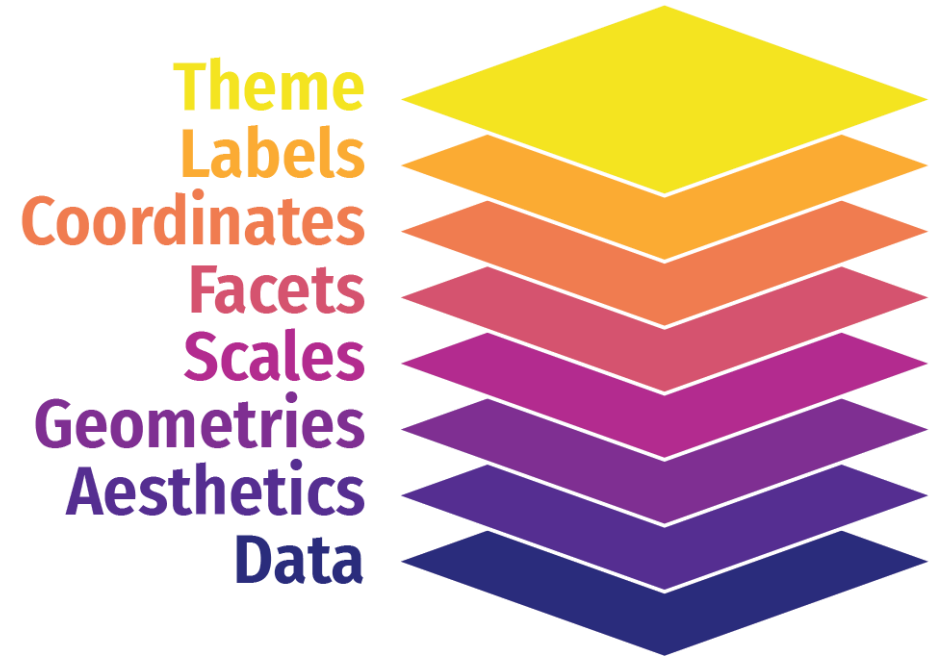


A true grammar

With the grammar of graphics,
we *do* talk about specific
chart *elements*

Map a column to the x-axis, fill by a
different variable, and `geom_col()` to
get stacked bars

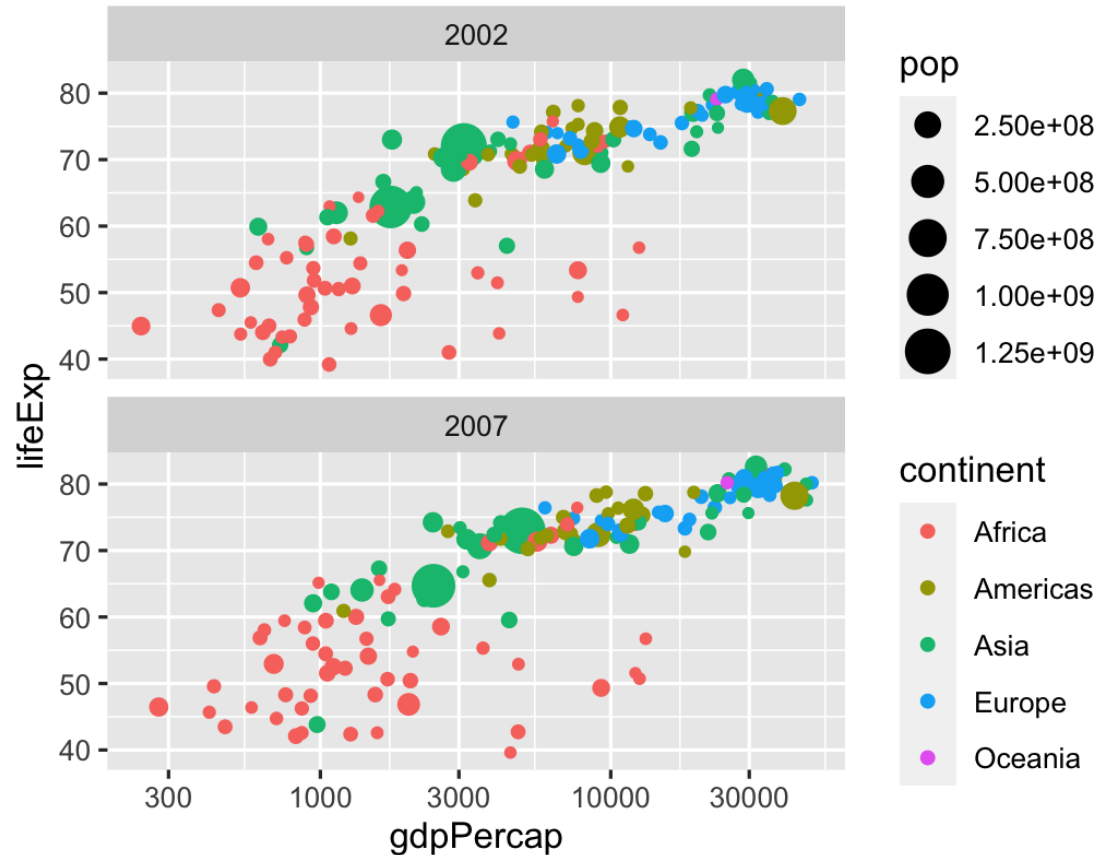
Geoms can be interchangeable
(e.g. switch `geom_violin()` to
`geom_boxplot()`)



Describing graphs with the grammar

Map wealth to the x-axis, health to the y-axis, add points, color by continent, size by population, scale the y-axis with a log, and facet by year

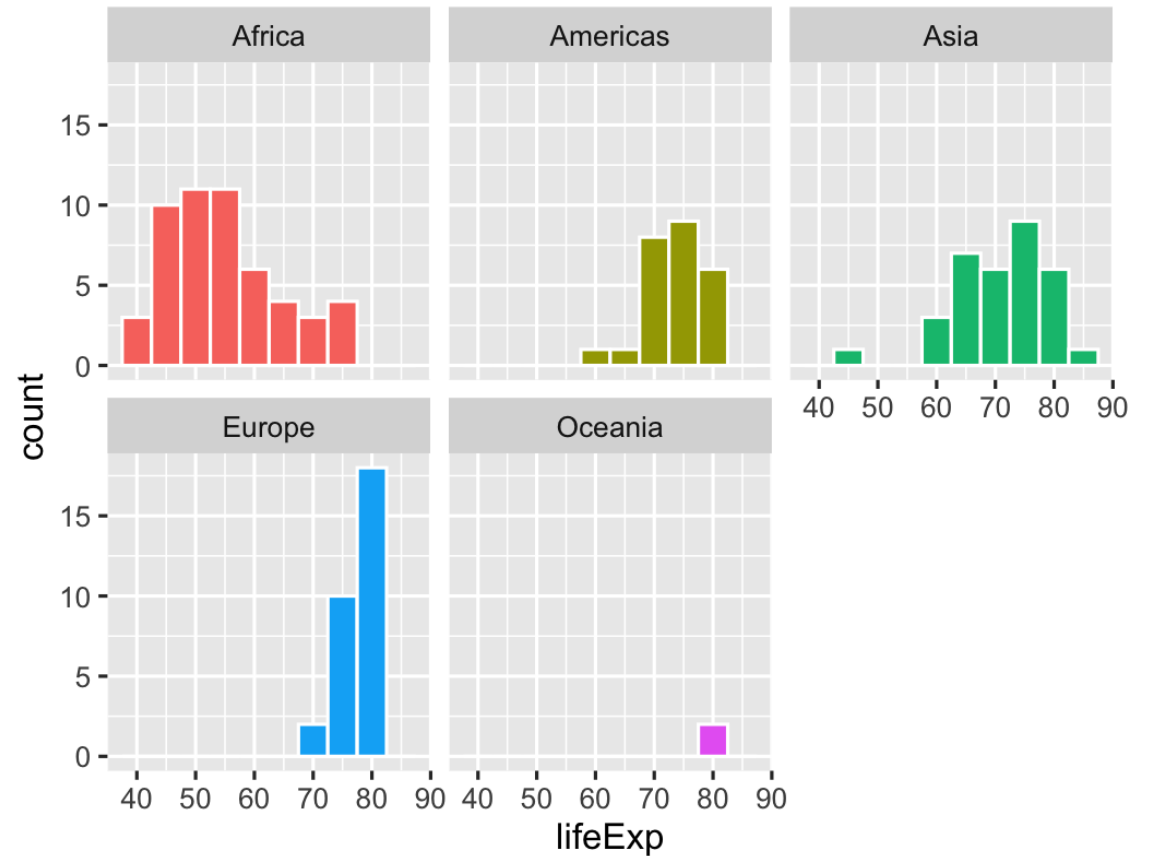
```
ggplot(filter(gapminder,  
            year %in% c(2002, 2007)),  
       aes(x = gdpPerCap,  
           y = lifeExp,  
           color = continent,  
           size = pop)) +  
  geom_point() +  
  scale_x_log10() +  
  facet_wrap(vars(year), ncol = 1)
```



