

# Mapping Data to Graphics

## Session 3

PMAP 8551/4551: Data Visualization with R  
Andrew Young School of Policy Studies  
Spring 2026

# Plan for today

Data, aesthetics, & the grammar of graphics

Grammatical layers

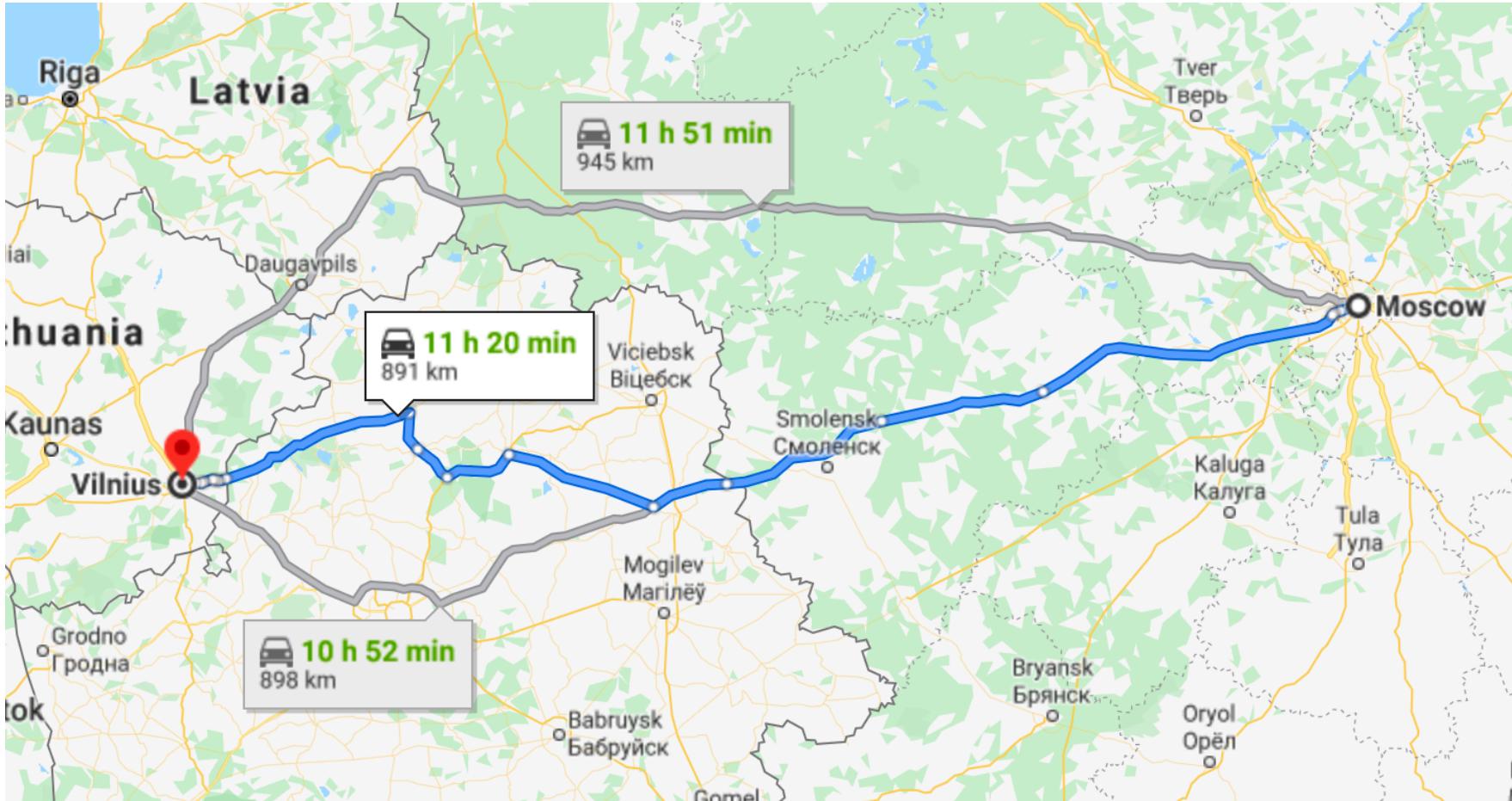
Aesthetics in extra dimensions

Tidy data

# Data, aesthetics, & the grammar of graphics

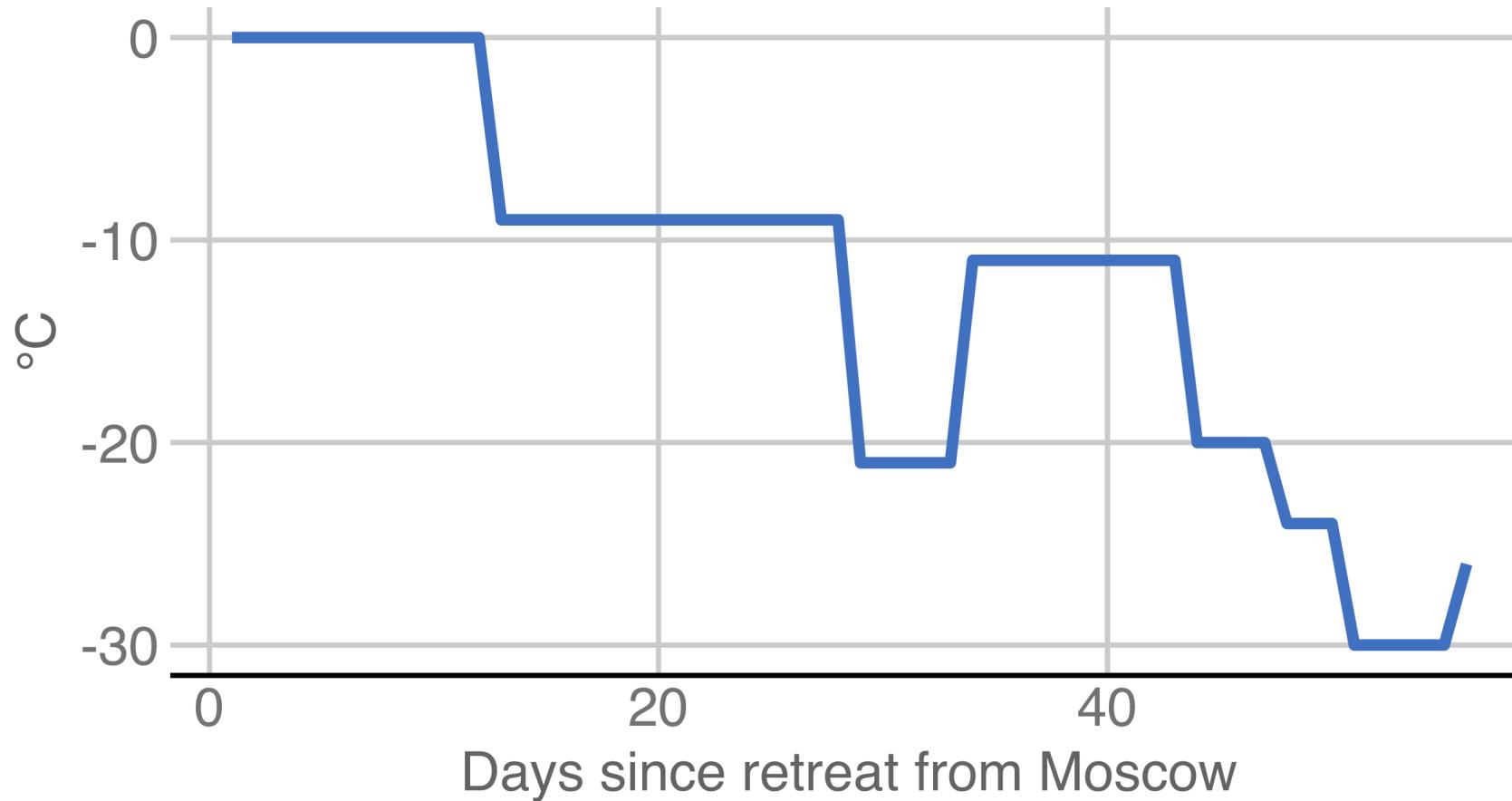


# Long distance!



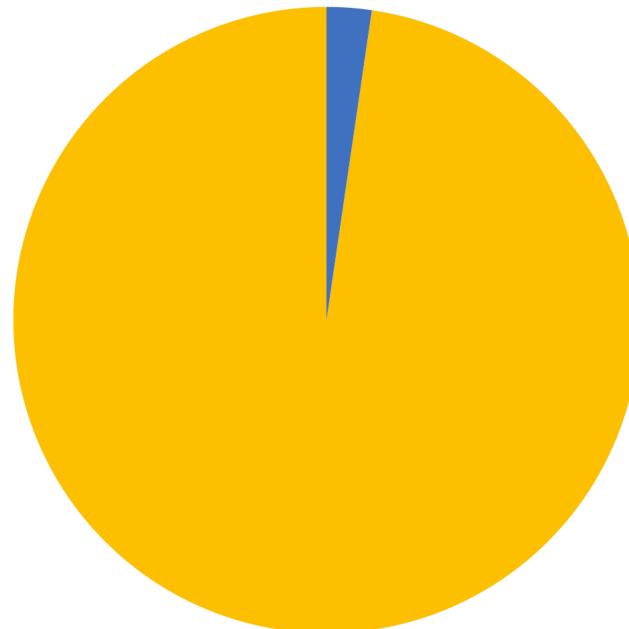
Moscow to Vilnius

# 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.

Dessinée 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 tracés des zones. Le rouge désigne les hommes qui ont été 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 Mobilow et qui rejoignirent Orsha en Witebsk, avaient toujours marché avec l'armée.

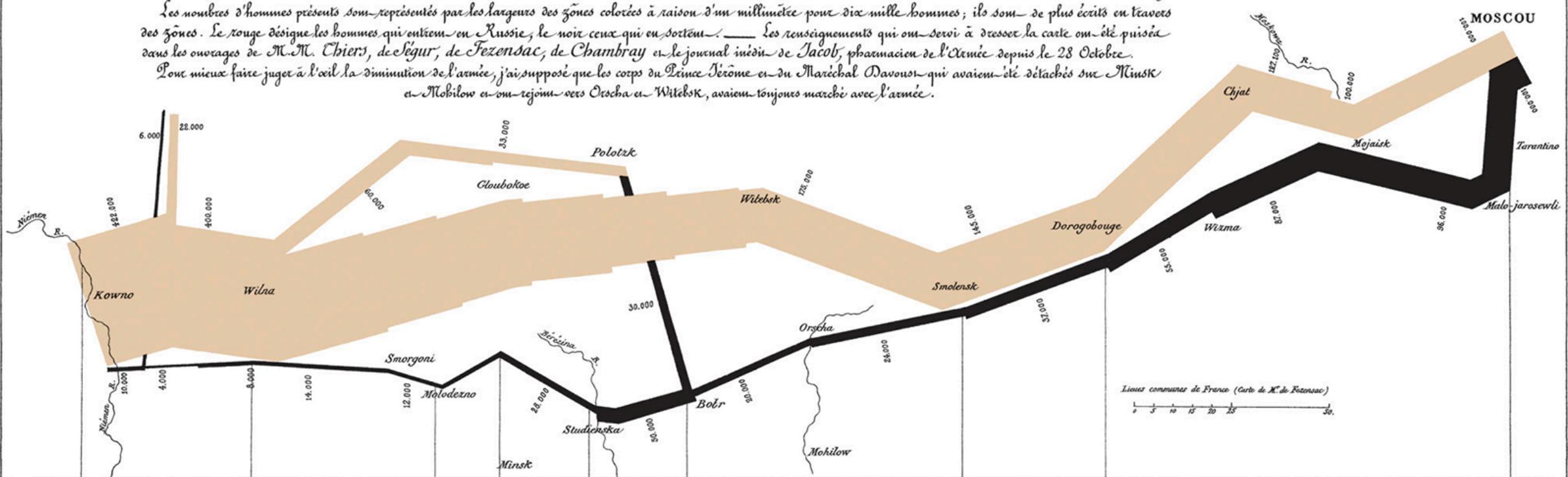
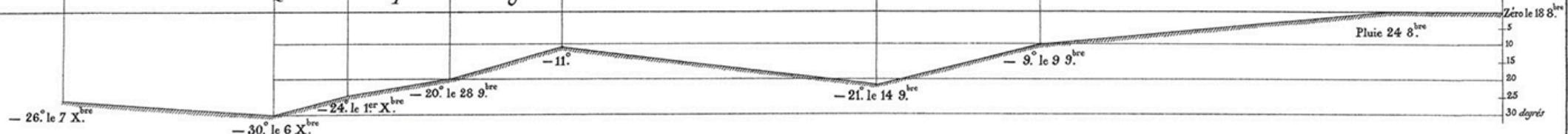
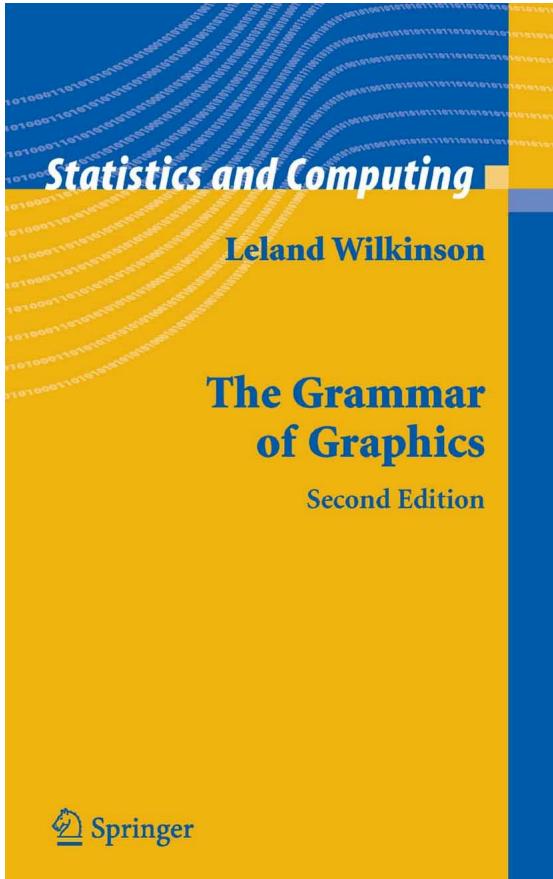


TABLEAU GRAPHIQUE de la température en degrés du thermomètre de Réaumur au dessous de zéro.

Les Cosaques passent au galop  
le Niemen gelé.



