# **ODDS** AND ENDS

#### Overview

Additional ggplot features and visual hypothesis test

Visual hypothesis testing

Writing functions

If there is time: Q-Q plots

#### Additional features of ggplot



#### Review/continuation of ggplot

A Frame: Coordinate system on which data is placed

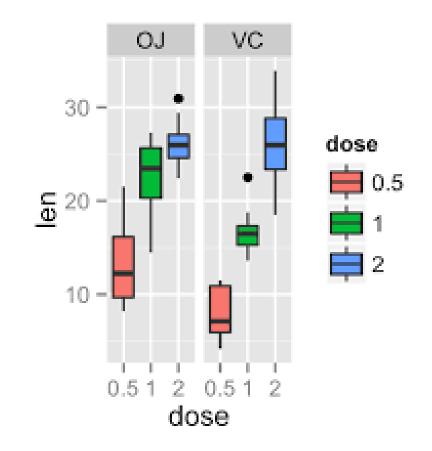
**Glyphs**: basic graphic unit representing cases or statistics

**Scales and guides**: shows how to interpret axes and other properties of the glyphs

**Facets**: allows for multiple side-by-side graphs based on a categorical variable

**Layers:** allows for more than one types of data to be mapped onto the same figure

**Theme**: contains finer points of display (e.g., font size, background color, etc.)



#### Review/continuation of ggplot

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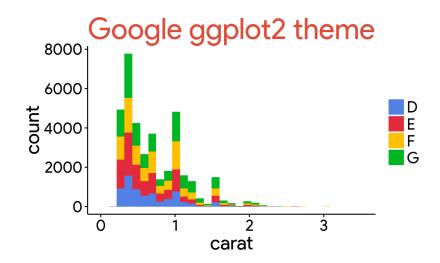
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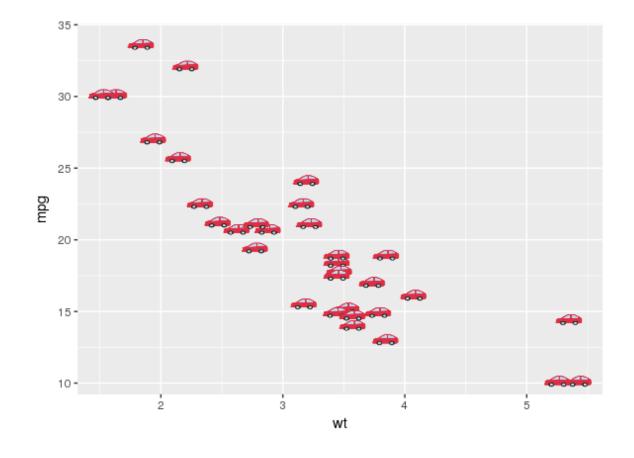
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#### ggplot bonus features: emojis

There are also additional packages that add more geoms

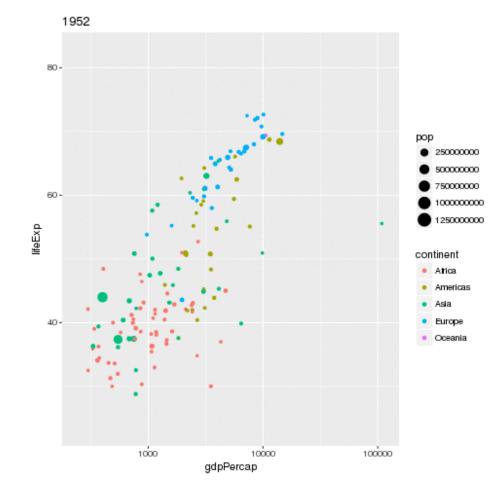
- > library(emoGG)
- > ggplot(mtcars, aes(wt, mpg)) + geom\_emoji(emoji="1f697")