Describing graphs with the grammar

Map health to the x-axis, add a histogram with bins for every 5 years, fill and facet by continent

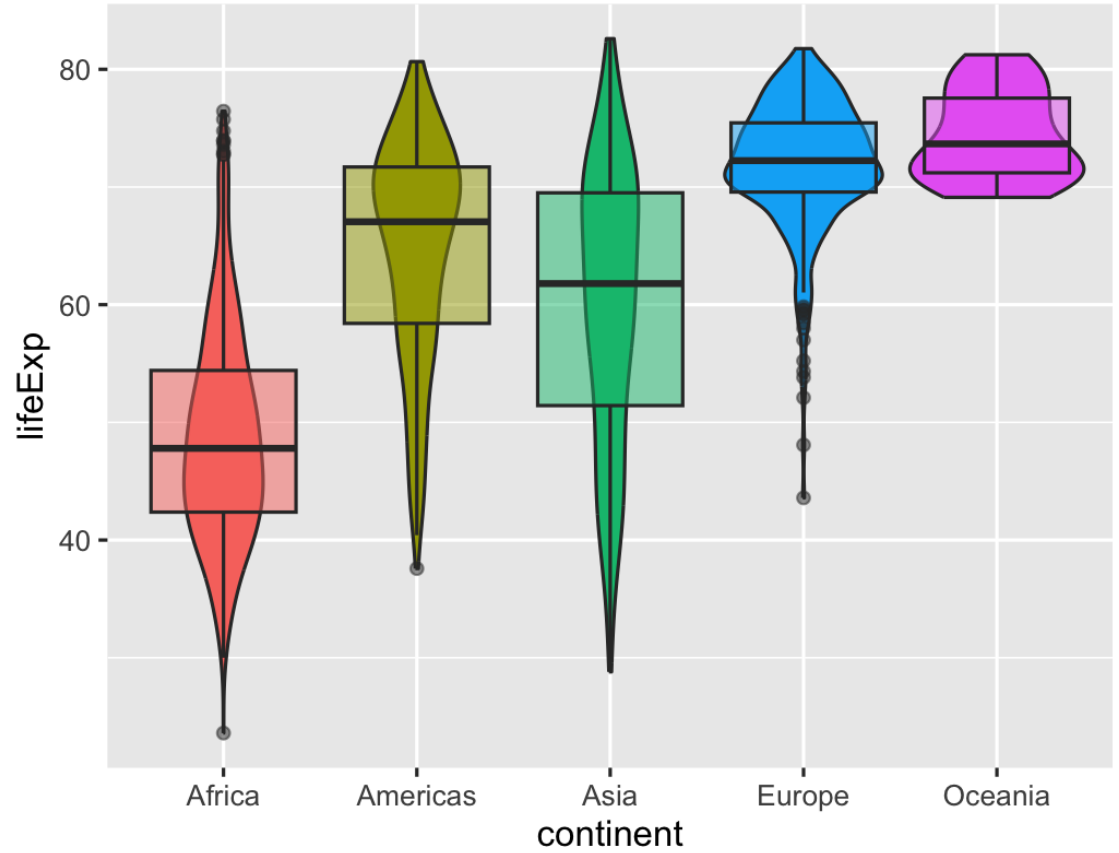
```
ggplot(gapminder_2007,  
       aes(x = lifeExp,  
           fill = continent)) +  
  geom_histogram(binwidth = 5,  
                color = "white") +  
  guides(fill = FALSE) + # Turn off legend  
  facet_wrap(vars(continent))
```



Describing graphs with the grammar

Map continent to the x-axis, health to the y-axis, add violin plots and semi-transparent boxplots, fill by continent

```
ggplot(gapminder,  
      aes(x = continent,  
          y = lifeExp,  
          fill = continent)) +  
  geom_violin() +  
  geom_boxplot(alpha = 0.5) +  
  guides(fill = FALSE) # Turn off legend
```



Scales

Scales change the properties of the variable mapping

Example layer

```
scale_x_continuous()
```

```
scale_x_continuous(breaks = 1:5)
```

```
scale_x_log10()
```

```
scale_color_gradient()
```

```
scale_fill_viridis_d()
```