# 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

---

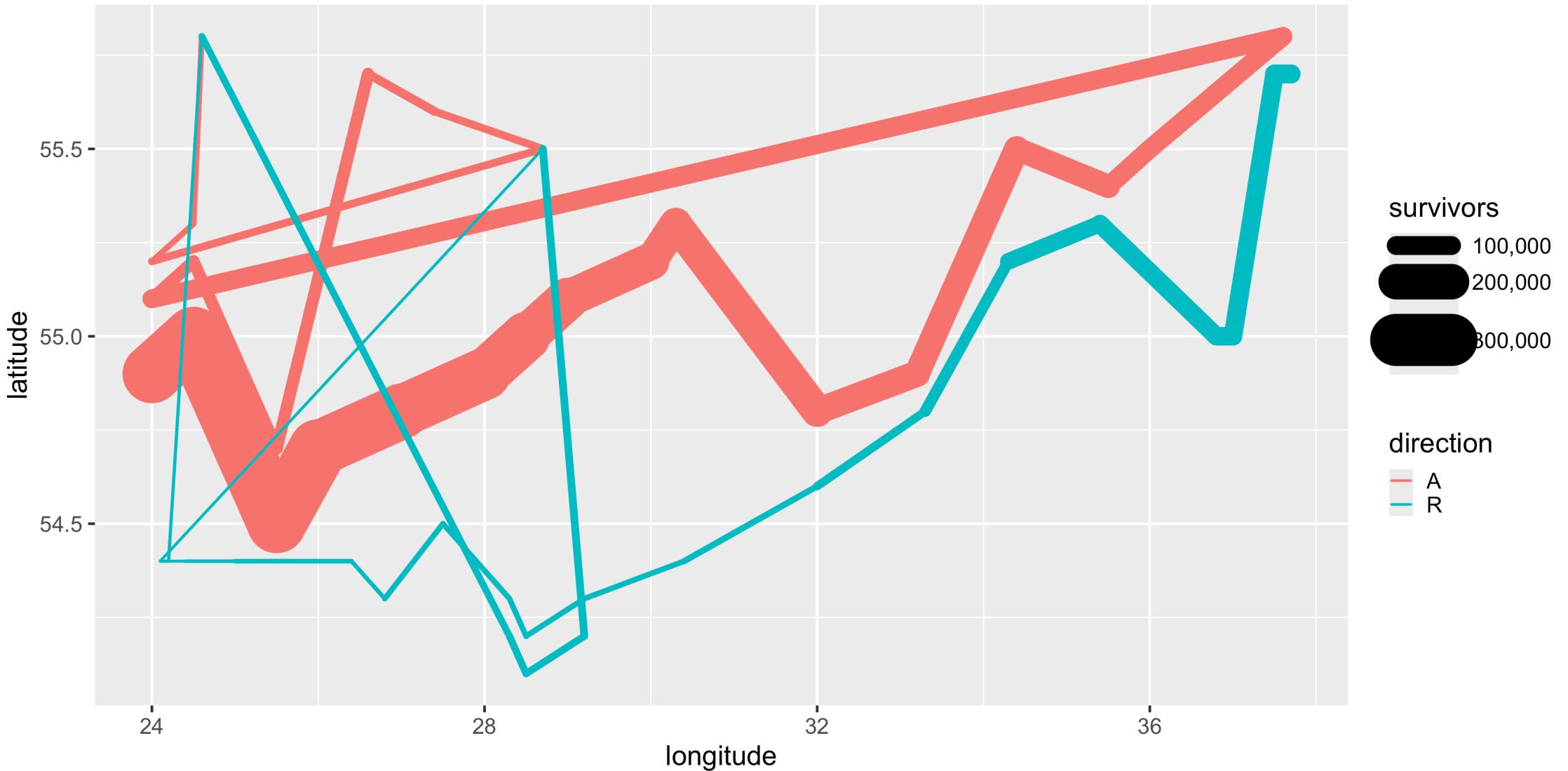
Data	aes()	geom
Longitude	x	geom_point()
Latitude	y	geom_point()
Army size	size	geom_path()
Army direction	color	geom_path()
Date	x	geom_line() + geom_text()
Temperature	y	geom_line() + geom_text()

---

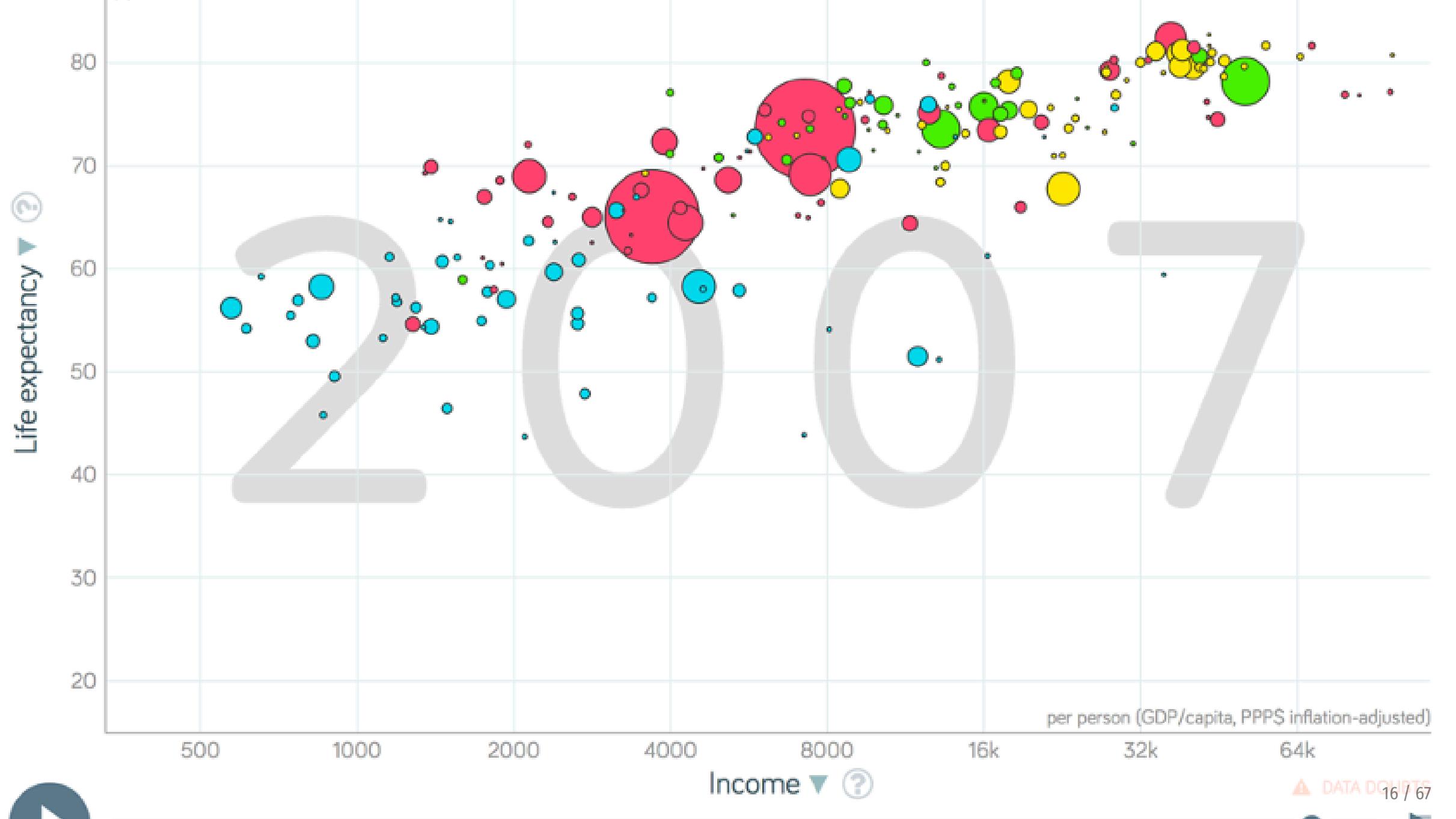
# `ggplot()` template

## This is a dataset named `troops`:

<b>longitude</b>	<b>latitude</b>	<b>direction</b>	<b>survivors</b>
24	54.9	A	340000
24.5	55	A	340000
...	...	...	...







# Mapping data to aesthetics

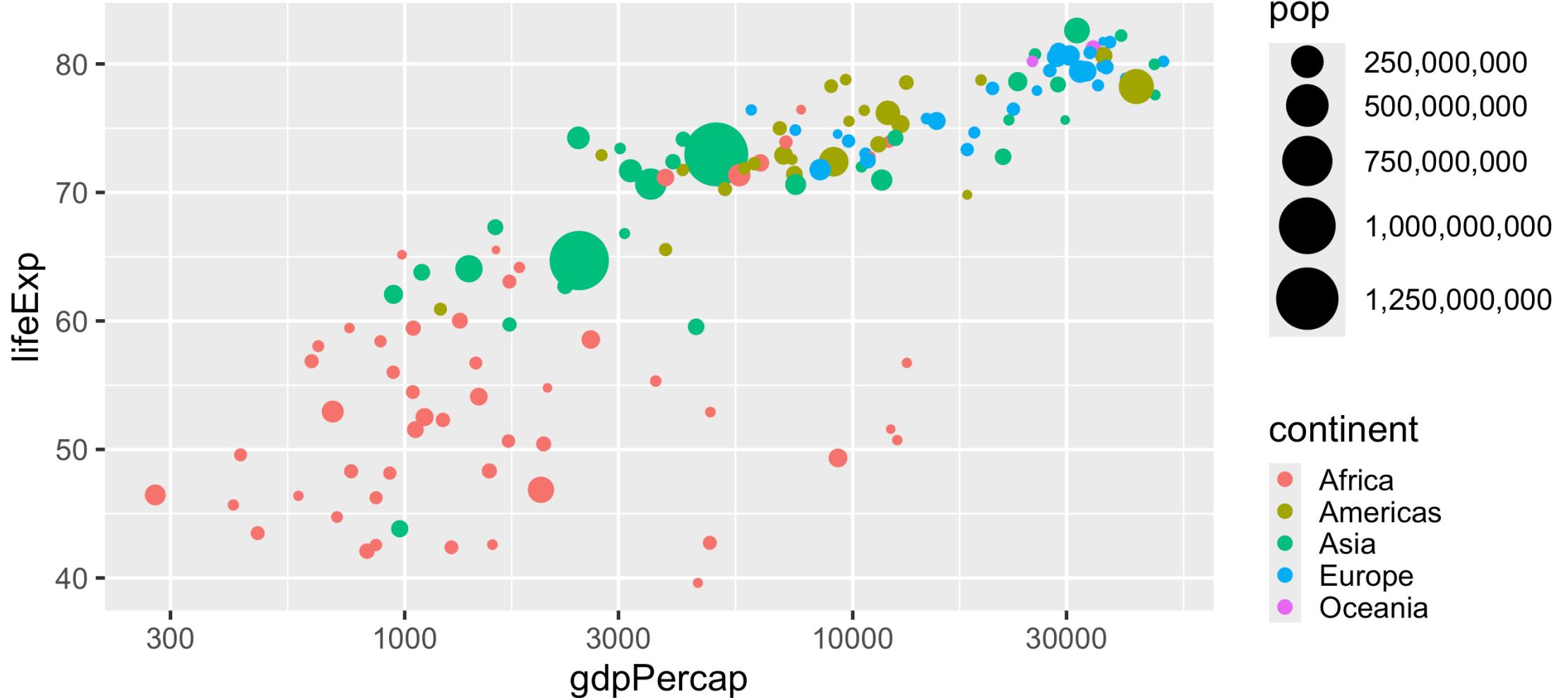
---

Data	aes()	geom
Wealth (GDP/capita)	x	geom_point()
Health (Life expectancy)	y	geom_point()
Continent	color	geom_point()
Population	size	geom_point()

This is a dataset named `gapminder_2007`:

<b>country</b>	<b>continent</b>	<b>gdpPercap</b>	<b>lifeExp</b>	<b>pop</b>
Afghanistan	Asia	974.5803384	43.828	31889923
Albania	Europe	5937.029526	76.423	3600523
...	...	...	...	...

# Health and wealth



# Grammatical layers

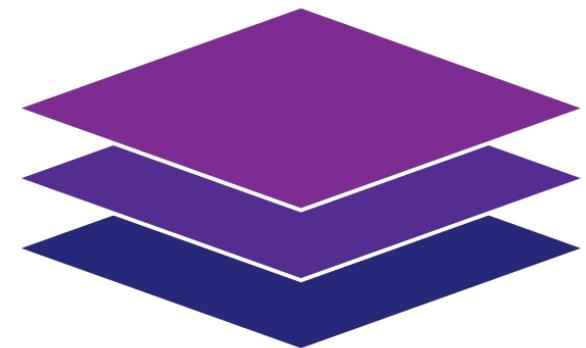
# Grammar components as layers

So far we know about data,  
aesthetics, and geometries

Think of these  
components as layers

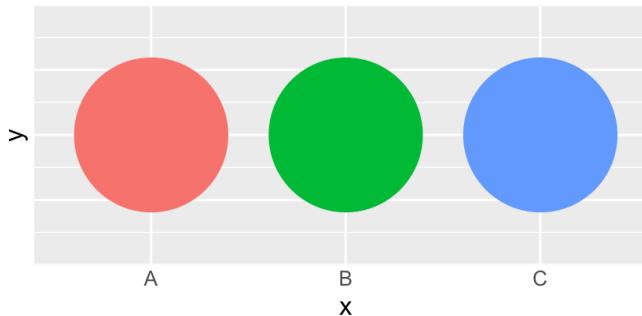
Add them to foundational  
`ggplot()` with +

Geometries  
Aesthetics  
Data

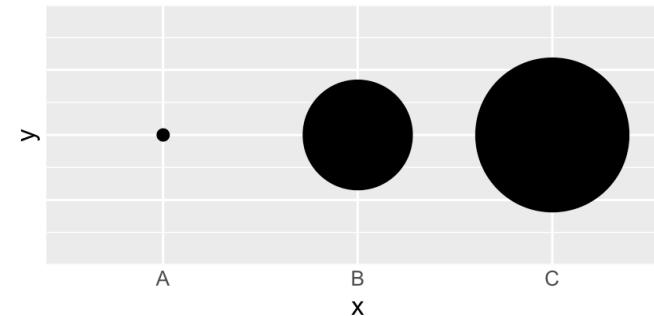


# Possible aesthetics

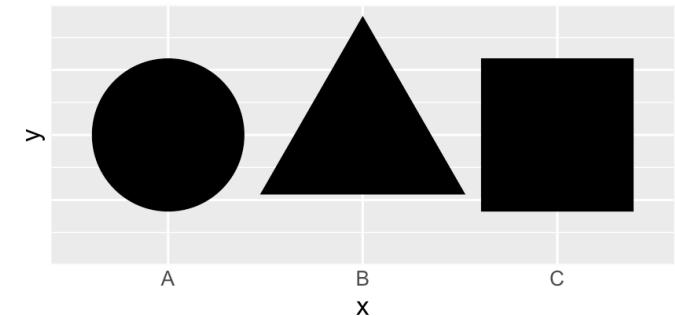
**color (discrete)**



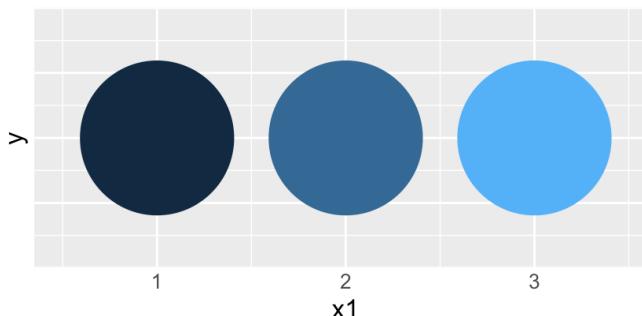
**size**



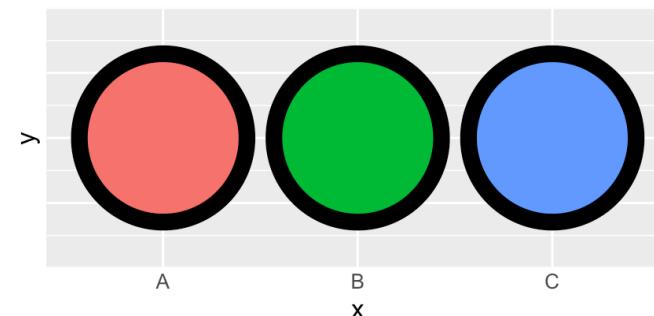
**shape**



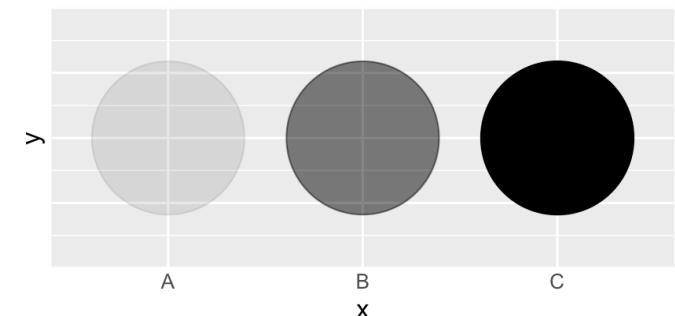
**color (continuous)**



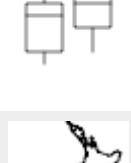
**fill**



**alpha**



# Possible geoms

Example geom	What it makes
	<code>geom_col()</code> Bar charts
	<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 and each class session will cover different ones.