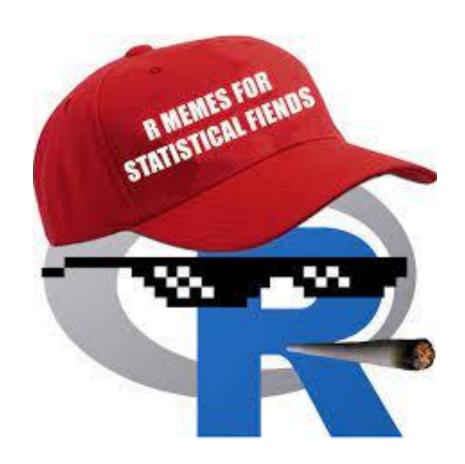
#### ggplot bonus features: animation

We can create animated images (gifs) using the gganimate package

```
library(gganimate)
ggplot(gapminder, aes(gdpPercap, lifeExp,
       size = pop, col = continent)) +
 geom_point(alpha = 0.7, show.legend = FALSE) +
 scale_x_log10() +
 # Here comes the gganimate specific parts
  labs(title = 'Year: {frame_time}',
        x = 'GDP per capita', y = 'life expectancy') +
  transition_time(year) +
  ease_aes('linear')
```



### Let's try it in R...

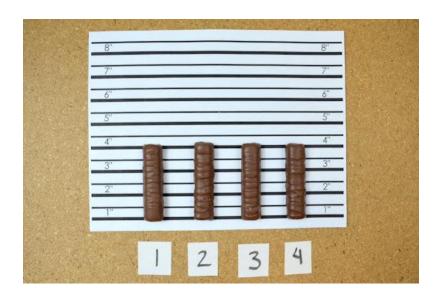


#### Visual hypothesis test

In visual hypothesis tests, we create data visualizations to try to assess whether particular relationships exist in our data.

One way this is done through a visual lineup.

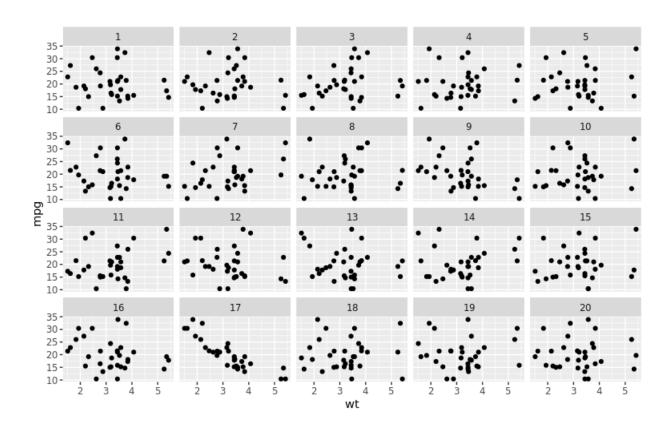




Which candy bar is 5<sup>th</sup> Avenue bar?

#### Visual hypothesis test

Which plot shows the true relationship between a car's weight and the number of miles per gallon a car gets?



Let's try it in R...



#### Writing functions

We've used many R functions in this class

Let's explore writing our own functions!



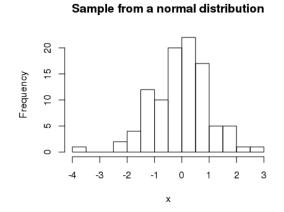
#### Quantile-Quantile plots

#### Density functions

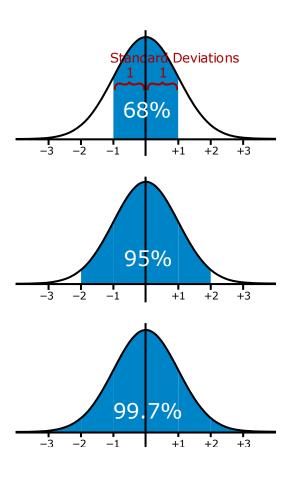
## $f(x,\mu,\sigma) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{(x-\mu)^2}{2\sigma^2}}$

A **density curves** are mathematical functions f(x) that are used to calculate probabilities





dnorm(x, 0, 1)
rand\_data <- rnorm(100, 0, 1)
hist(rand\_data)</pre>



How can you assess whether data comes from a particular distribution?