What it does

Make the x-axis continuous

Manually specify axis ticks

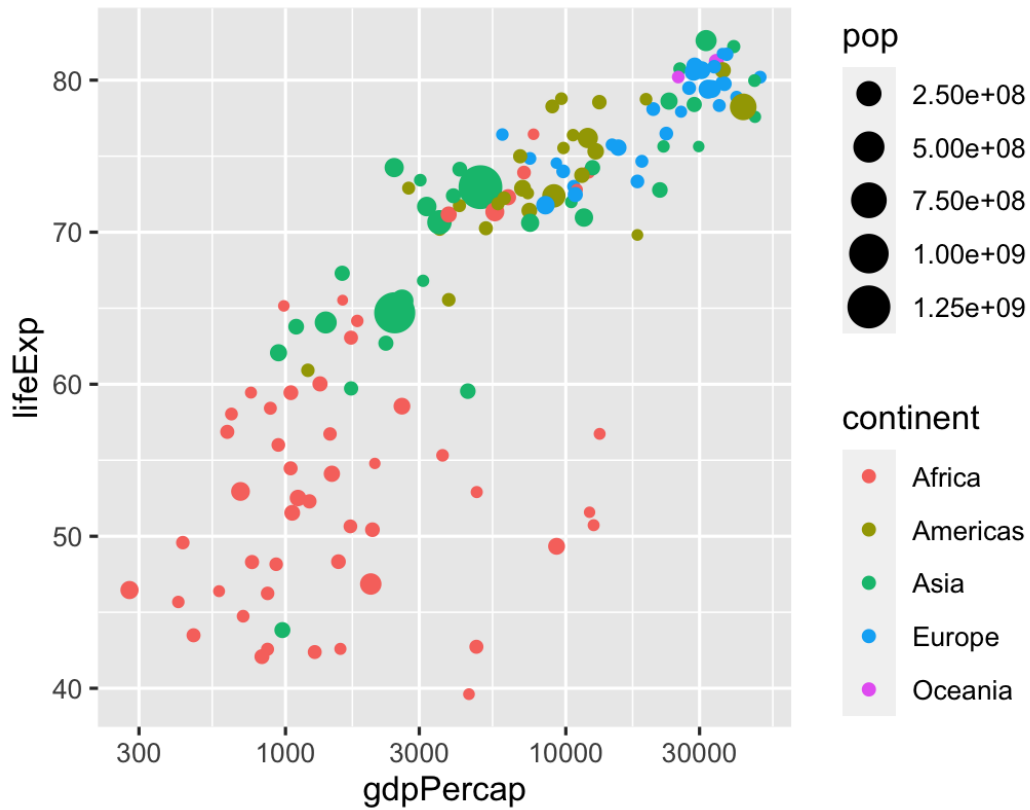
Log the x-axis

Use a gradient

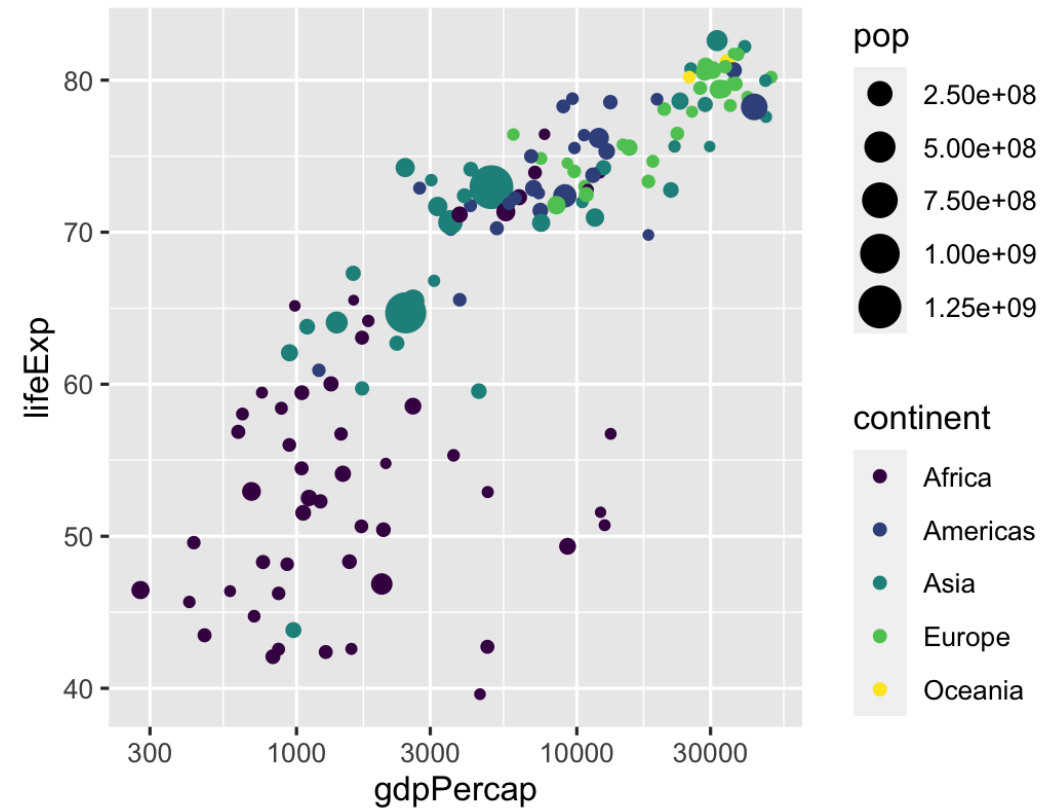
Fill with discrete viridis colors

Scales

`scale_x_log10()`



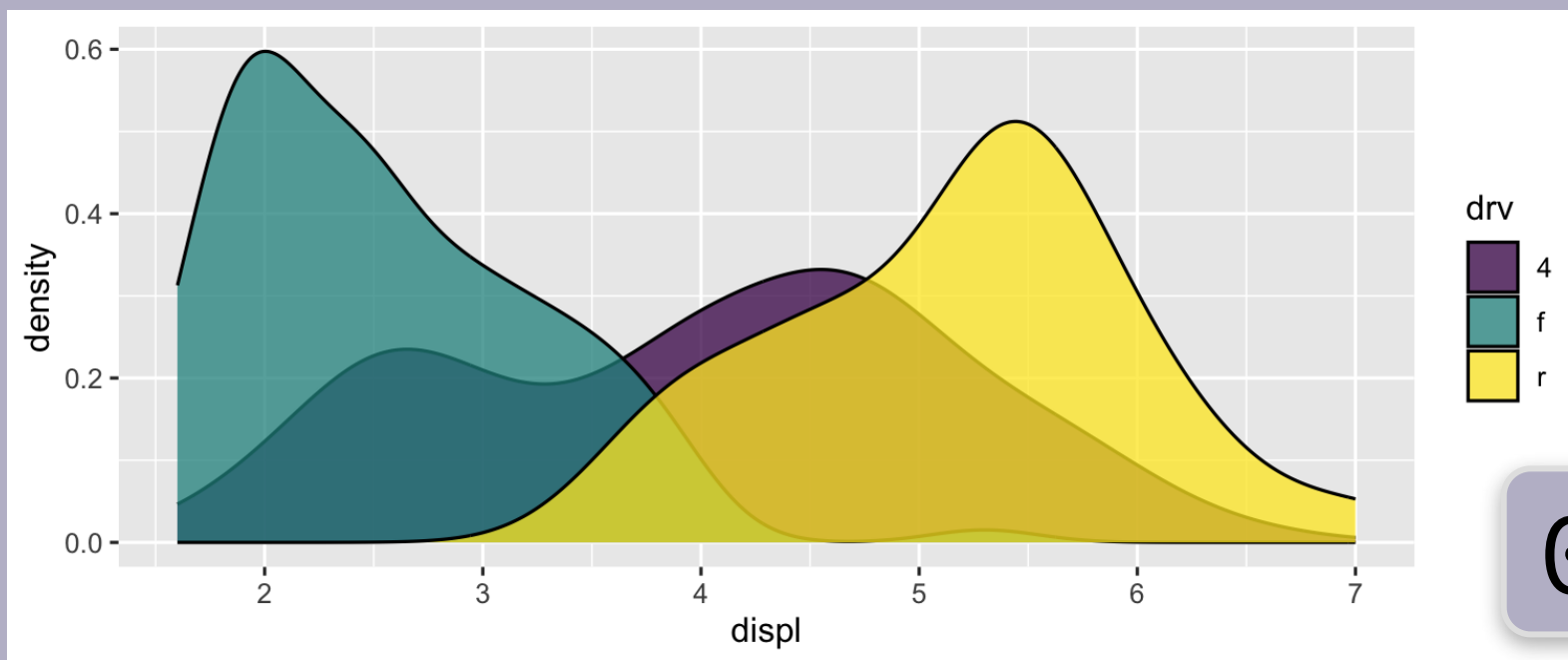
`scale_color_viridis_d()`



Your turn #7

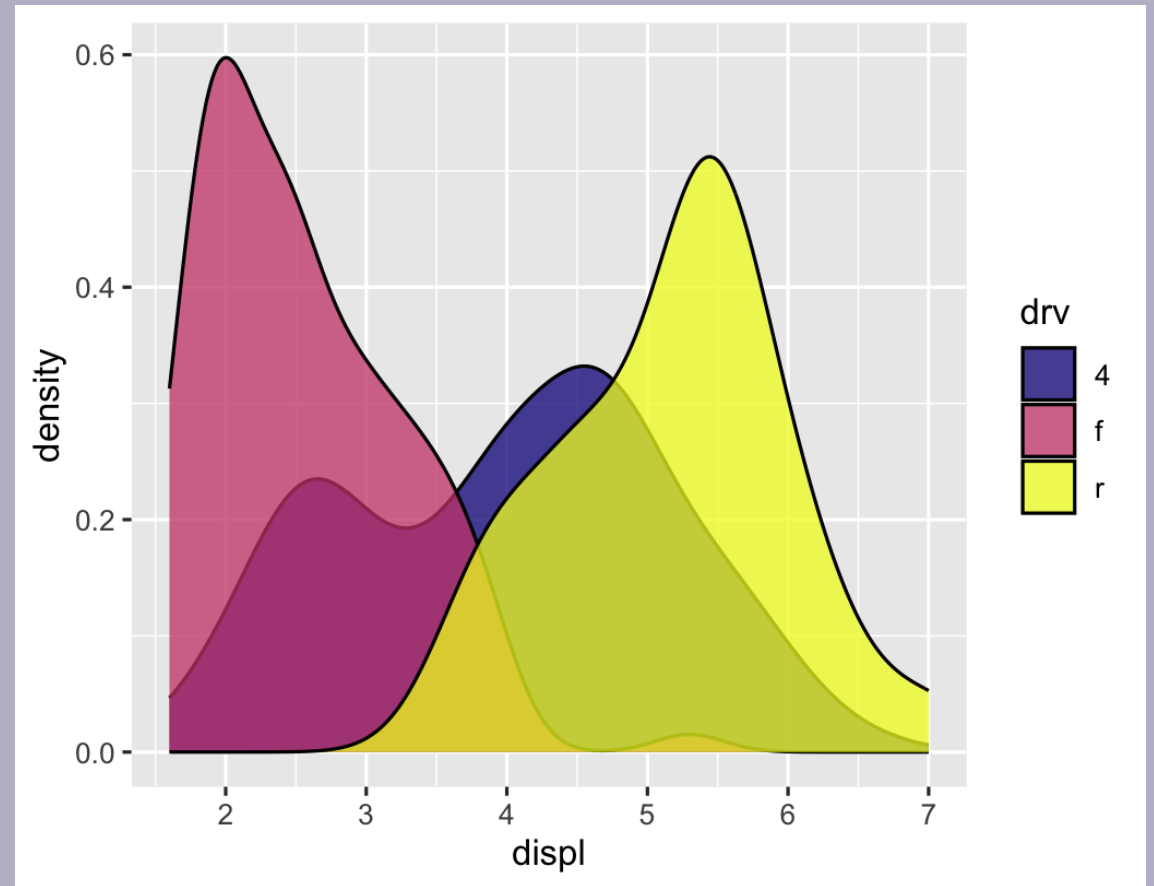
Make this density plot of `displ` filled by `drv`.
Use the viridis fill scale.

For bonus fun, try a different viridis option like `plasma` OR `inferno`.