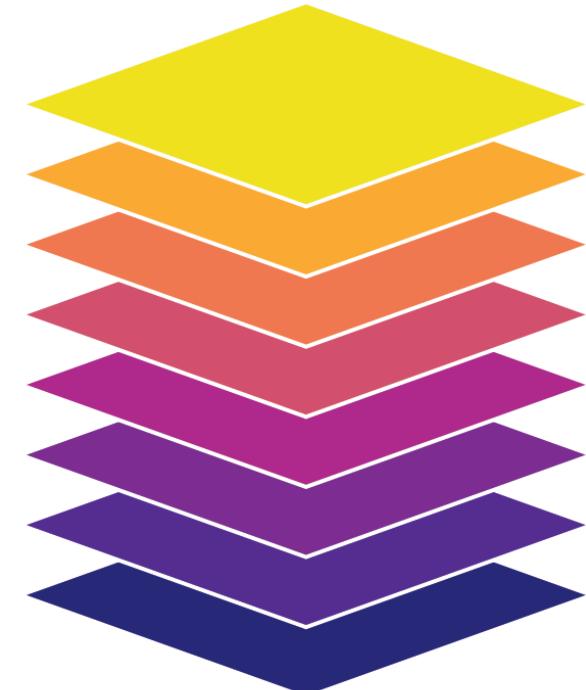
See [the {ggplot2} documentation](#) for complete examples of all the different geom layers

# Additional layers

There are many other grammatical layers we can use to describe graphs!

