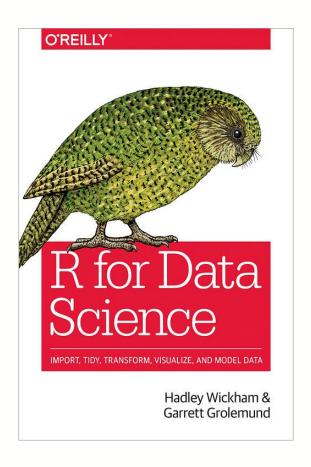
# Session 3: Introduction to ggplot2

# Acknowledgement

This session shadows Chapter 3 of the excellent:

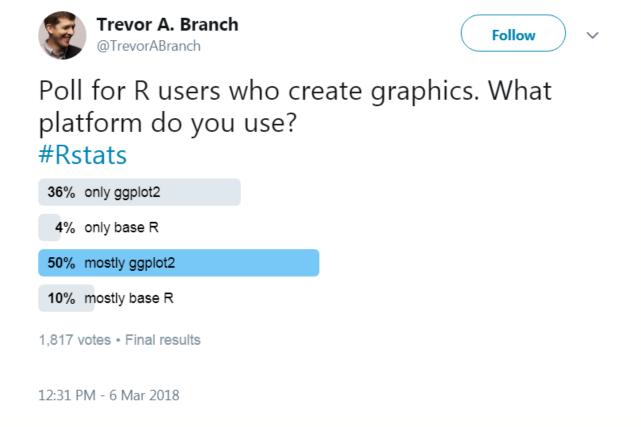


# ggplot2

Is one of several plotting systems in R

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# Why ggplot2?

1. Highly versatile

2. Relatively easy to make good-looking plots

 It meshes well with other tools we will be learning

# ggplot2

ggplot2 is part of the tidyverse, so:

library(tidyverse)

### mpg data

Data on car efficiency\*. 38 models produced in both 1999 and 2008. Please type:



\*Source: US Environment Protection Agency\* https://fueleconomy.gov/

### mpg data

test <- mpg

Now, whenever we type, **test**, it will refer to the mpg data.

test is a data frame.

#### What is a data frame?

A data frame is a rectangular collection of variables (in columns) and observations (in rows).

id	gender	score
1	F	10.24
2	F	5.98
3	M	7.62

### tibble = data frame

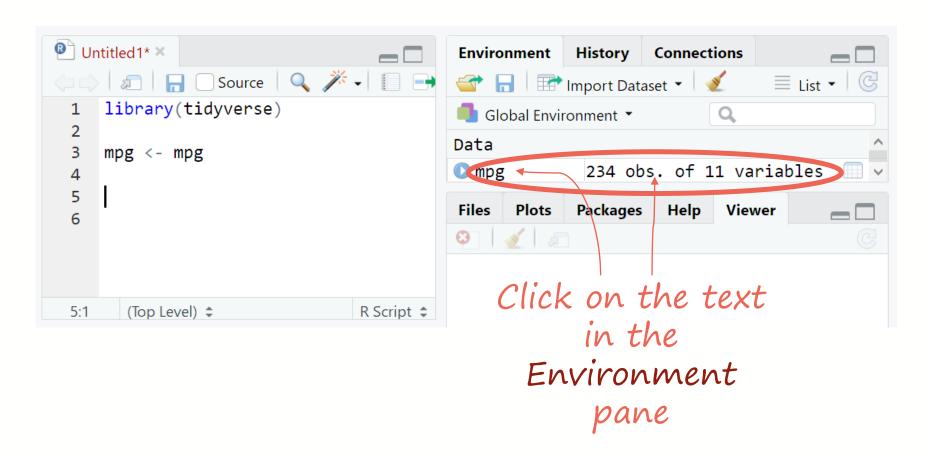
In the tidyverse you will see the term "tibble".