03:00

```
ggplot(mpg,  
       aes(x = displ,  
           fill = drv)) +  
  geom_density(alpha = 0.75) +  
  scale_fill_viridis_d(option = "plasma")
```



Facets

Facets show subplots for different subsets of data

Example layer

```
facet_wrap(vars(continent))
```

What it does

Plot for each continent

```
facet_wrap(vars(continent, year))
```

Plot for each continent/year

```
facet_wrap(..., ncol = 1)
```

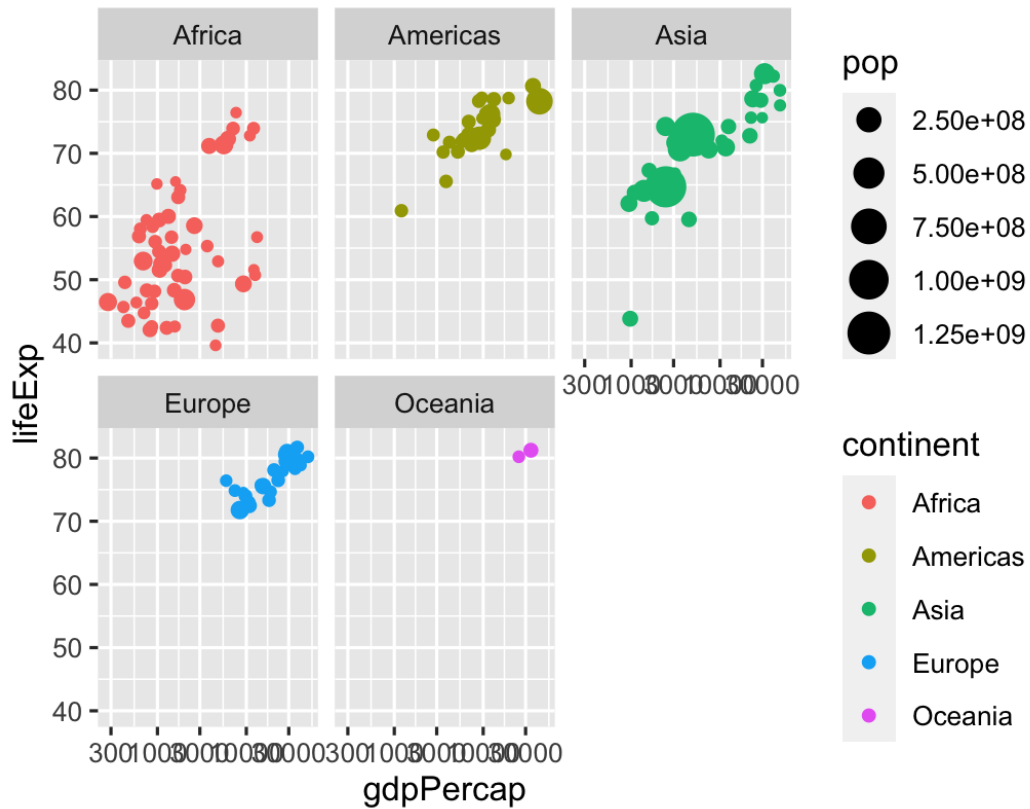
Put all facets in one column

```
facet_wrap(..., nrow = 1)
```