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

Theme  
Labels  
Coordinates  
Facets  
Scales  
Geometries  
Aesthetics  
Data



# Scales

**Scales change the properties of the variable mapping**

## Example layer

`scale_x_continuous()`

## What it does

Make the x-axis continuous

`scale_x_continuous(breaks = 1:5)`

Manually specify axis ticks

`scale_x_log10()`

Log the x-axis

`scale_color_gradient()`

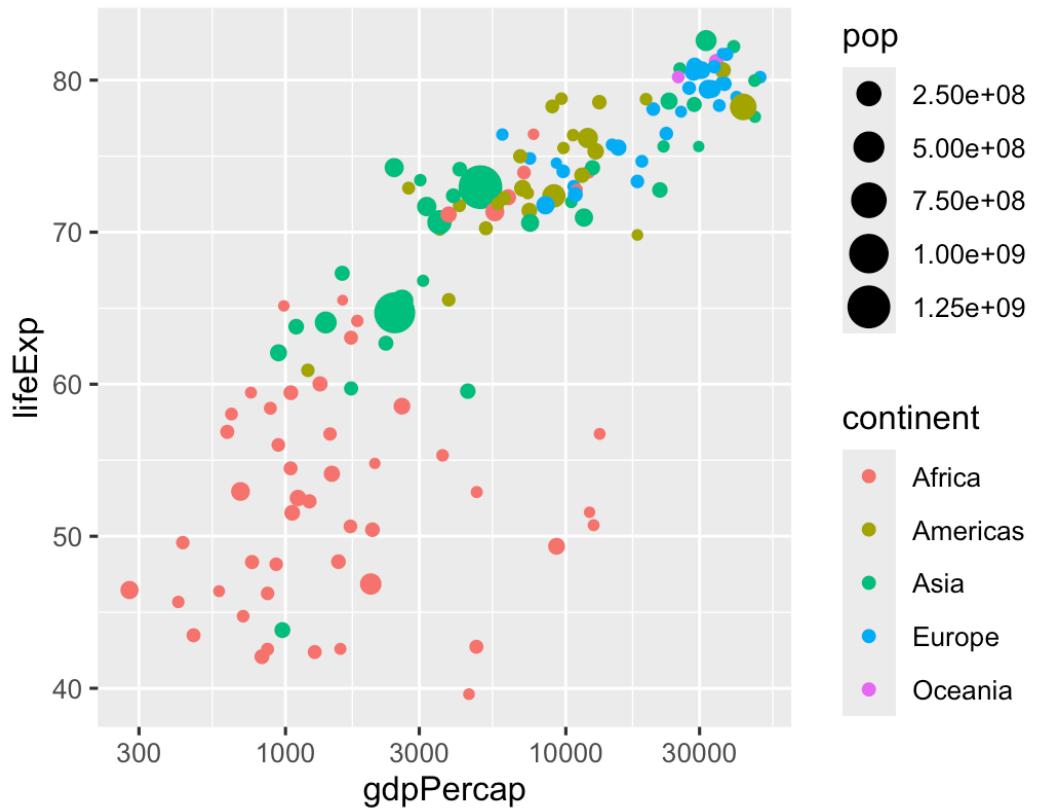
Use a gradient

`scale_fill_viridis_d()`

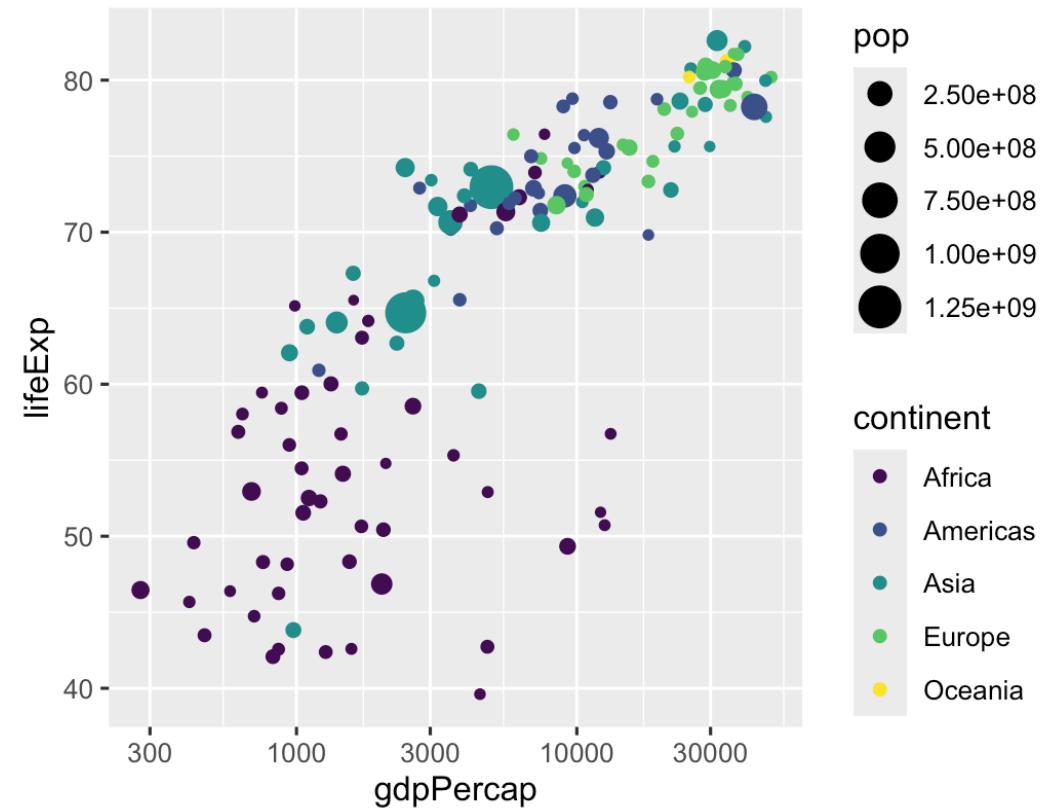
Fill with discrete viridis colors

# Scales

`scale_x_log10()`



`scale_color_viridis_d()`



# Facets

**Facets show subplots for different subsets of data**

## Example layer

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

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

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

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

## What it does

Plot for each continent

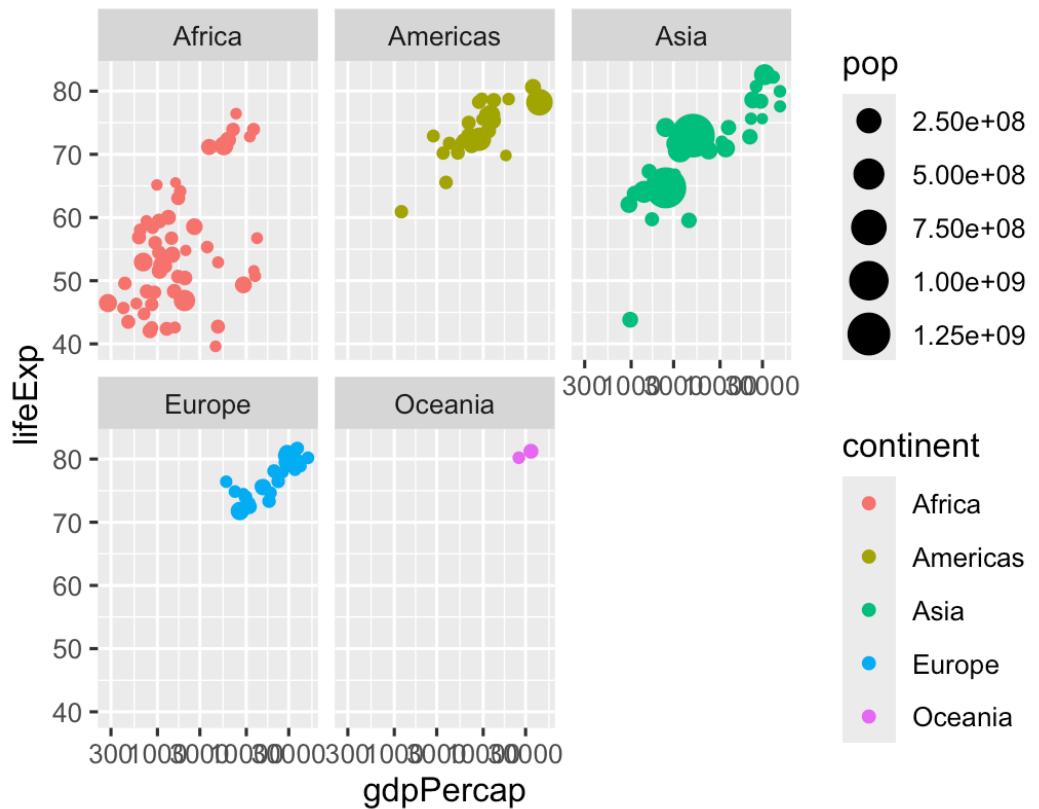
Plot for each continent/year

Put all facets in one column

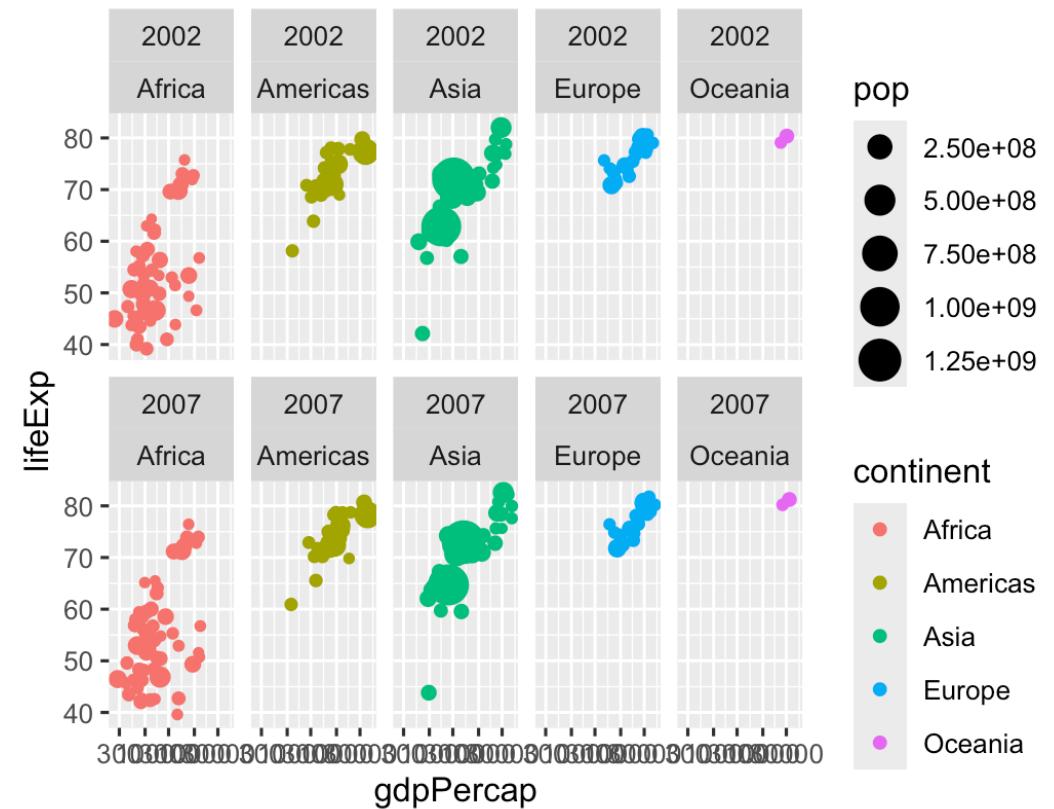
Put all facets in one row

# Facets

`facet_wrap(vars(continent))`



`facet_wrap(vars(continent, year))`



# Coordinates

## Change the coordinate system

### Example layer

```
coord_cartesian()
```

### What it does

Plot for each continent

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

Zoom in where y is 1–10

```
coord_flip()
```

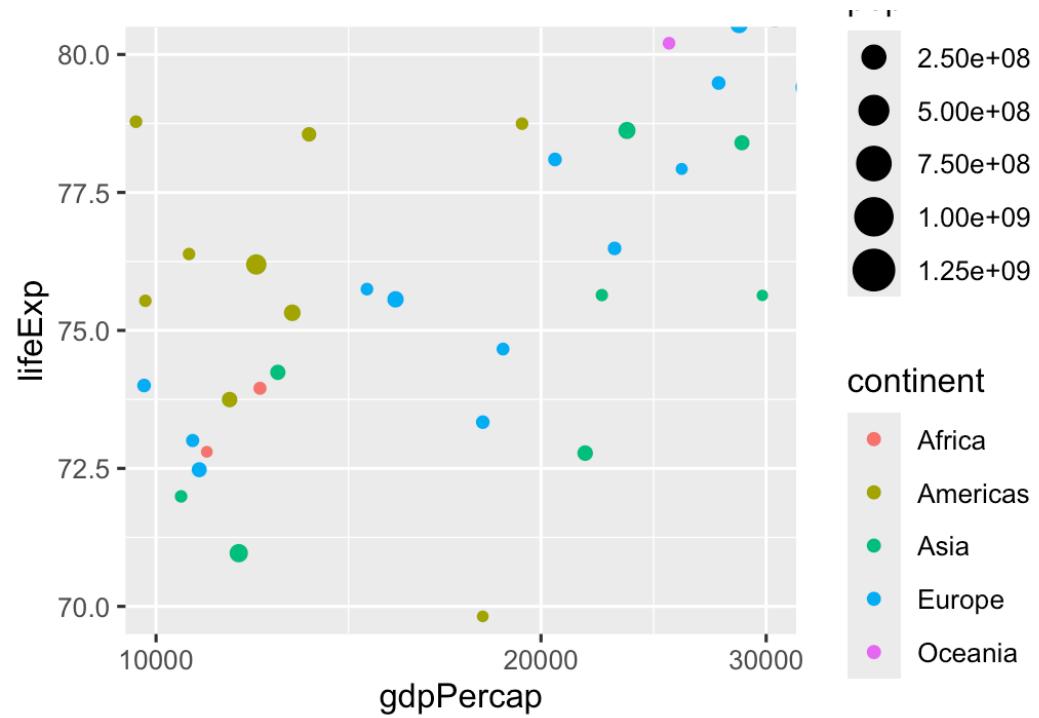
Switch x and y

```
coord_polar()
```

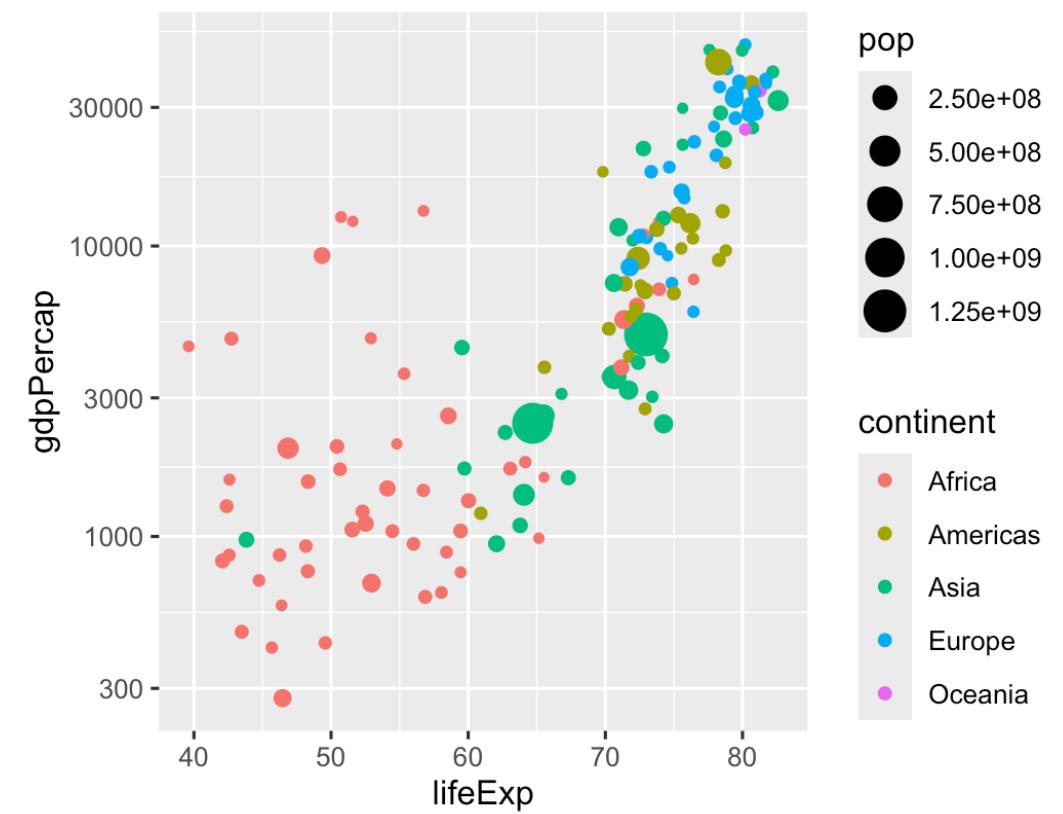
Use circular polar system

# 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

```
labs(title = "Neat title")
```

## What it does

Title

```
labs(caption = "Something")
```

Caption

```
labs(y = "Something")
```

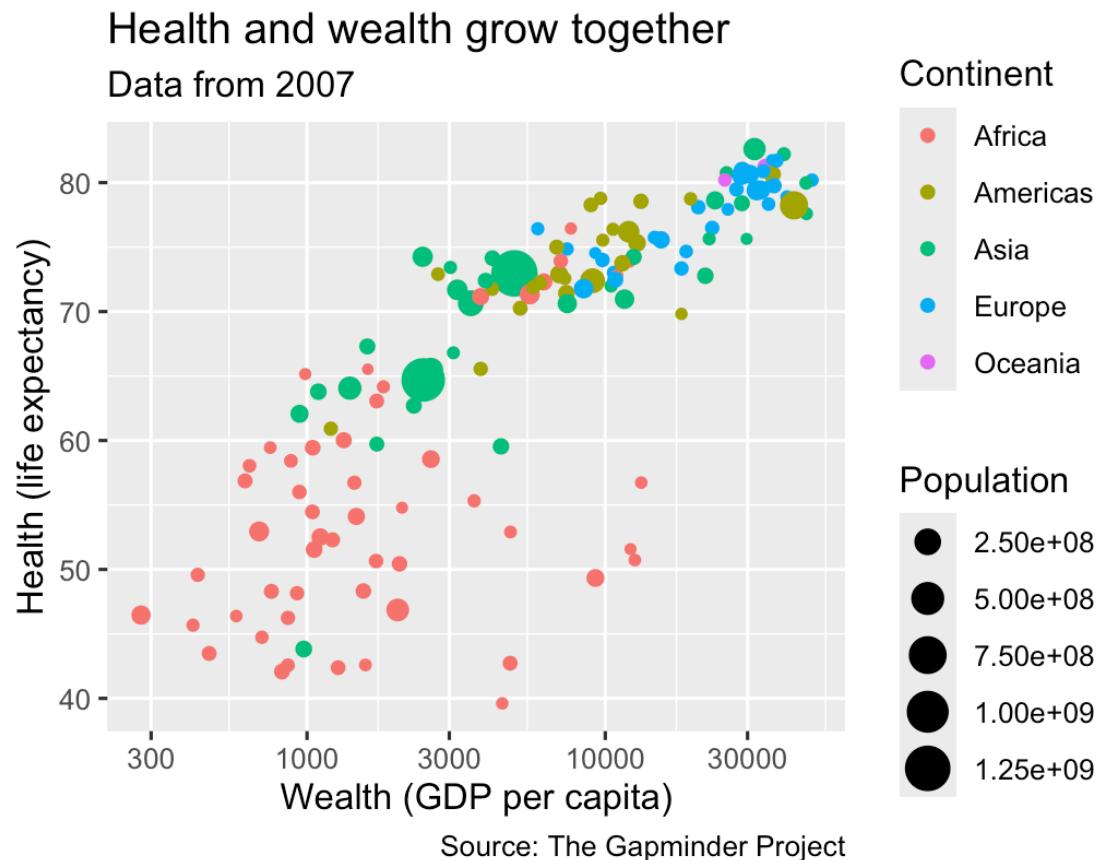
y-axis

```
labs(size = "Population")
```

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

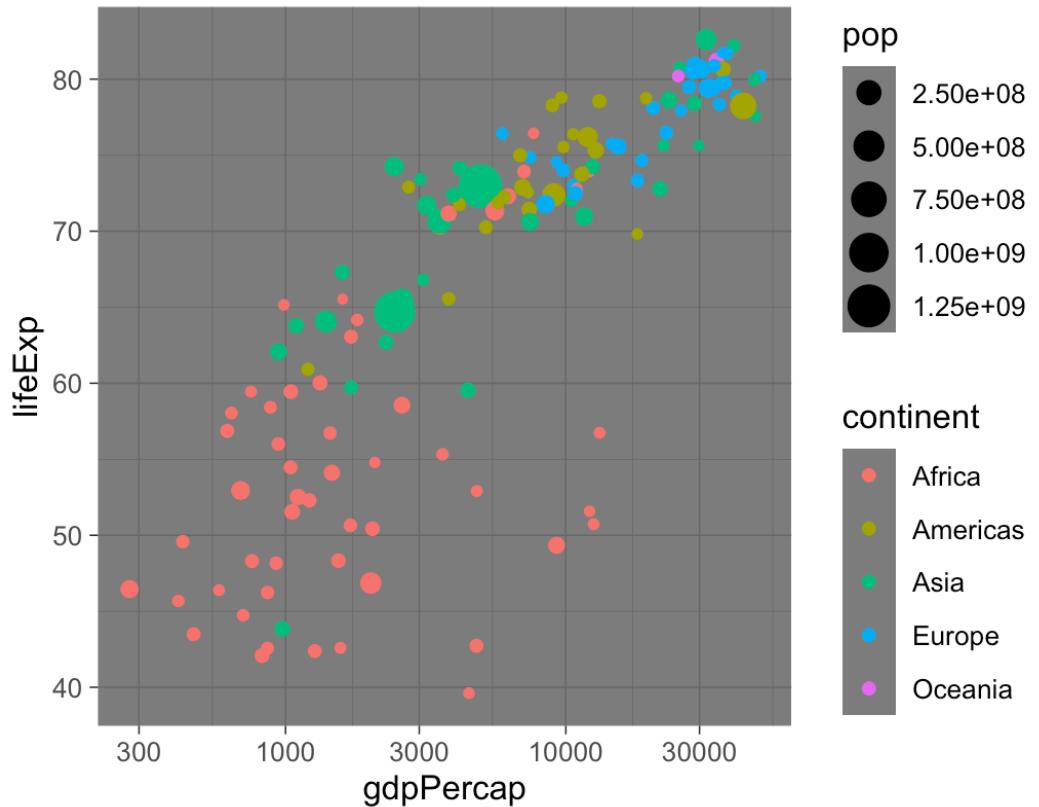
There are many built-in themes

---

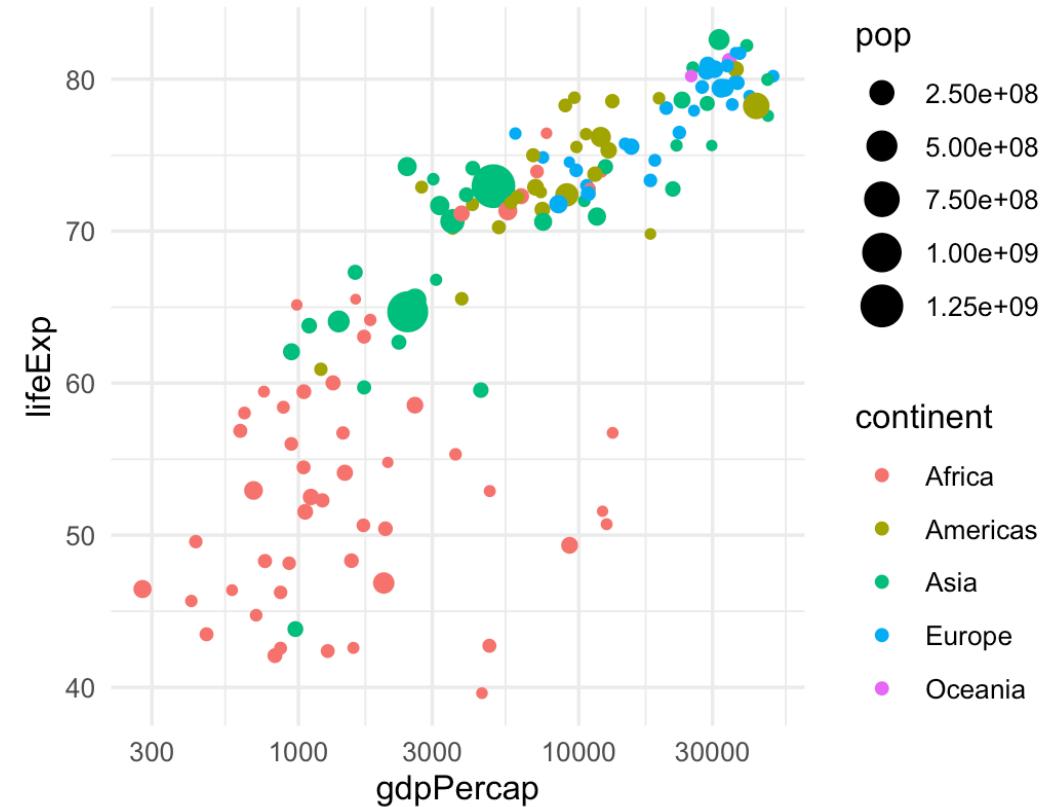
<b>Example layer</b>	<b>What it does</b>
theme_grey()	Default grey background
theme_bw()	Black and white
theme_dark()	Dark
theme_minimal()	Minimal

# Theme

`theme_dark()`



`theme_minimal()`



# Theme

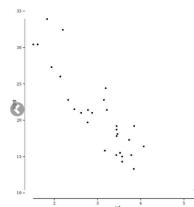
There are collections of pre-built themes online,  
like the `{ggthemes}` package

## ggthemes



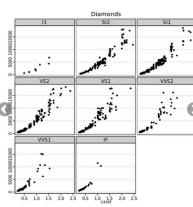
`theme_wsj`

Wall Street Journal theme



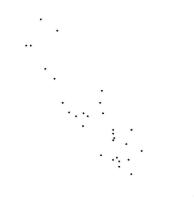
`theme_tufte`

Tufte Maximal Data, Minimal Ink Theme



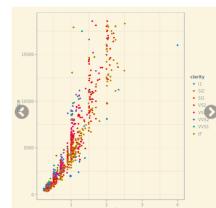
`theme_stata`

Themes based on Stata graph schemes



`theme_solid`

Theme with nothing other than a background color



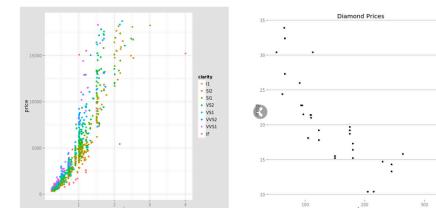
`theme_solarized`

ggplot color themes based on the Solarized palette



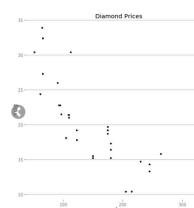
`theme_map`

Clean theme for maps



`theme_igray`

Inverse gray theme

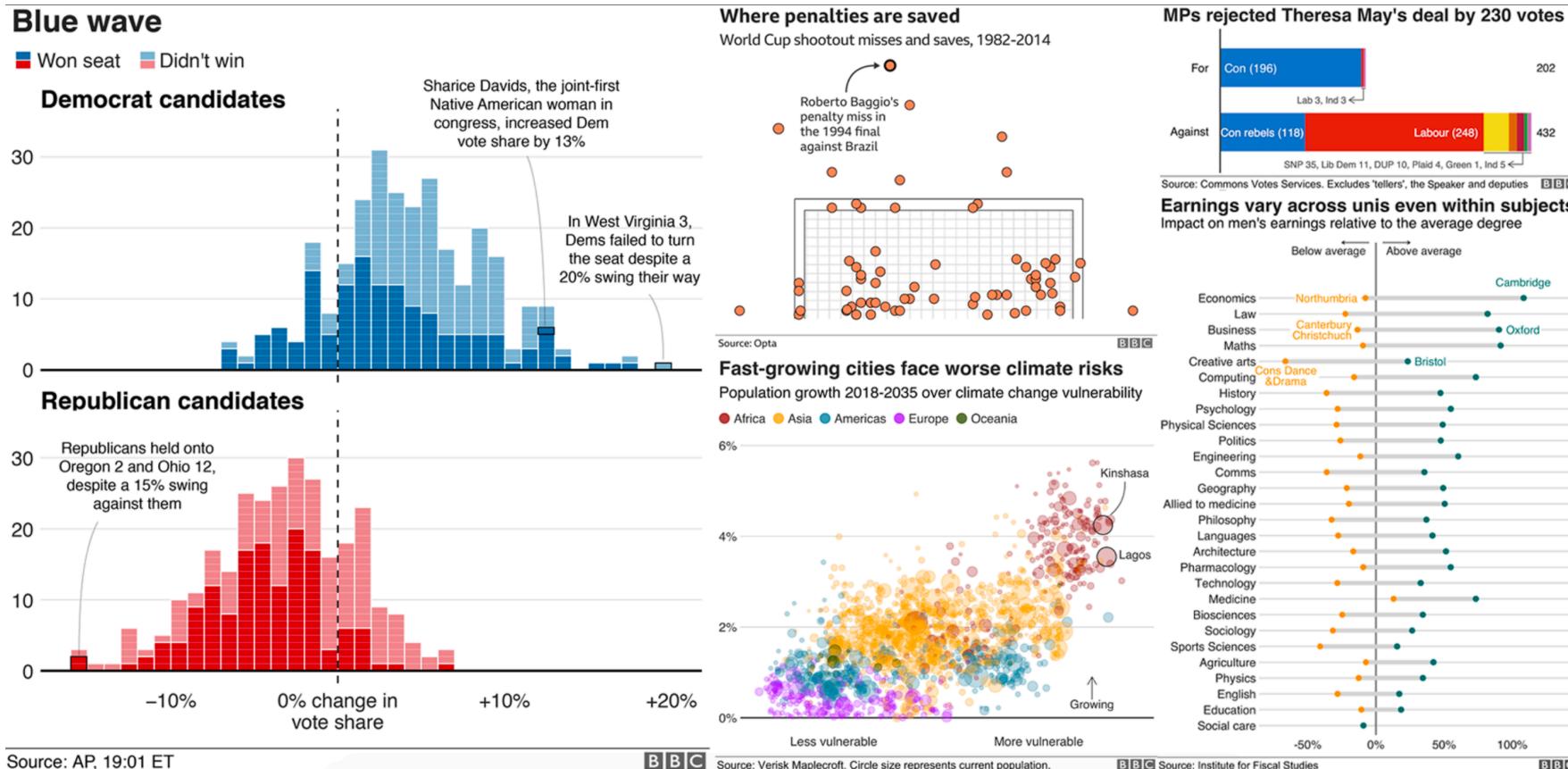


`theme_hc`

Highcharts JS theme

# Theme

Organizations often make their own custom themes, like the BBC



# Theme options

Make theme adjustments with `theme()`

There are a billion options here!  
We have a whole class session dedicated to this!