We'll take "tibble" to be synonymous with "data frame".

id	gender	score
1	F	10.24
2	F	5.98
3	M	7.62

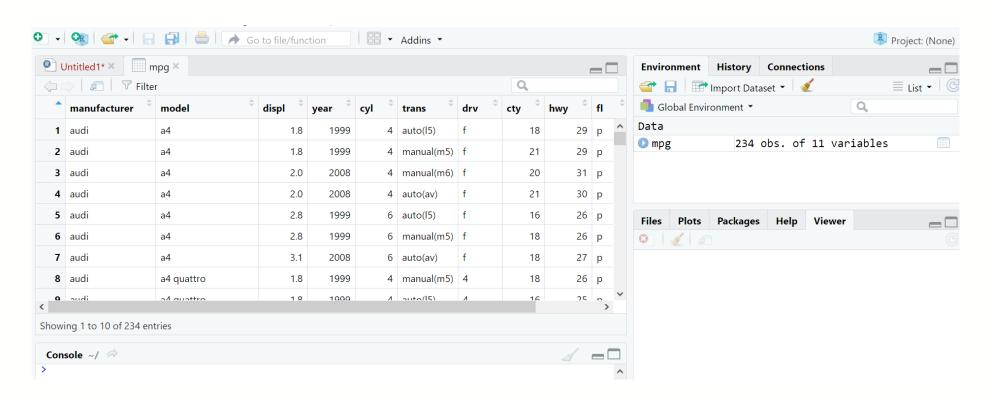
# Viewing the data

Several ways to examine a data frame. Option 1:



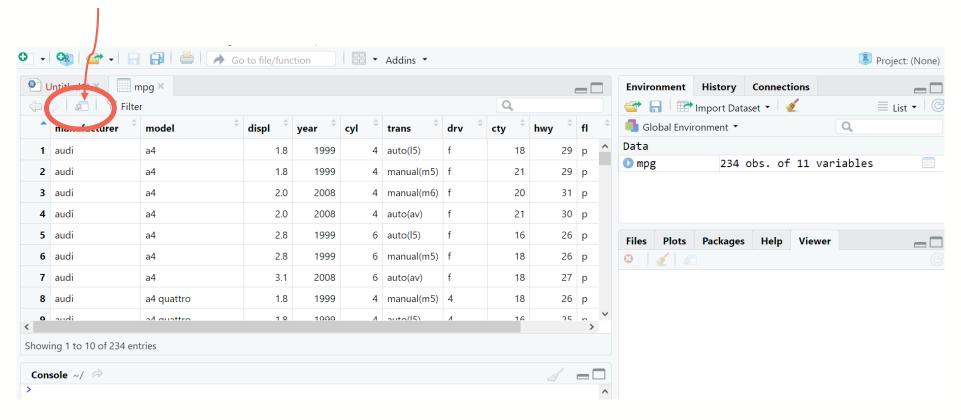
# Viewing the data

This brings up a view of the data in a new tab:



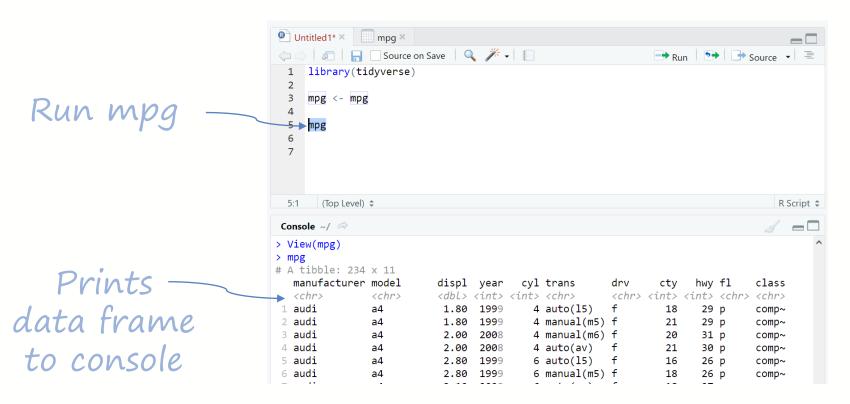
# Viewing the data

Click here to show the data frame in a new window\*



### Option 2: Preview in Console

Type the name of the dataset into editor/console, and run the line (Ctrl + Enter).