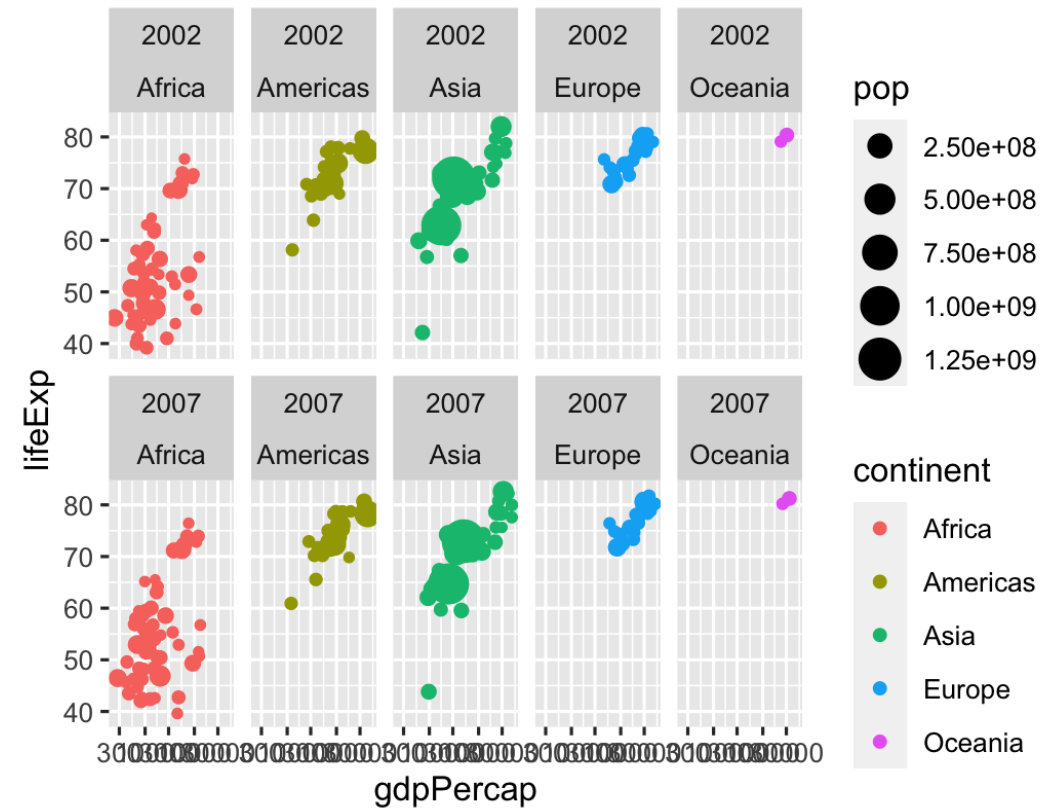
Put all facets in one row

Facets

```
facet_wrap(vars(continent))
```

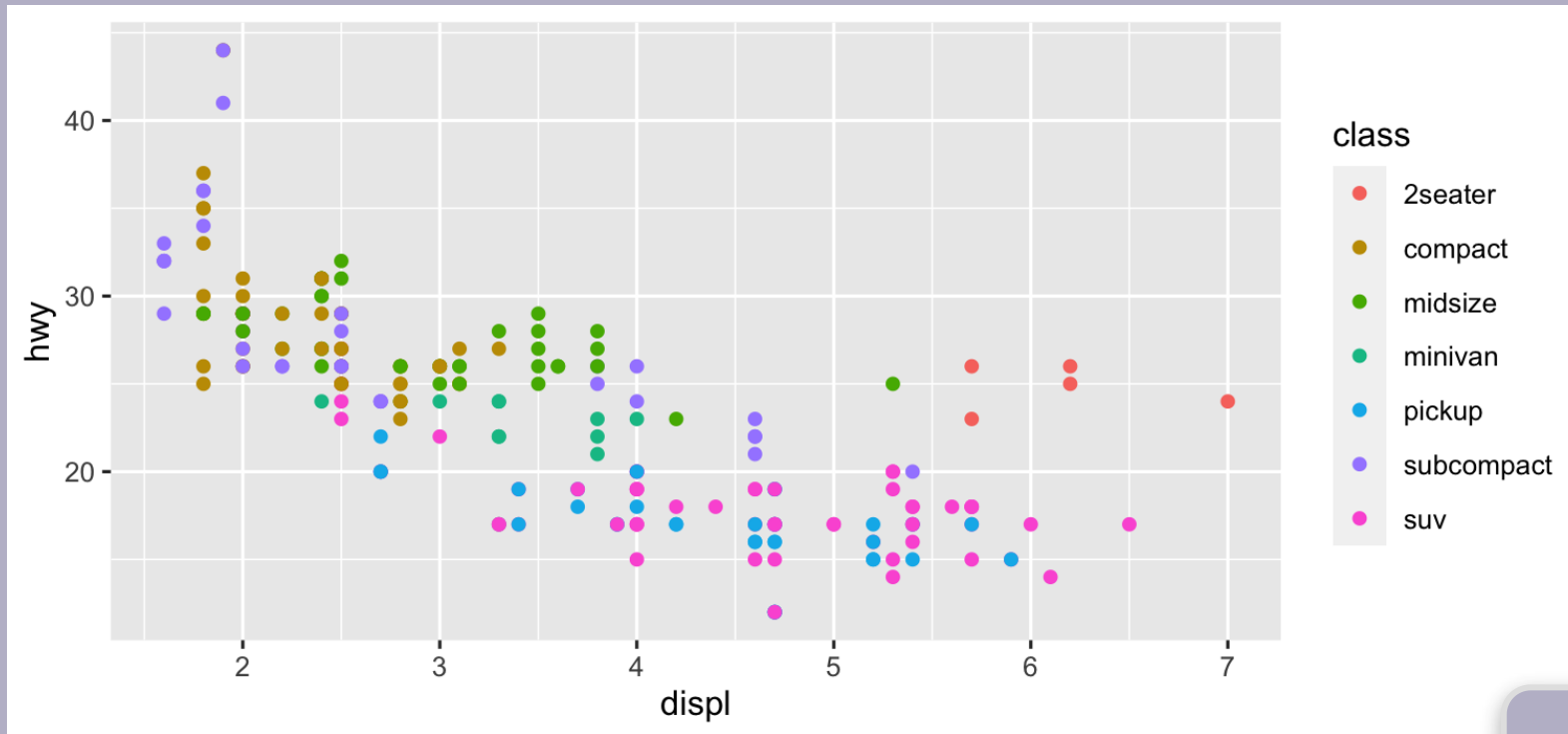


```
facet_wrap(vars(continent, year))
```