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

# So many possibilities!



These were just a few examples of layers!

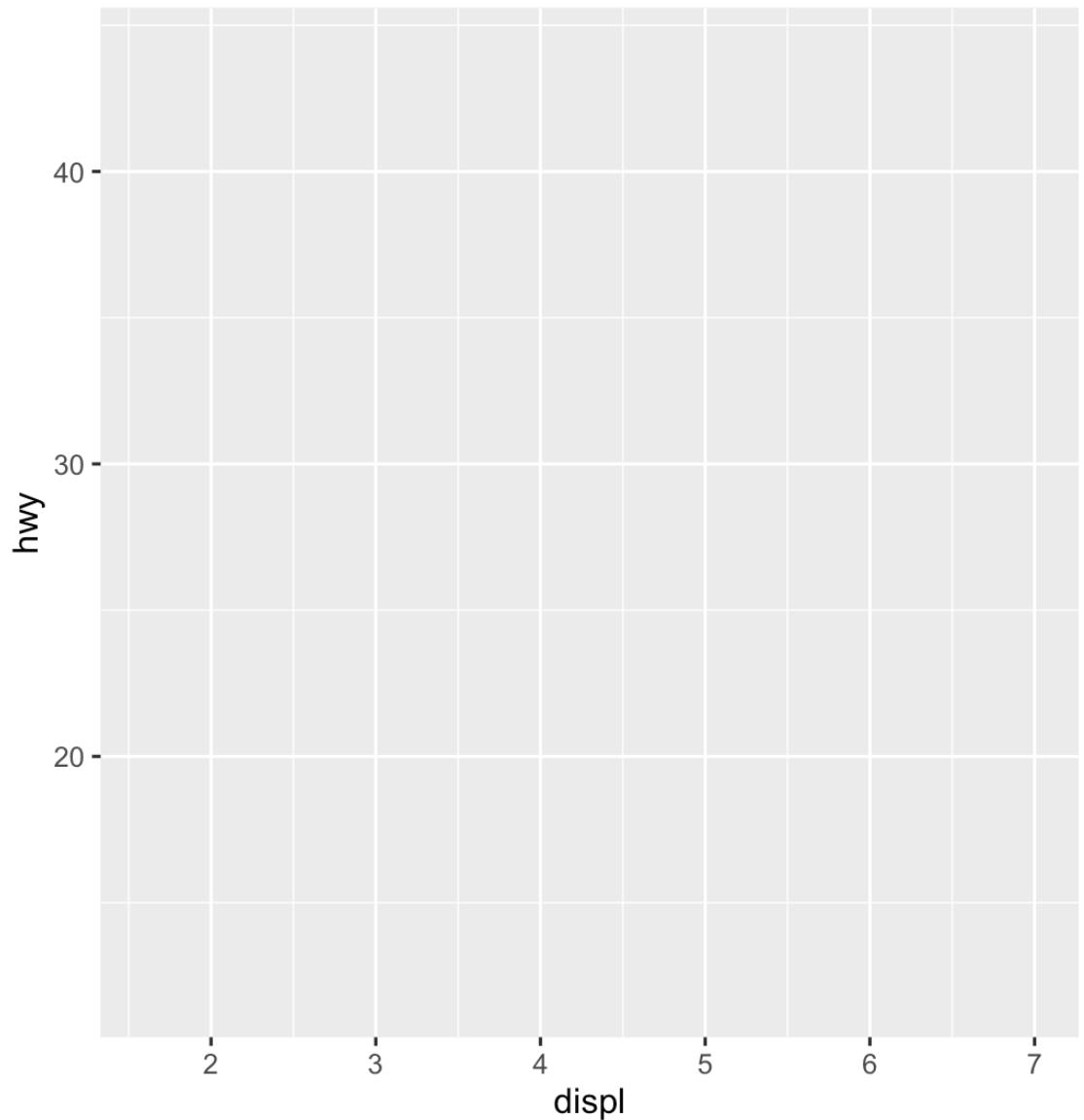
See the `{ggplot2}` documentation for complete examples of everything you can do