# Q. How many cars? What variables do we have?

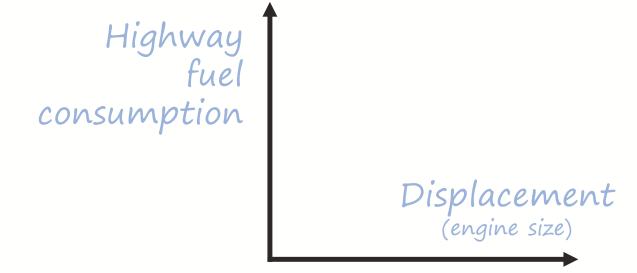
# Graphics with ggplot2

# The simple graph has bought more information to the data analyst's mind than any other device

- John Tukey

# Q. Do cars with large engines (displ) use more fuel than cars with small engines?

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# Q. Do cars with large engines (displ) use more fuel than cars with small engines?

```
Note that
R is case
ggplot(data = test) +
geom_point(aes(x = displ, y = hwy))
```

### Breakdown

```
1. We begin our plot with ggplot()
ggplot()
we name our dataset layer(s) with +

ggplot(data = test) +

geom_point(aes(x = displ, y = hwy))
```

# How do we move from data to graphic?

# Pen and paper exercise: Create a graphic from the data below.

year	time (sec)
1930	12.0
1960	11.3
1990	10.5

# Pen and paper exercise: Create a graphic from the data below.

Now note down all the subtle (unconscious?) choices you made when creating the graphic.

year	time (sec)
1930	12.0
1960	11.3
1990	10.5

1. What shape will represent the data?



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- 1. What shape will represent the data? (geom)
- 2. What visual (aesthetic) attributes do we give to the geom?



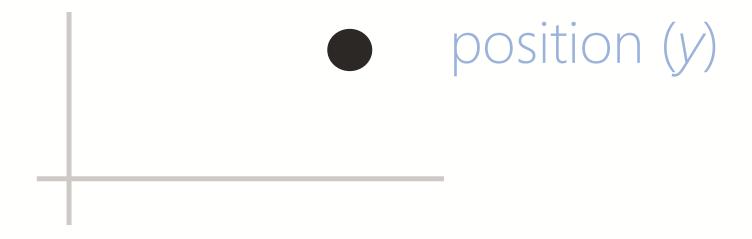
- 1. What shape will represent the data? (geom)
- 2. What visual (aesthetic) attributes do we give to the geom?



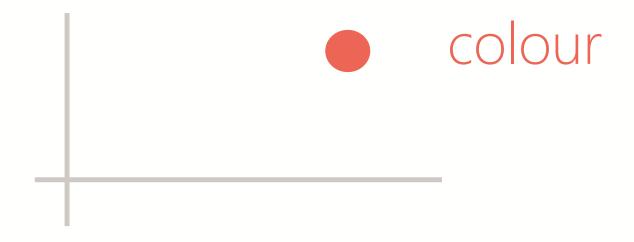
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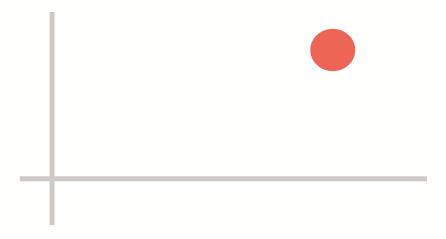
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## A statistical graphic

Maps data variables to geometric objects.

aesthetic attributes of



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Maps data variables to geometric objects.

aesthetic attributes of

```
ggplot(data = test) +
```

 $geom_point(aes(x = displ, y = hwy))$ 

Here, other aes() properties: size, colour, etc. are set by default

### Functions ()

ggplot(), geom\_point(), and aes() are functions.
Arguments (inputs) in a function are separated by commas

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Arguments (inputs) in a function are separated by commas
```

```
Here, we provide

geom_point() with one
argument : aes()

geom_point(test) +

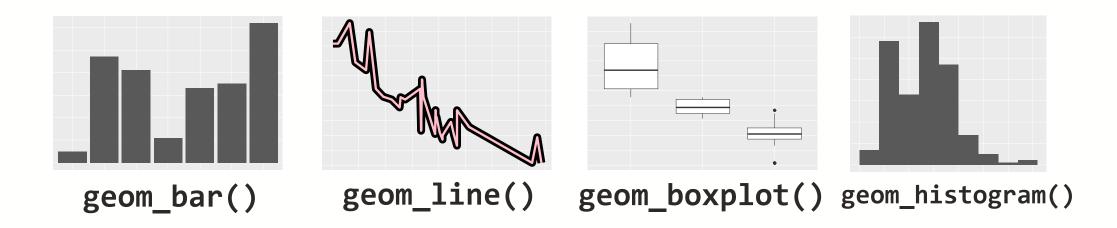
geom_point(aes(displ, hwy))
```

#### Shorthand

As ggplot knows the order of essential arguments, I will use this convention from now on:

### geoms

We tend to describe plots in terms of the geom used:



### Layering geoms

We can display more than one geom in a plot:

```
ggplot(test) +
  geom_point(aes(displ, hwy)) +
  geom_line(aes(displ, hwy))
```

Note: Nonsense graphic: used to illustrate principle only

# Layering geoms

We can display more than one geom in a plot:

Note: geom\_line used to illustrate principle only

# Layering geoms

To avoid duplication, we can pass the local **aes()** to **ggplot()**. This will make it a global value:

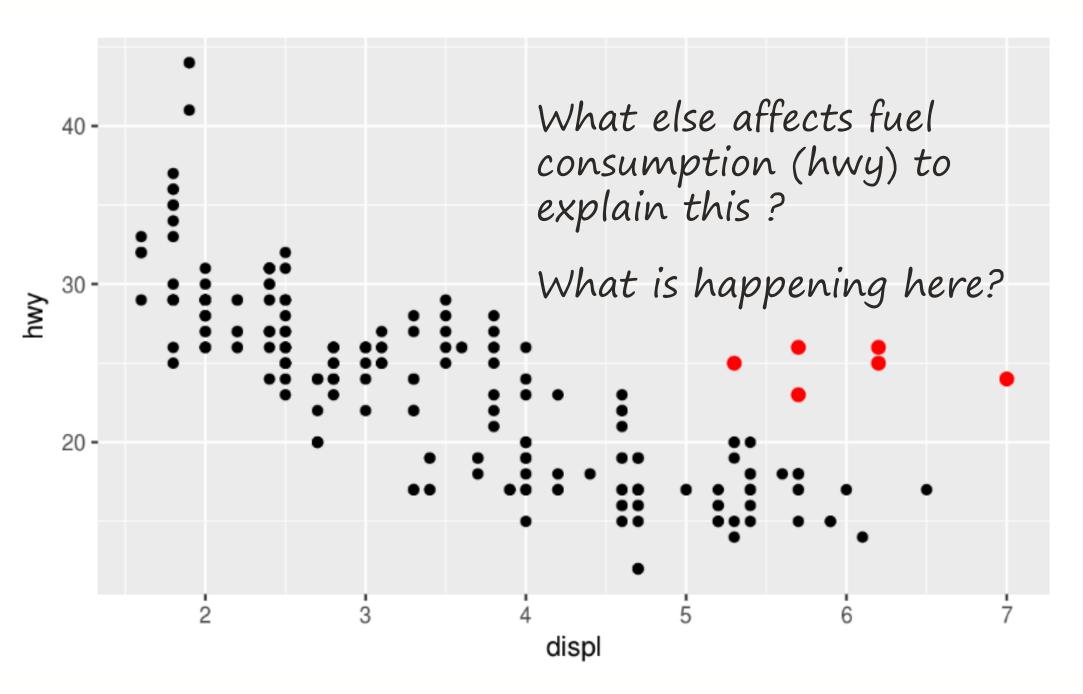
Note: geom\_line used to illustrate principle only

#### Your turn

A **geom\_smooth()** layer can help us identify patterns. Add geom\_smooth to our original plot:

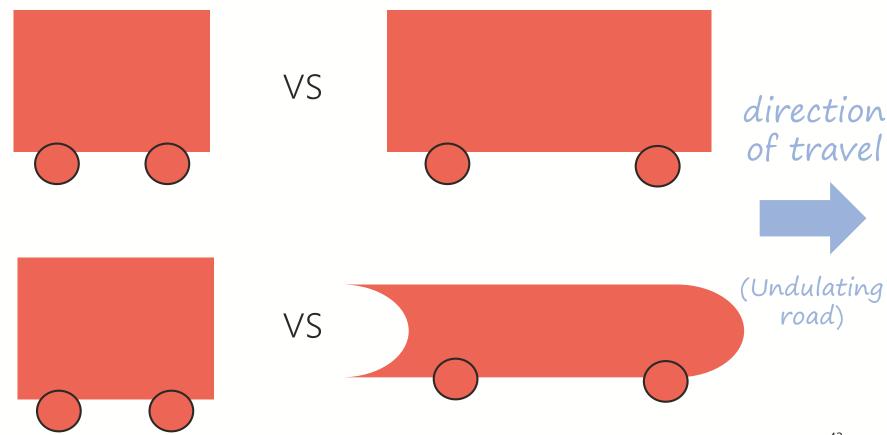
```
ggplot(data = test) +
  geom_point(aes(x = displ, y = hwy))
```

And (if you like) re-write in shorthand



### Play your cars right

Same engine, material, speed. Which is more fuel efficient?



### Hypothesis

Anomalous cars are lighter and/or aerodynamic.

Are they sports cars?

→ aesthetic attribute

We can map point colour to the class variable

- so a different colour for each class - to find out.

### Adding another variable

Remember: Arguments within the **aes**thetic wrapper describe how variables are mapped:

#### Outcome

The anomalous points are (mostly) two-seater cars.

Likely to be sports cars, therefore more aerodynamic and lighter.

#### All red

If you wished to apply the same colour to all points, the colour does not vary so argument goes outside aes():