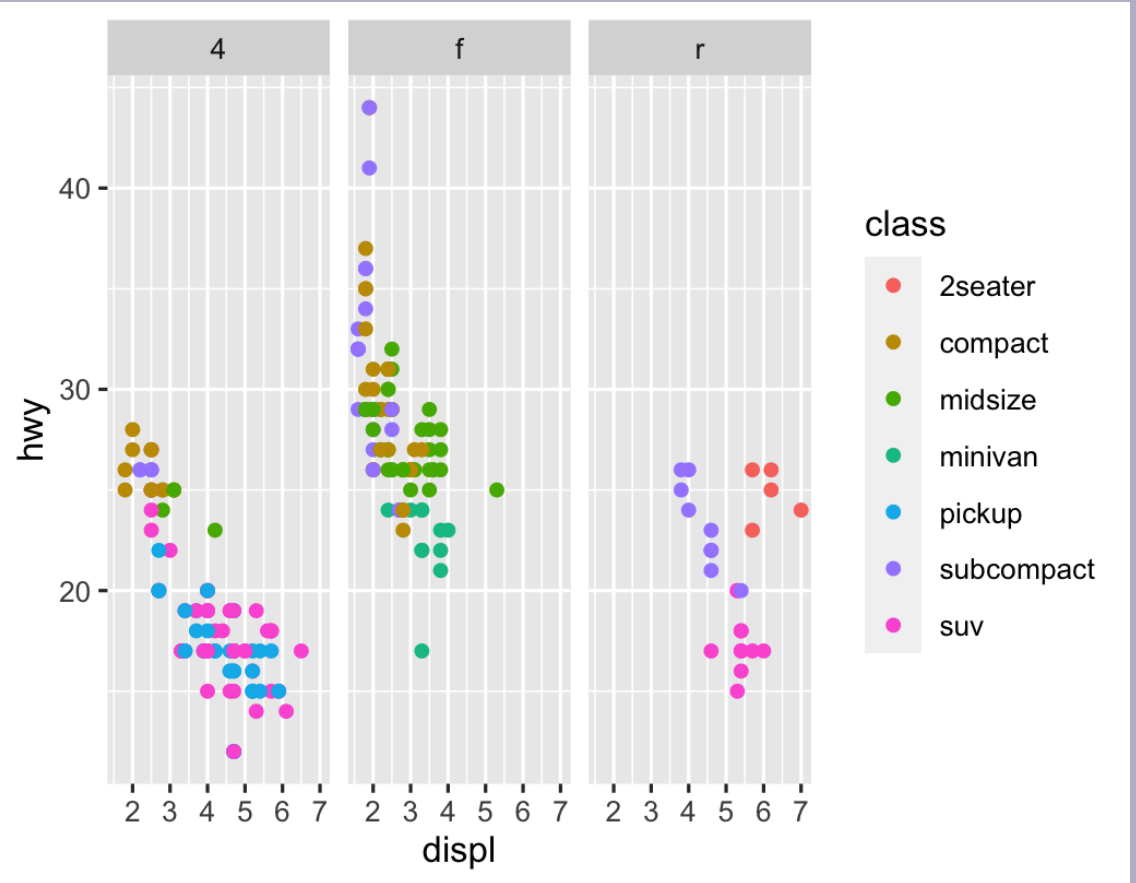
Your turn #8

Facet this scatterplot by `drv`. Are there any interesting trends?



03:00

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



Coordinates

Change the coordinate system

Example layer

```
coord_cartesian()
```

What it does

Standard x-y coordinate system

```
coord_cartesian(ylim = c(1,  
10))
```

Zoom in where y is 1–10

```
coord_flip()
```

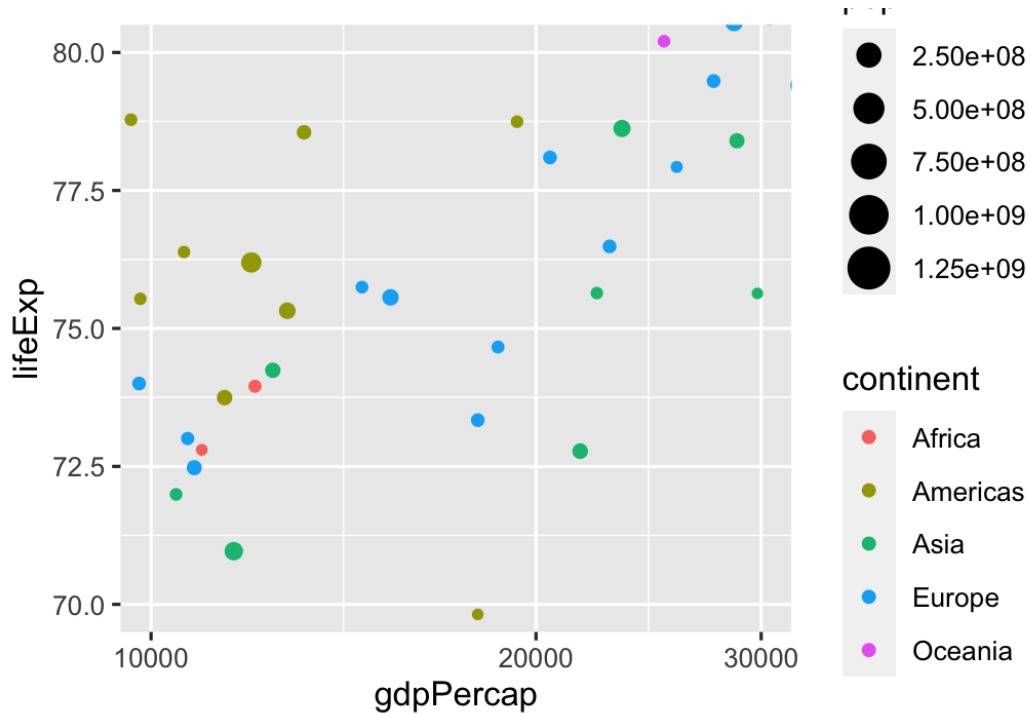
Switch x and y

```
coord_polar()
```

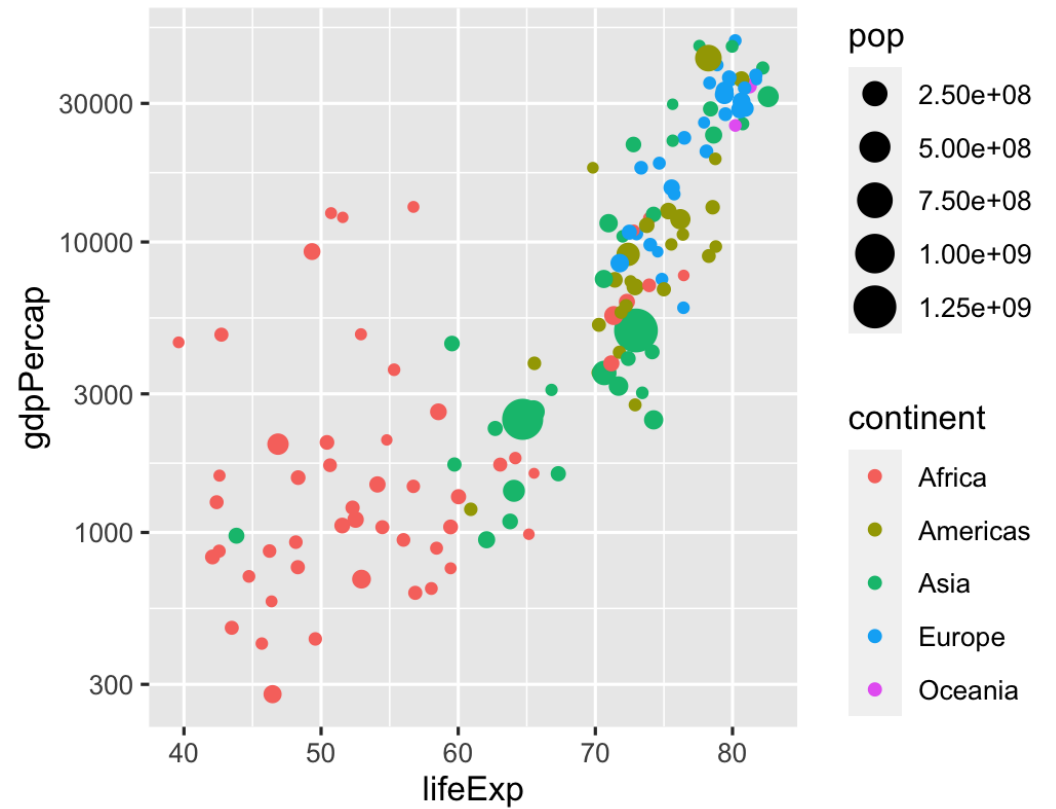
Use circular polar system

Coordinates

```
coord_cartesian(ylim = c(70, 80),  
               xlim = c(10000, 30000))
```



```
coord_flip()
```