# Putting it all together

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

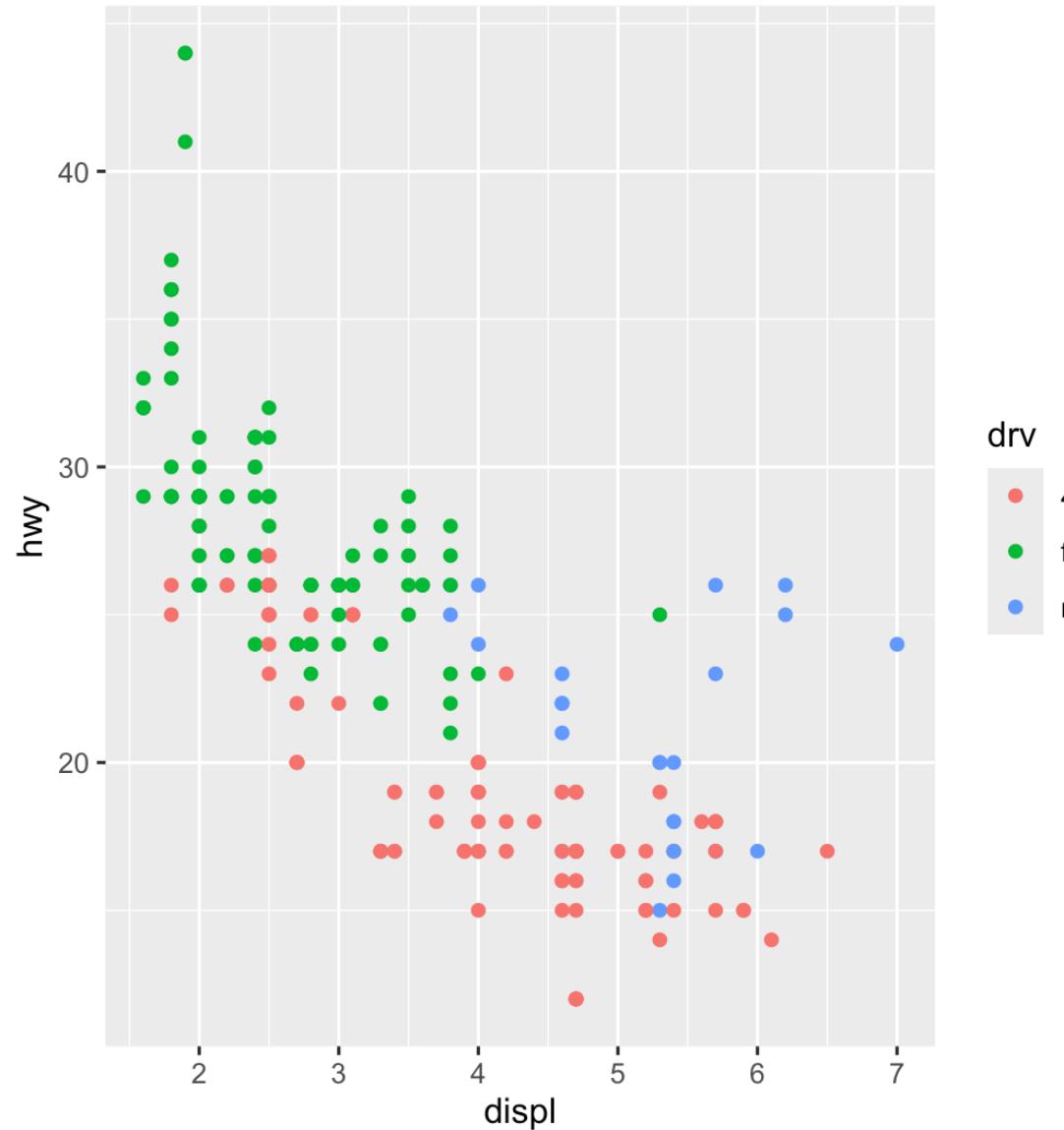
# Start with data and aesthetics

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



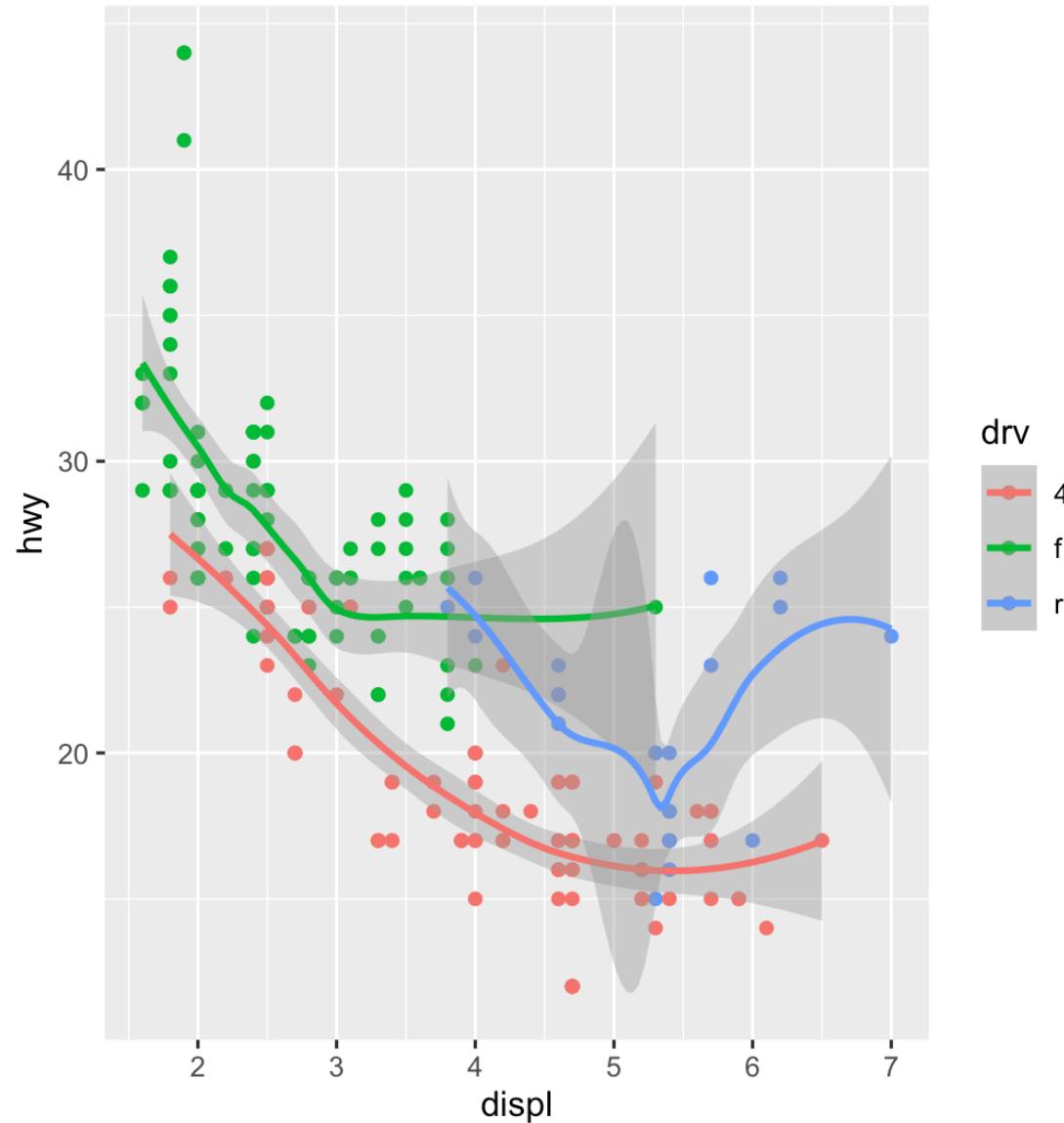
# Add a point geom

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



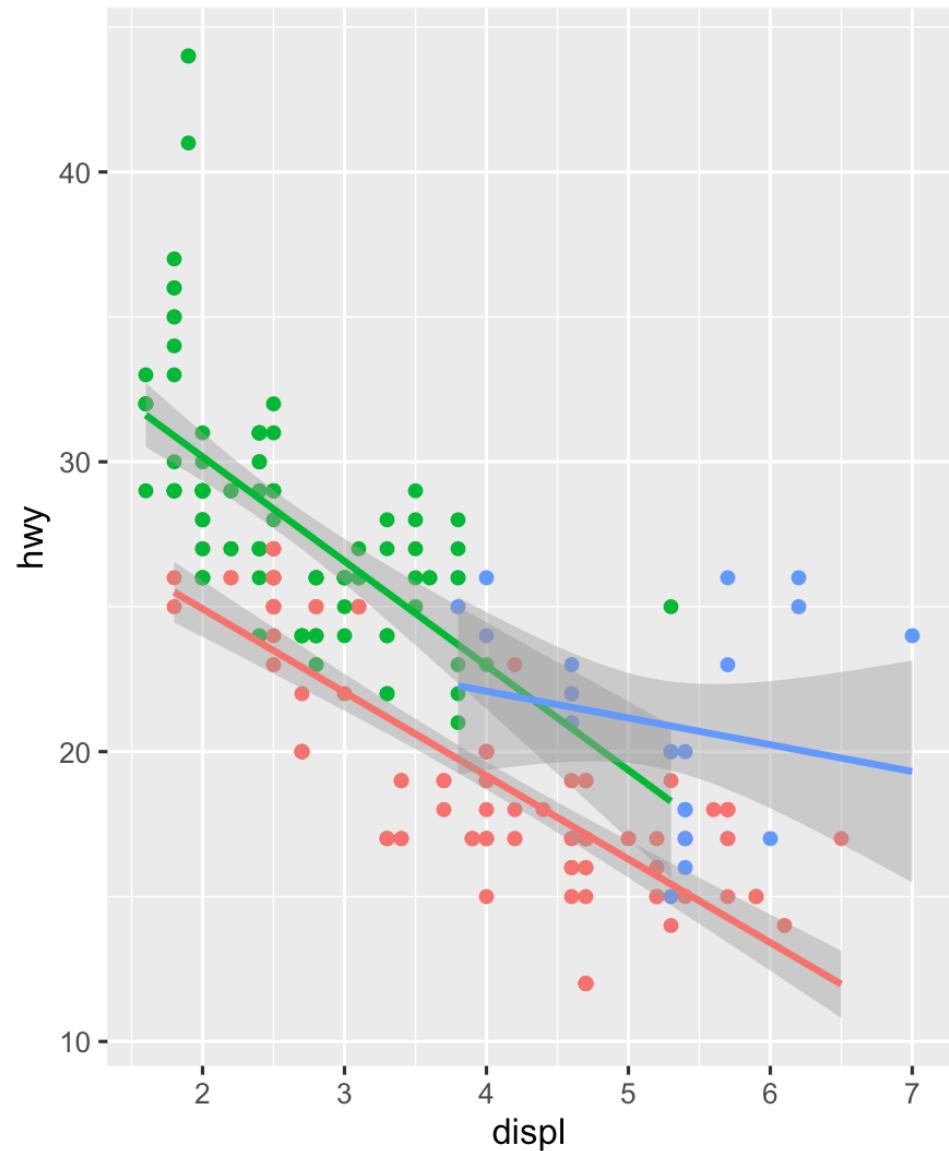
# Add a smooth geom

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



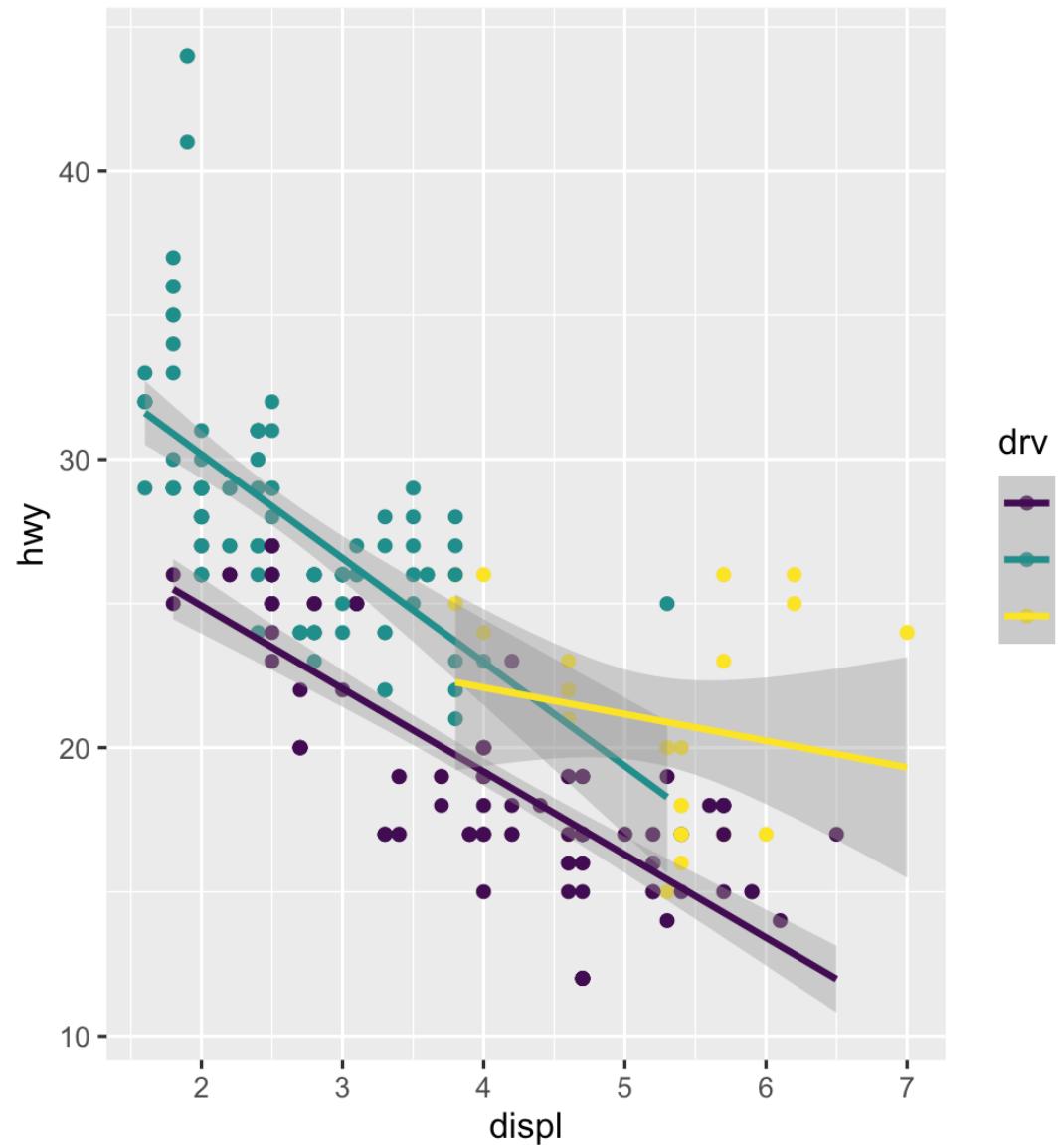
# Make it straight

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



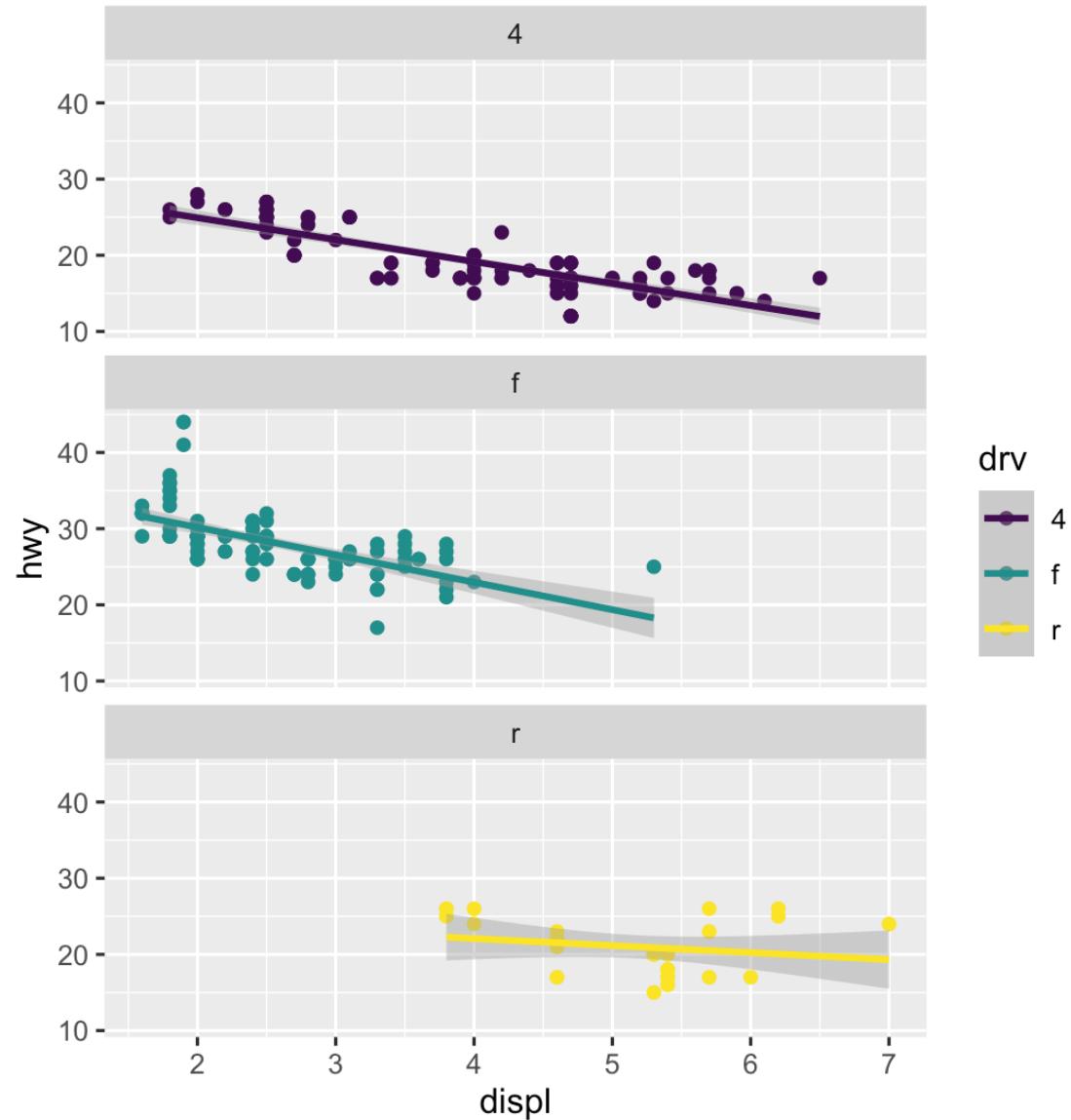
# Use a viridis color scale

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



# Facet by drive

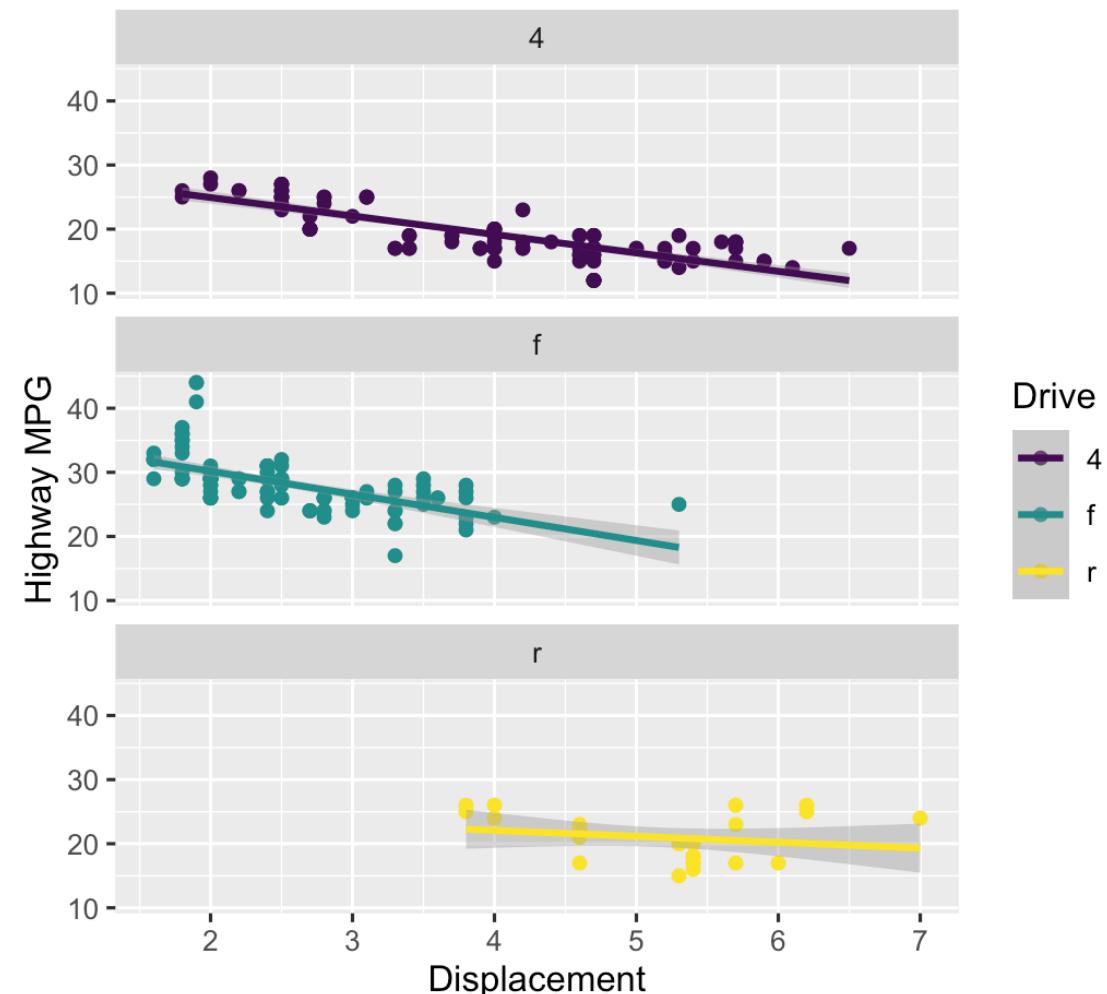
```
ggplot(data = mpg,  
       mapping = aes(x = displ,  
                      y = hwy,  
                      color = drv)) +  
  
  geom_point() +  
  geom_smooth(method = "lm") +  
  scale_color_viridis_d() +  
  facet_wrap(vars(drv), ncol = 1)
```



# 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")
```

Heavier cars get lower mileage  
Displacement indicates weight(?)

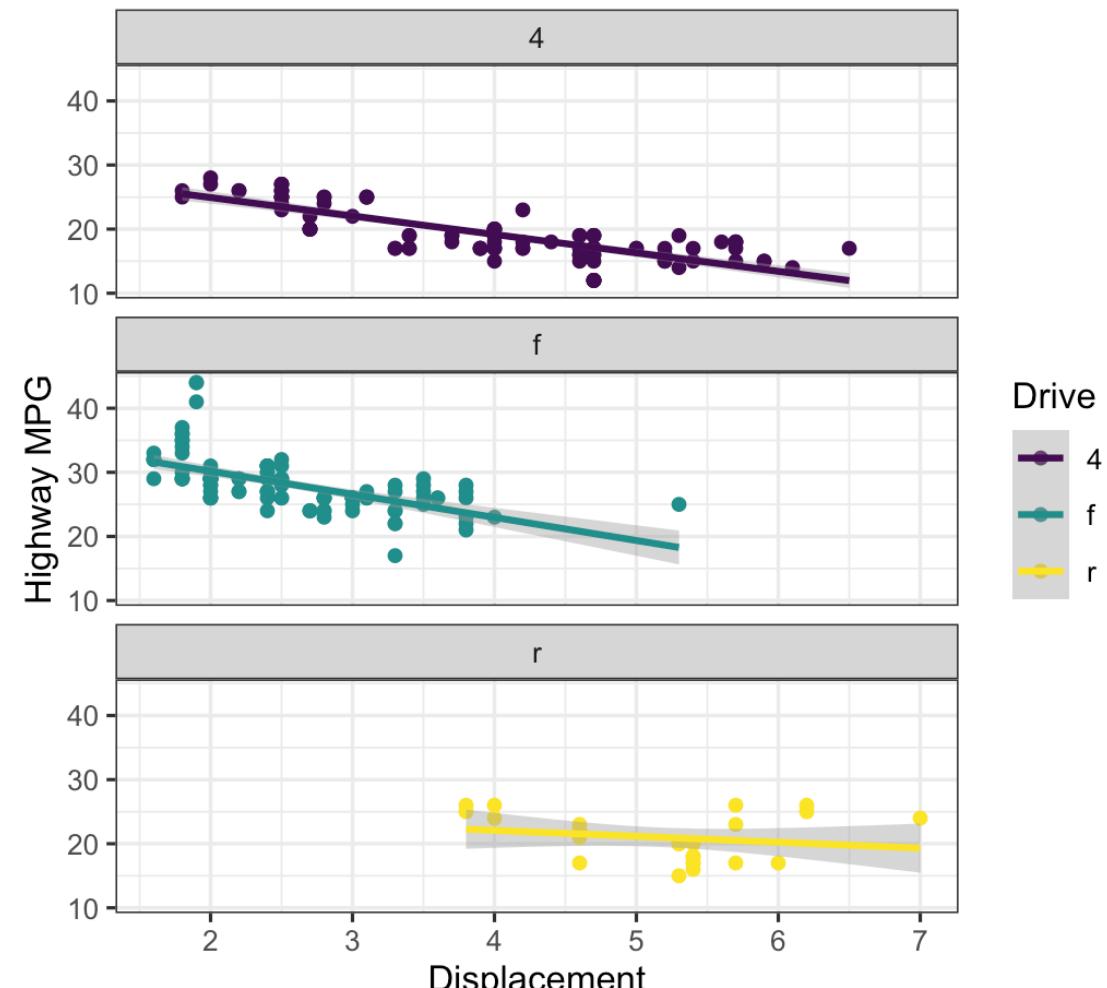


I know nothing about cars

# 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()
```

Heavier cars get lower mileage  
Displacement indicates weight(?)