```
ggplot(test) +
  geom_point(aes(displ, hwy), colour = "red")
```

### Small multiples

An alternative way to display additional variables is with small multiples. We do this with facet\_wrap()

```
ggplot(test) +
  geom_point(aes(displ, hwy)) +
  facet_wrap(vars(class))
```

### Small multiples

An alternative way to display additional variables is with small multiples. We do this with facet\_wrap()

# Demonstrating geoms: (note these are simple, unpolished graphics)

# Q. How are "cty" values distributed? Histogram

```
ggplot(test, aes(cty)) +
  geom_histogram()
```

# Q. How are "cty" values distributed? Histogram

```
ggplot(test, aes(cty)) +
  geom_histogram(binwidth = 4)
```

# Q. Distribution of engine size in each class? Box plot

```
ggplot(test, aes(class, displ)) +
  geom_boxplot()
```

# Q. Number of models by manufacturer? Bar plot

```
ggplot(test, aes(manufacturer)) +
  geom_bar()
```

# Q. Number of models by manufacturer? Bar plot - flipped

```
ggplot(test, aes(manufacturer)) +
  geom_bar()+
  coord_flip()
```

# Two variable bar plot (more common than geom\_bar)

```
ggplot(test, aes(manufacturer, hwy)) +
   geom_col()
```

### Reorder a two variable bar plot:

```
Name of variable by which to reorder x
ggplot(data, aes(reorder(x, a), y)) +
geom_col()
```

#### Plot labels

```
ggplot(mpg, aes(class, displ)) +
 geom violin()+
  labs(title = "Displacement by class",
       subtitle = "Any subtitle",
       y = "Displacement",
       caption = "Source: US EPA")
```

or: pdf, jpg as you wish

### ggsave("plot\_name.png")

#### By default:

- saves most recent ggplot to your working directory
- saves a plot in the same dimensions as plot window

Tip for now: adjust dimensions of plot pane in RStudio as you wish, then save.

### Save your script!

Think of your script as the "real" part of your analysis.

File → Save As... → ggplot\_intro.R

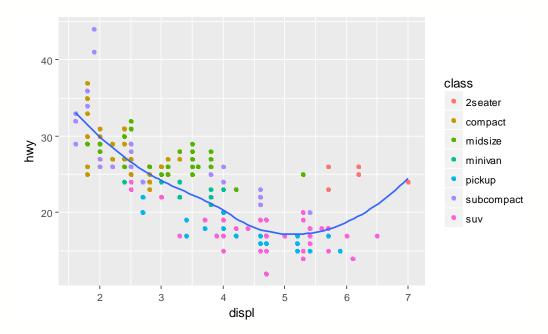
### accidental aRt

https://twitter.com/accidental aRt

# Addendum: A review of local and global aesthetics

### Two layers:

```
ggplot(mpg) +
  geom_point(aes(displ, hwy, colour = class)) +
  geom_smooth(aes(displ, hwy))
```



### Duplicate aes attributes:

### Global and local

```
ggplot(mpg, aes(displ, hwy)) +

geom_point(aes(colour = class)) + Note that
colour = class
geom_smooth()

must remain within
aes() of geom_point, or
it will be applied to
geom_smooth
```

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# End