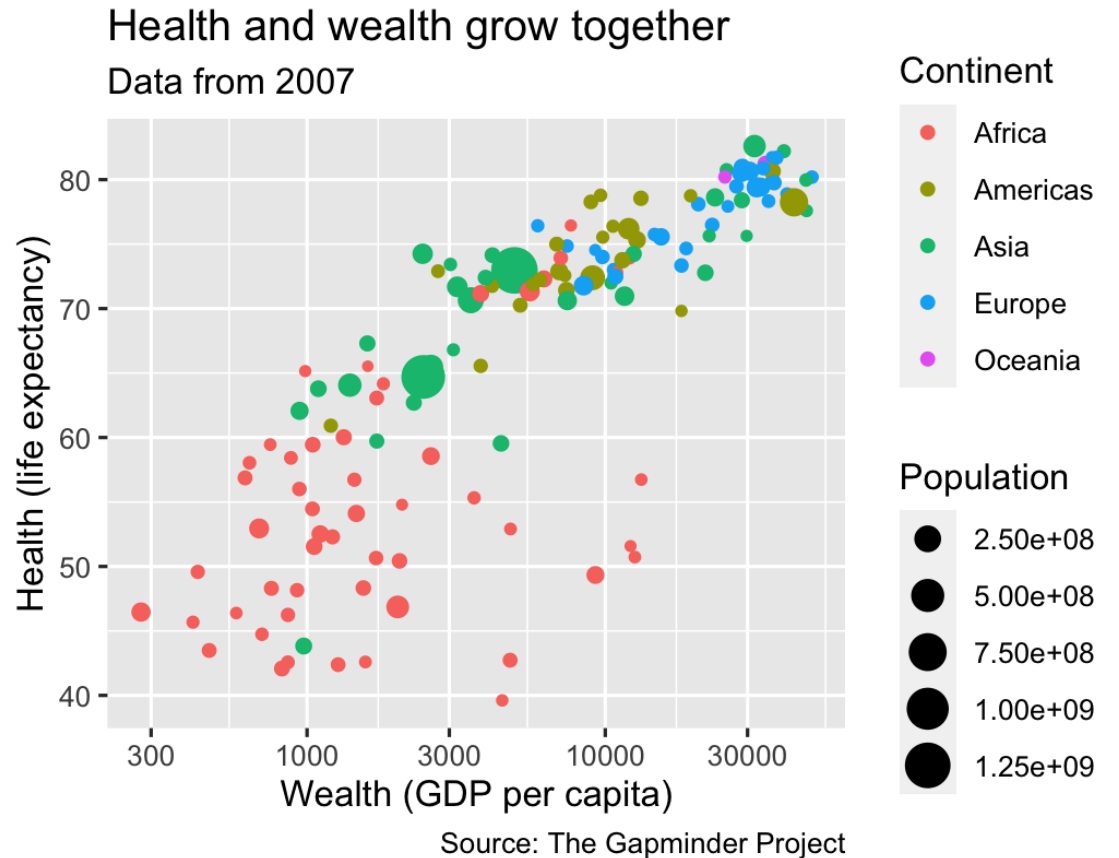
Labels

Add labels to the plot with a single `labs()` layer

Example layer	What it does
<code>labs(title = "Neat title")</code>	Title
<code>labs(caption = "Something")</code>	Caption
<code>labs(y = "Something")</code>	y-axis
<code>labs(size = "Population")</code>	Title of size legend

Labels

```
ggplot(gapminder_2007,  
       aes(x = gdpPercap, y = lifeExp,  
           color = continent, size = pop)) +  
  geom_point() +  
  scale_x_log10() +  
  labs(title = "Health and wealth grow together",  
        subtitle = "Data from 2007",  
        x = "Wealth (GDP per capita)",  
        y = "Health (life expectancy)",  
        color = "Continent",  
        size = "Population",  
        caption = "Source: The Gapminder Project")
```



Theme

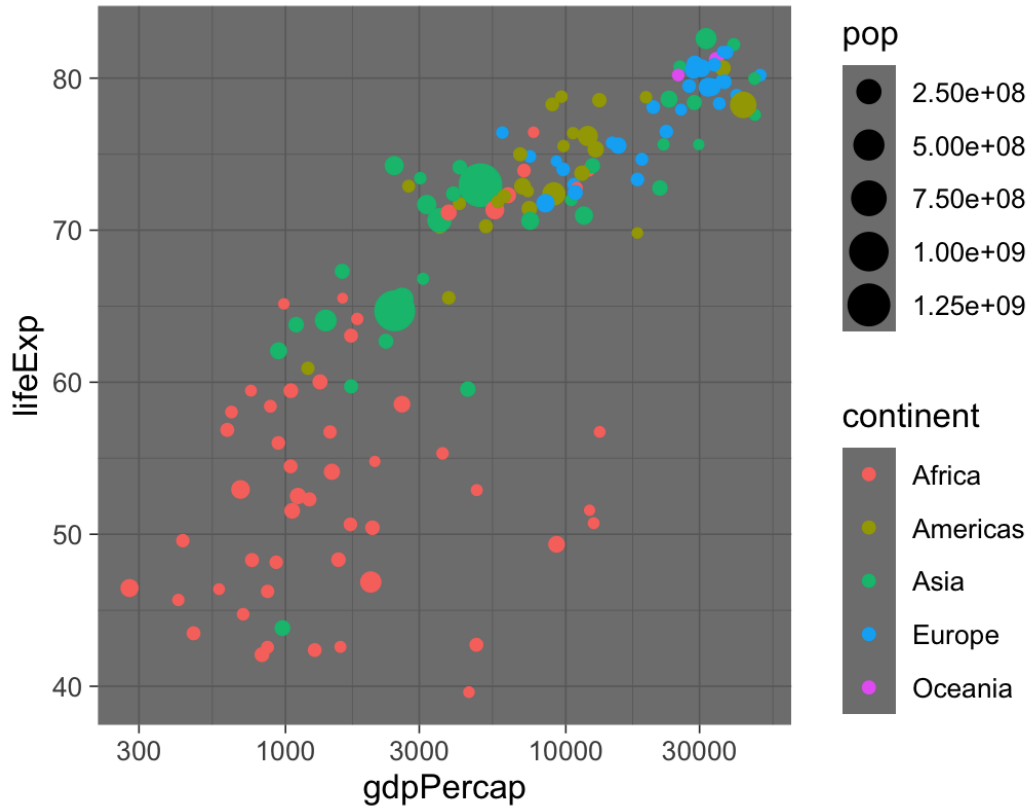
Change the appearance of anything in the plot