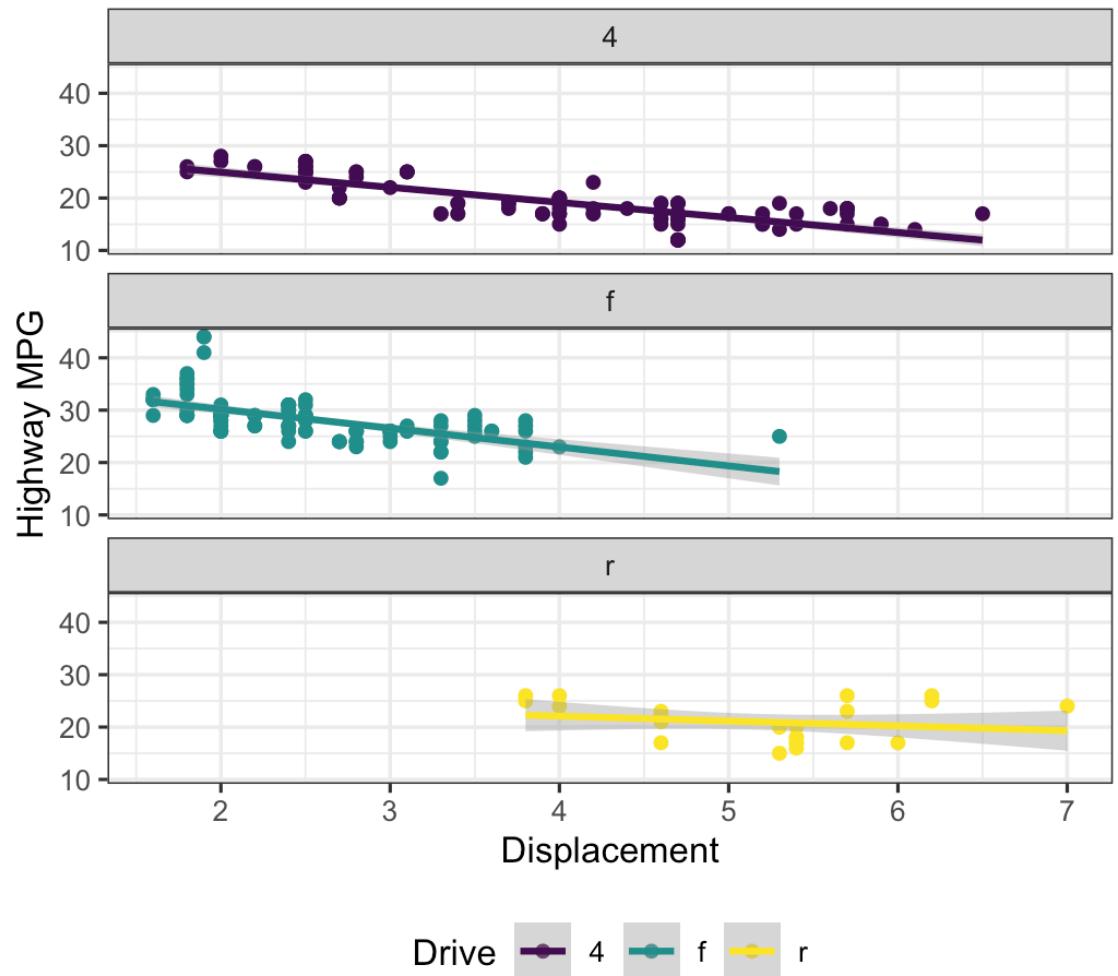
I know nothing about cars

# 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(?)



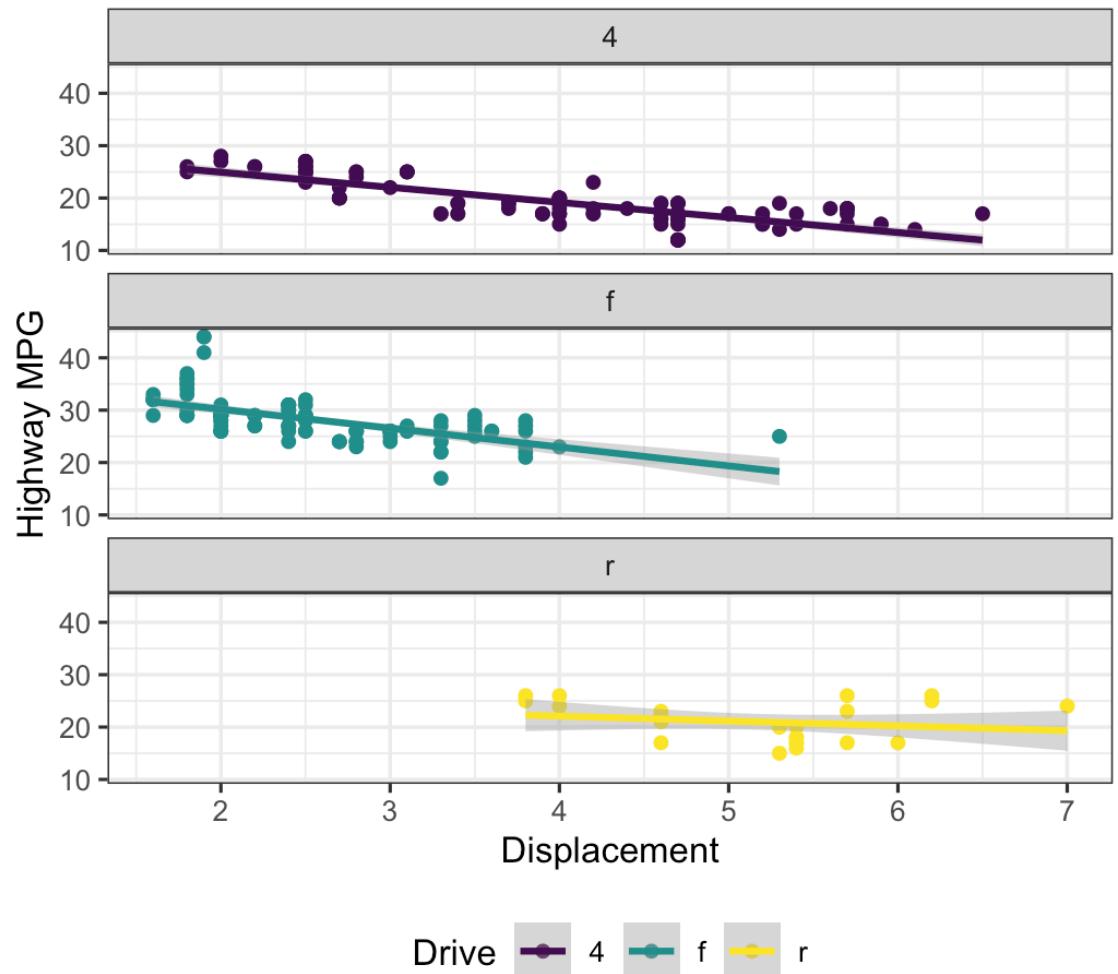
I know nothing about cars

# 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(?)

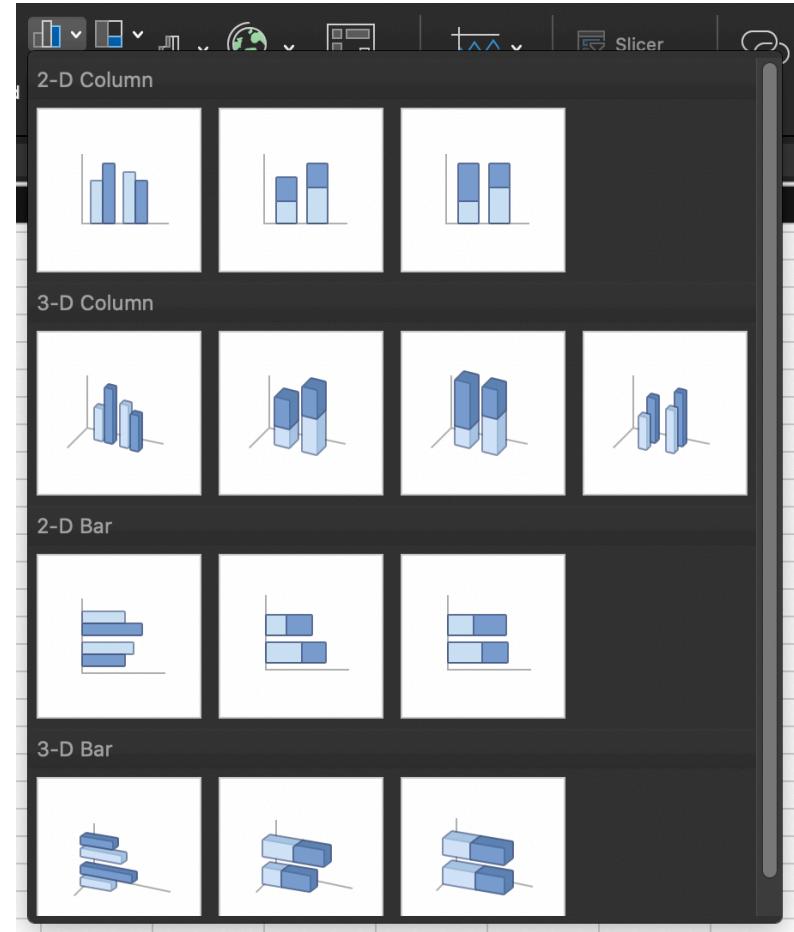


I know nothing about cars

# 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



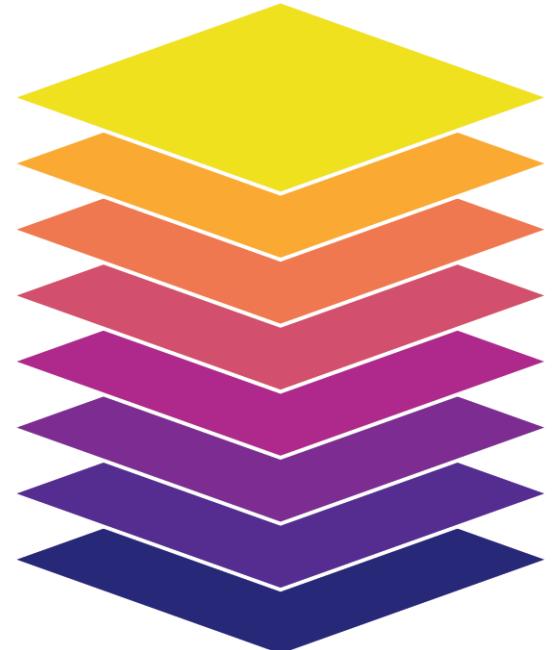
# A true grammar

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()`)

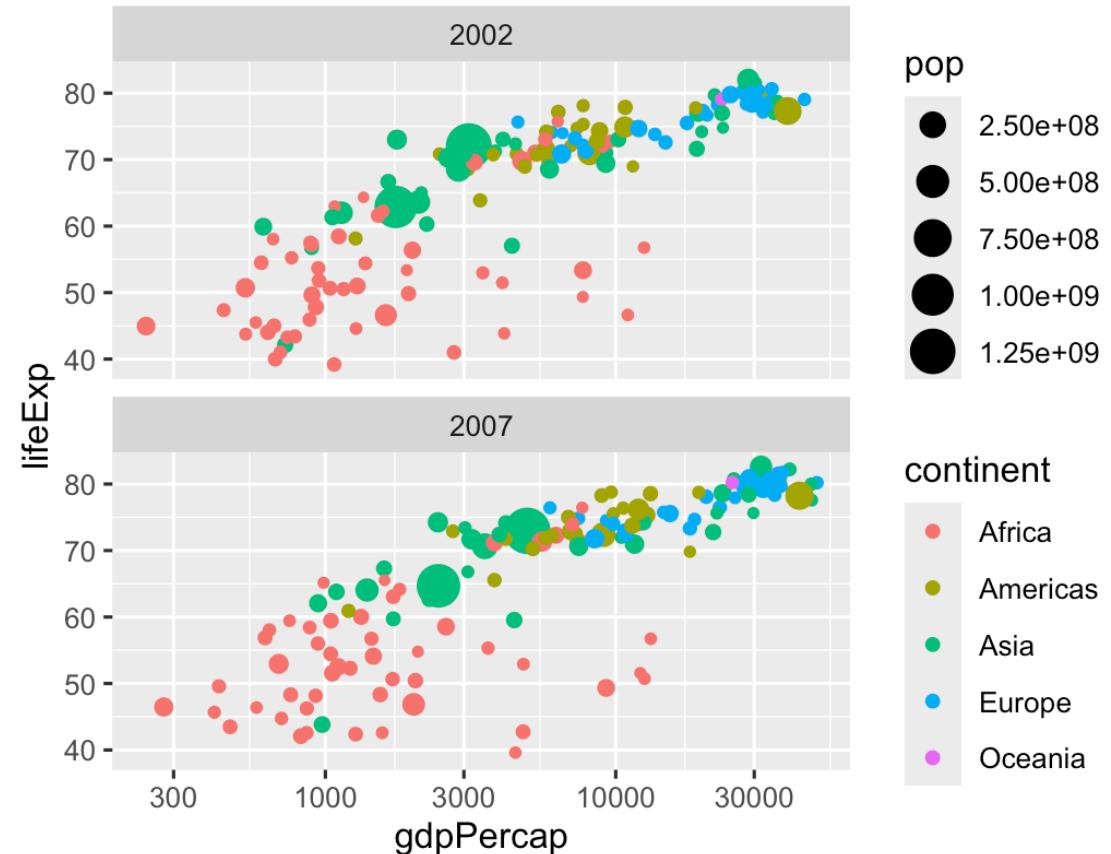
Theme  
Labels  
Coordinates  
Facets  
Scales  
Geometries  
Aesthetics  
Data



# 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

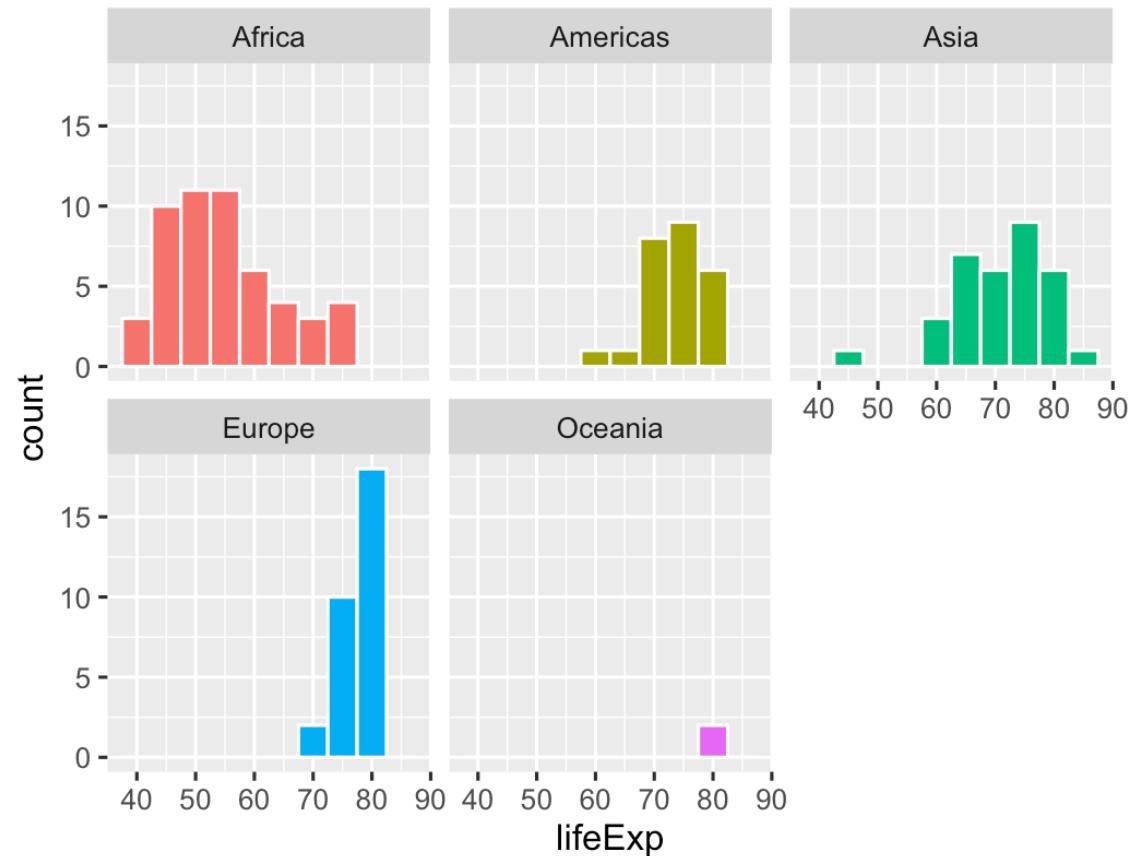
```
ggplot(data = filter(gapminder, year %in% c(2002, 2007)),  
       mapping = aes(x = gdpPercap,  
                     y = lifeExp,  
                     color = continent,  
                     size = pop)) +  
  geom_point() +  
  scale_x_log10() +  
  facet_wrap(vars(year), ncol = 1)
```