#### Quantiles

As you know, to get the probability (area) from a normal distribution we can use the pnorm function

```
pnorm(x, mu, sigma)
```

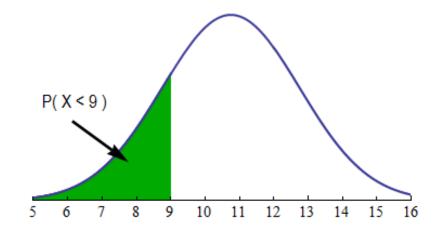
e.g., 
$$P(X < 9; \mu = 11, \sigma = 3)$$
  
pnorm(9, 11, 3) = 0.2525

The quantile function is the inverse of the probability functions.

For a given probably p, (area between 0 and 1), it tells us the x value such that P(X < x) = p.

```
qnorm(p, mu, sigma)
```

e.g., 
$$P(X < ?; \mu = 11, \sigma = 3) = 0.252$$
  
qnorm(.2525, 11, 3) = 9



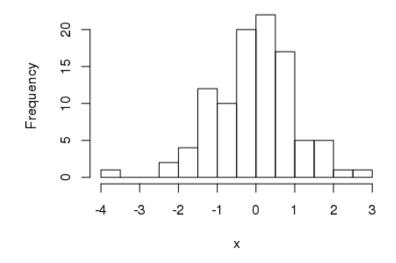


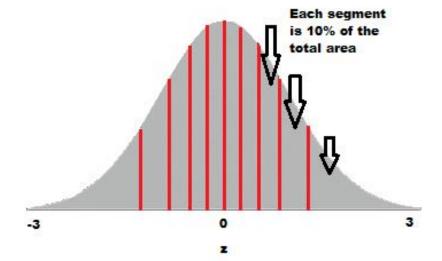
#### Quantile-quantile plots (Q-Q plots)

Quantile-quantile plots (Q-Q plots) can be used to assess whether a data sample comes from a particular distribution

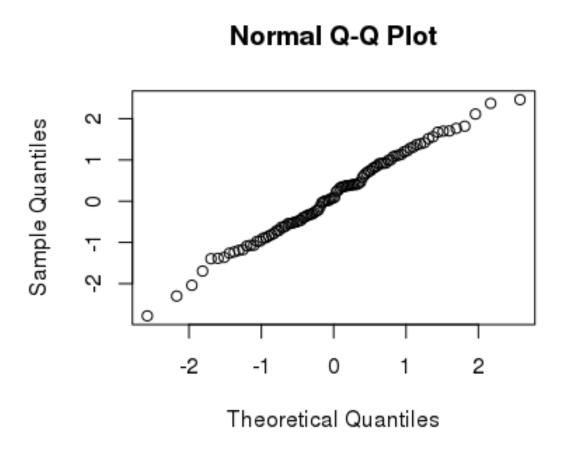
Q-Q plots show the observed quantile values from a data sample against the theoretical quantile values from a known distribution

#### Sample from a normal distribution

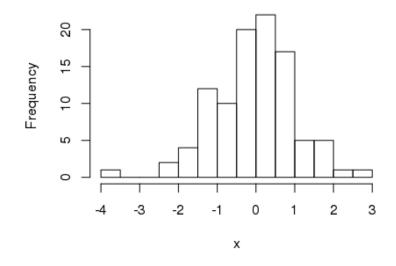


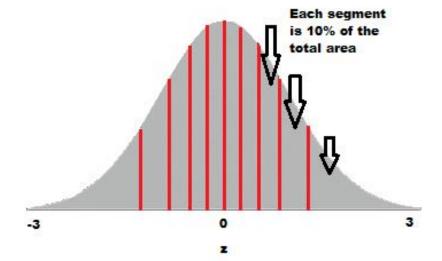


#### Quantile-quantile plots (Q-Q plots)



#### Sample from a normal distribution





## Let's try it in R...