There are many built-in themes

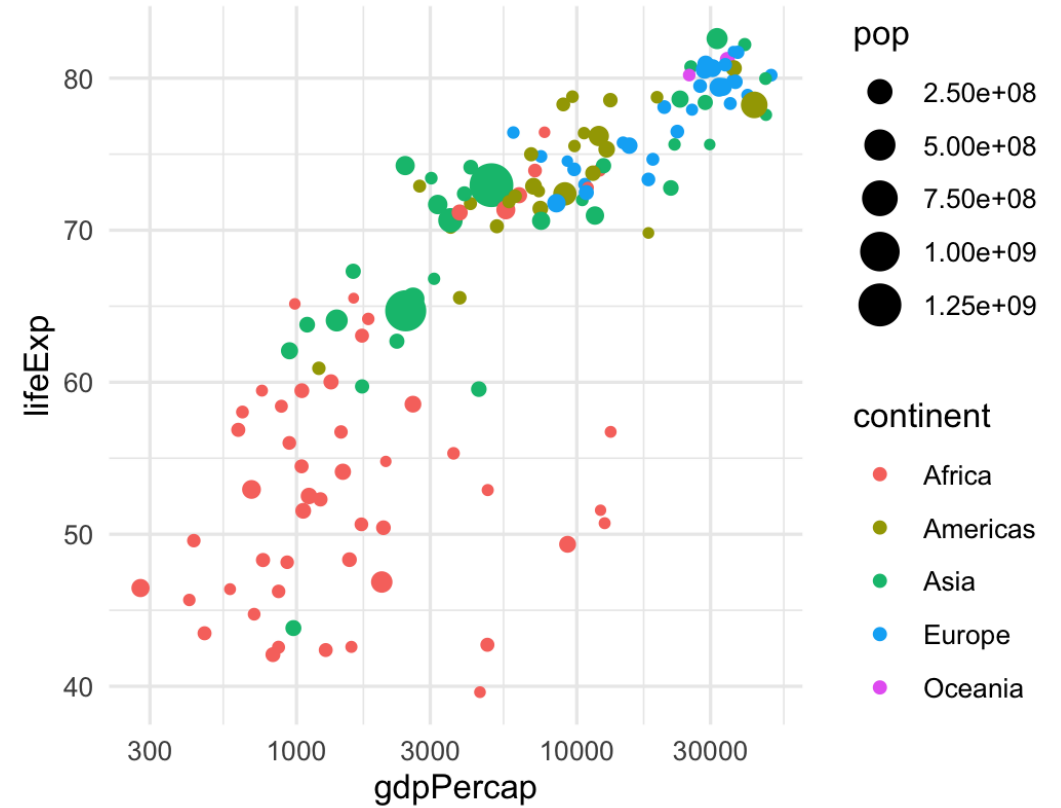
Example layer	What it does
<code>theme_grey()</code>	Default grey background
<code>theme_bw()</code>	Black and white
<code>theme_dark()</code>	Dark
<code>theme_minimal()</code>	Minimal

Theme

`theme_dark()`



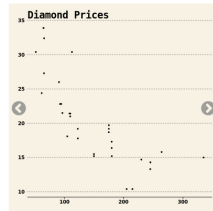
`theme_minimal()`



Theme

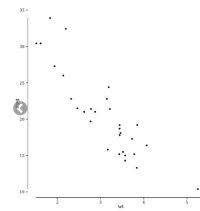
There are collections of pre-built themes online,
like **the ggthemes package**

ggthemes



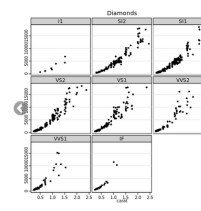
theme_wsj

Wall Street Journal theme



theme_tufte

Tufte Maximal Data, Minimal Ink Theme



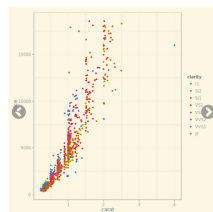
theme_stata

Themes based on Stata graph schemes



theme_solid

Theme with nothing other than a background color



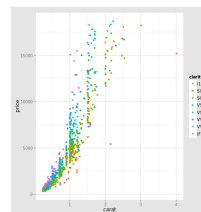
theme_solarized

ggplot color themes based on the Solarized palette



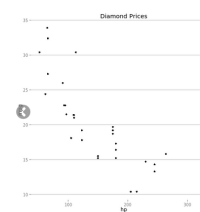
theme_map

Clean theme for maps



theme_igray

Inverse gray theme



theme_hc

Highcharts JS theme

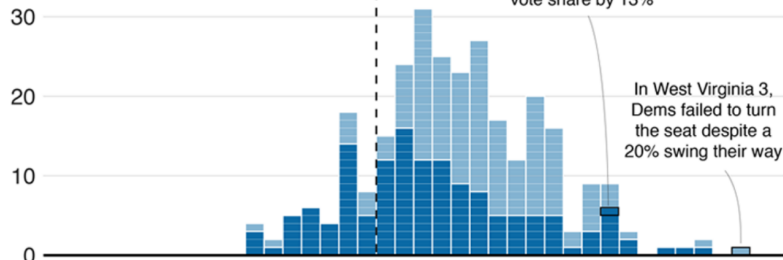
Theme

Organizations often make their own custom themes, like the BBC

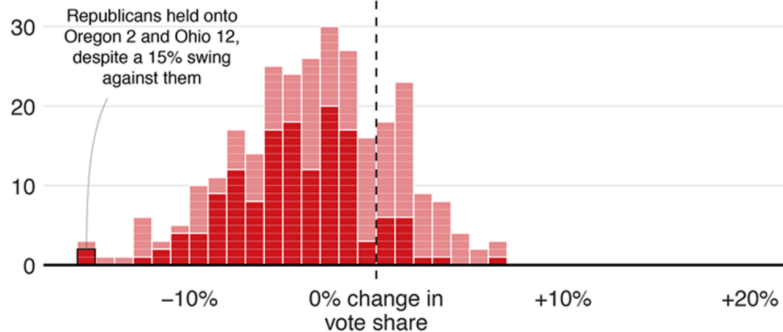
Blue wave

■ Won seat ■ Didn't win

Democrat candidates



Republican candidates

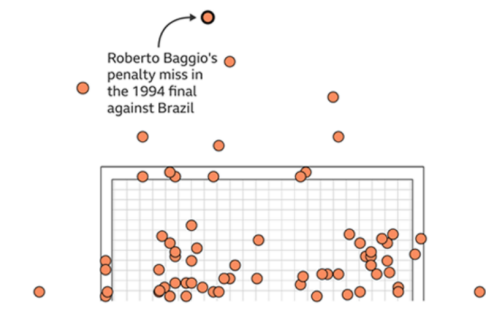


Source: AP, 19:01 ET

BBC

Where penalties are saved

World Cup shootout misses and saves, 1982-2014

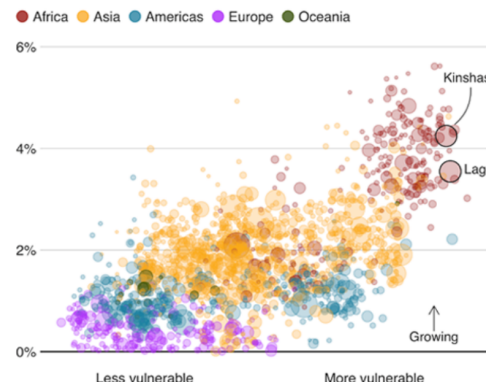


Source: Opta

BBC

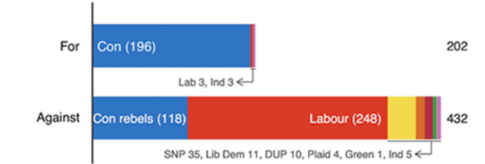
Fast-growing cities face worse climate risks

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

Theme options

Make theme adjustments with `theme()`

There are a billion options here!

```
theme_bw() +  
theme(legend.position = "bottom",  
      plot.title = element_text(face = "bold"),  
      panel.grid = element_blank(),  
      axis.title.y = element_text(face = "italic"))
```

Saving graphs

Use `ggsave()` to save a plot to your computer

Store plot as an object, feed it to `ggsave()`

```
my_plot <- ggplot(...)
```

```
ggsave("plot_name.pdf", my_plot, width = 5, height = 3.5)
```

```
ggsave("plot_name.png", my_plot, width = 5, height = 3.5)
```

Next up

**Transforming and
manipulating data with dplyr**