# 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

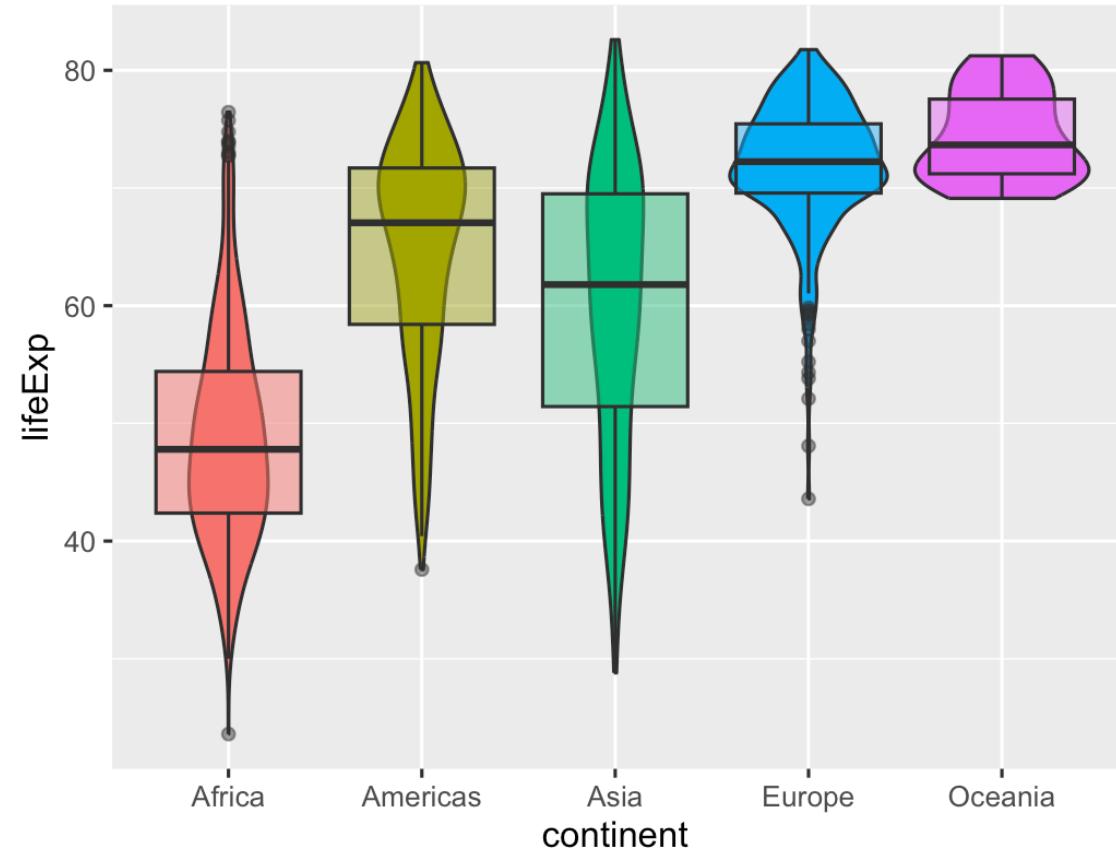
```
ggplot(data = gapminder_2007,  
       mapping = aes(x = lifeExp,  
                      fill = continent)) +  
  geom_histogram(binwidth = 5,  
                 color = "white") +  
  guides(fill = "none") + # Turn off legend  
  facet_wrap(vars(continent))
```



# 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(data = gapminder,  
       mapping = aes(x = continent,  
                      y = lifeExp,  
                      fill = continent)) +  
  geom_violin() +  
  geom_boxplot(alpha = 0.5) +  
  guides(fill = "none") # Turn off legend
```

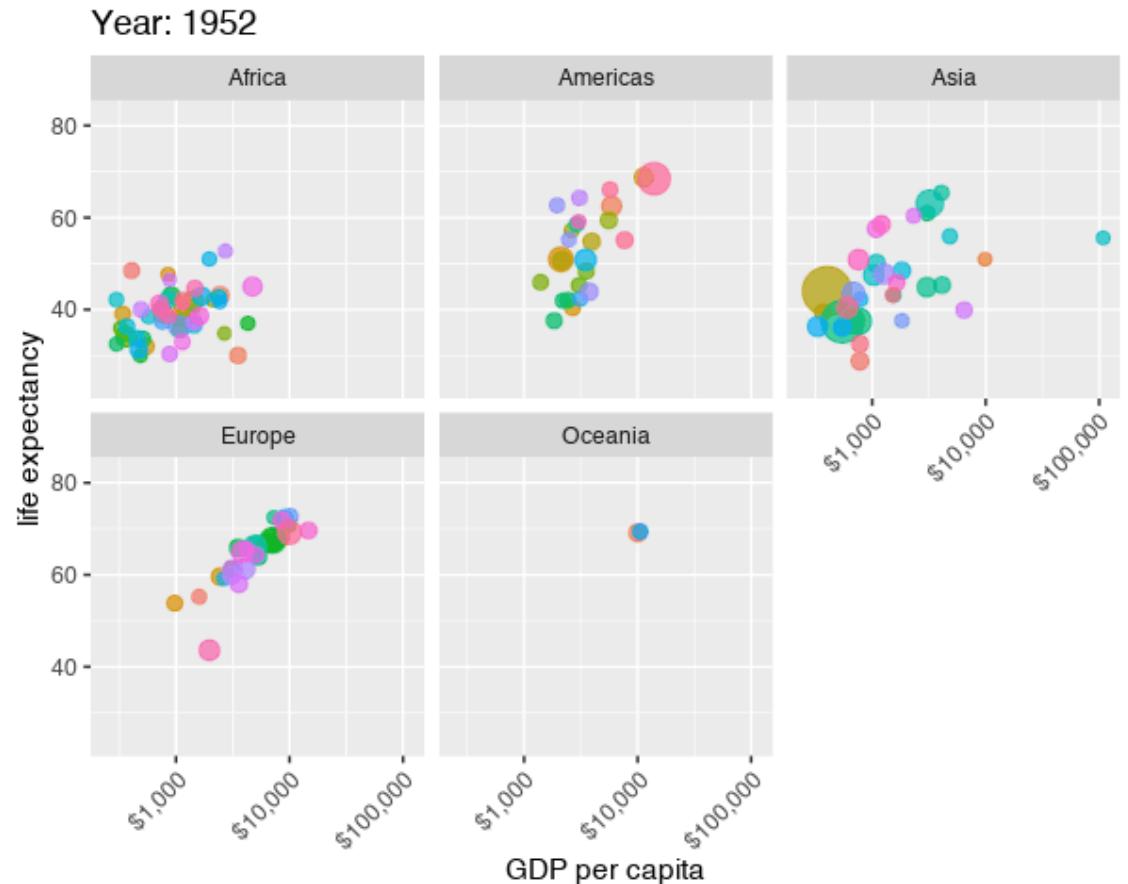


# Aesthetics in extra dimensions

# Time

Use `{gganimate}` to map variables to a time aesthetic

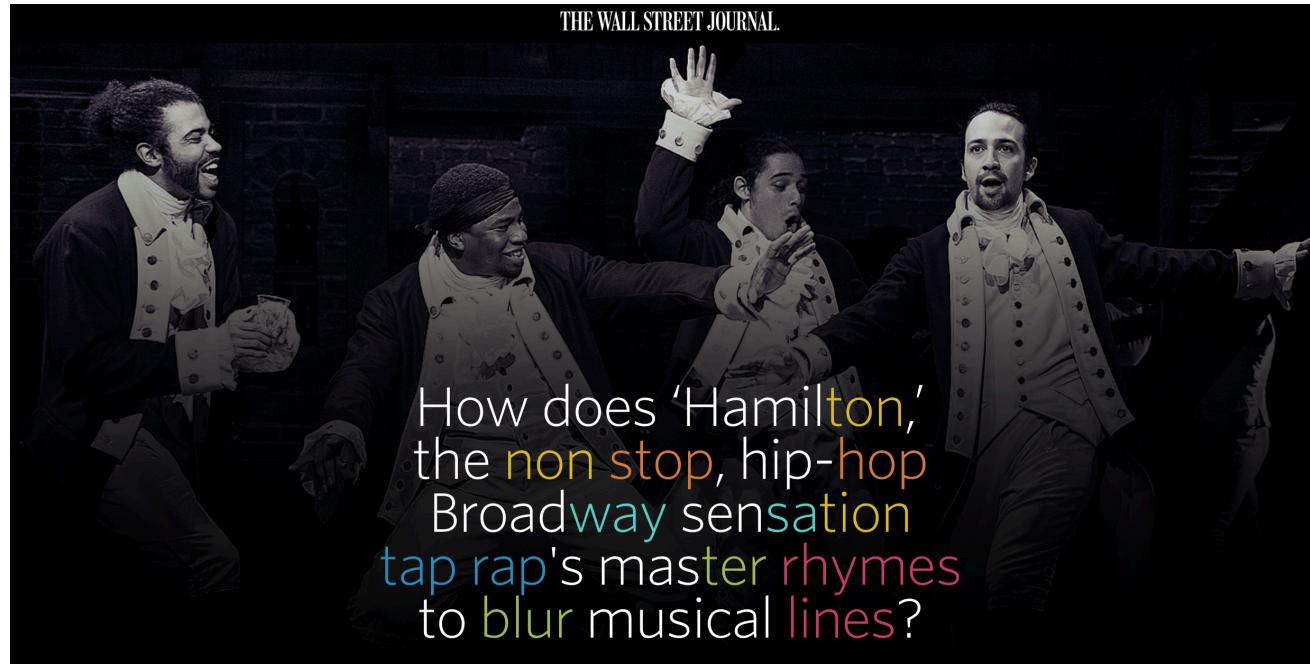
```
ggplot(gapminder, aes(x = gdpPercap, y = lifeExp, size = pop, color = continent)) +  
  geom_point(alpha = 0.7) +  
  scale_size(range = c(2, 12)) +  
  scale_x_log10(labels = scales::label_dollar) +  
  guides(size = "none", color = "none") +  
  facet_wrap(~continent) +  
  # Special gganimate stuff  
  labs(title = 'Year: {frame_time}', x = 'GDP per capita') +  
  transition_time(year) +  
  ease_aes('linear')
```

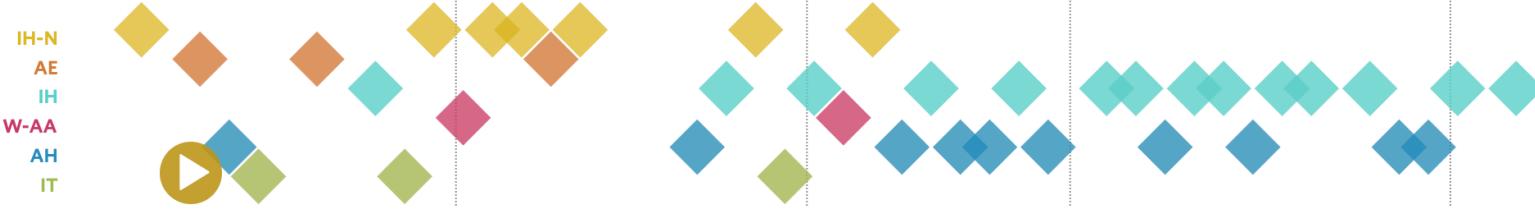


# Sound

Visualize internal rhyming schemes in music

<http://graphics.wsj.com/hamilton/>





Daveed Diggs

I'm in the cabinet I am complicit in  
 Watching him grabbing at power and kissing it  
 If Washington isn't gon' listen  
 To disciplined dissidents this is the difference  
 This kid is out

"Washington On Your Side" on "Hamilton (Original Broadway Cast Recording)"



Kendrick Lamar

Trapped inside your desire to fire bullets that stray  
 Track attire just tell you I'm tired and ran away  
 I should ask a choir "What do you require  
 to sing a song that acquire me to have faith?"

"good kid" on "good kid, m.A.A.d city"

# Animation, time, and sound

The Weather Channel Uses Animation to Show Dangers of Storm Surge



# Tidy data

# Data shapes

**For `ggplot()` to work,  
your data needs to be in a `tidy` format**

This doesn't mean that it's clean—  
it refers to the *structure* of the data

All the packages in the `{tidyverse}` work best with  
tidy data; that's why it's called that!

# Tidy data

Each variable has its own column

Each observation has its own row

Each value has its own cell

country	year	cases	population
Afghanistan	1990	745	19807071
Afghanistan	2000	2666	20595360
Brazil	1999	37737	172006362
Brazil	2000	80488	174504898
China	1999	212258	1272015272
China	2000	21766	128042583

variables

country	year	cases	population
Afghanistan	1990	745	19807071
Afghanistan	2000	2666	20595360
Brazil	1999	37737	172006362
Brazil	2000	80488	174504898
China	1999	212258	1272015272
China	2000	21766	128042583

observations

country	year	cases	population
Afghanistan	1990	745	19807071
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China	1999	212258	1272015272
China	2000	21766	128042583

values

From chapter 12 of *R for Data Science*

# Untidy data example

Real world data is often untidy, like this:

	A	B	C	D
1	Number of incidents			
2				
3	Office	2015	2016	2017
4	Utah County	134	145	167
5	Salt Lake County	302	334	331
6	Davis County	254	288	299
7	Juab County	78	82	87
8				
9	<b>bold = needs verification</b>			
10	yellow = compiled from different source			
11				

# Tidy data example

Here's the tidy version of that same data:

	A	B	C	D	E
1	Office	Year	Incidents	Needs Verification	Different Source
2	Utah County	2015	134	FALSE	FALSE
3	Salt Lake County	2015	302	TRUE	FALSE
4	Davis County	2015	254	FALSE	FALSE
5	Juab County	2015	78	FALSE	FALSE
6	Utah County	2016	145	FALSE	TRUE
7	Salt Lake County	2016	334	FALSE	FALSE
8	Davis County	2016	288	FALSE	FALSE
9	Juab County	2016	82	TRUE	TRUE
10	Utah County	2017	167	TRUE	FALSE
11	Salt Lake County	2017	331	FALSE	FALSE
12	Davis County	2017	299	FALSE	TRUE
13	Juab County	2017	87	FALSE	FALSE

This is plottable!

# Wide vs. long

Tidy data is also called "long" data

wide

id	x	y	z
1	a	c	e
2	b	d	f

long

id	key	val
1	x	a
2	x	b
1	y	c
2	y	d
1	z	e
2	z	f

# Moving from wide to long

Nowadays, `gather()` is called `pivot_longer()` and `spread()` is called `pivot_wider()`

wide

	x	y	z
id	x	y	z
1	a	c	e
2	b	d	f