Quantitative Reasoning

# Working with Numerical data

## KEY CONCEPTS

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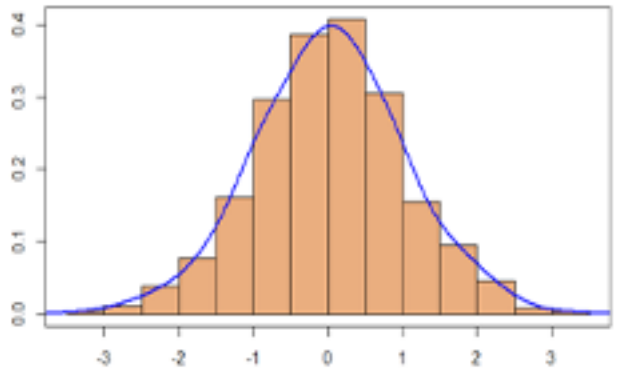
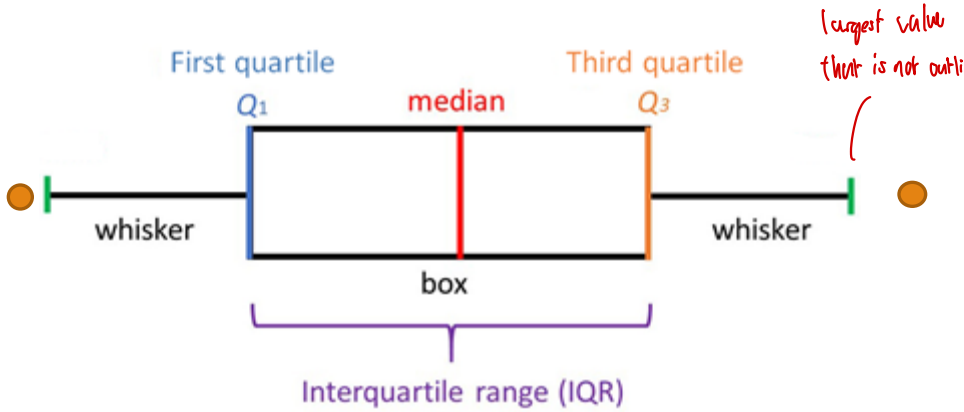
# Outline

Distributions (single variable)

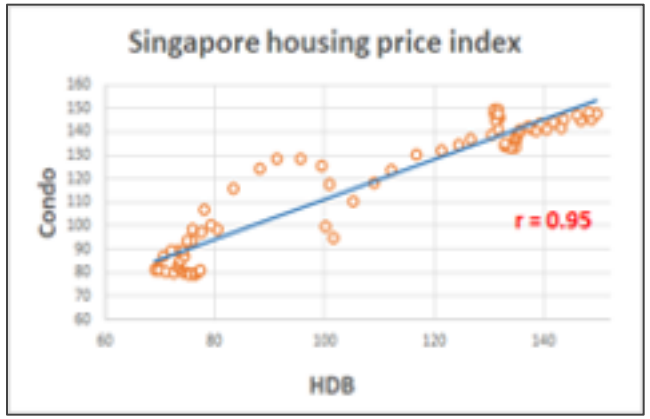
Numerical data

Association (2 variables)

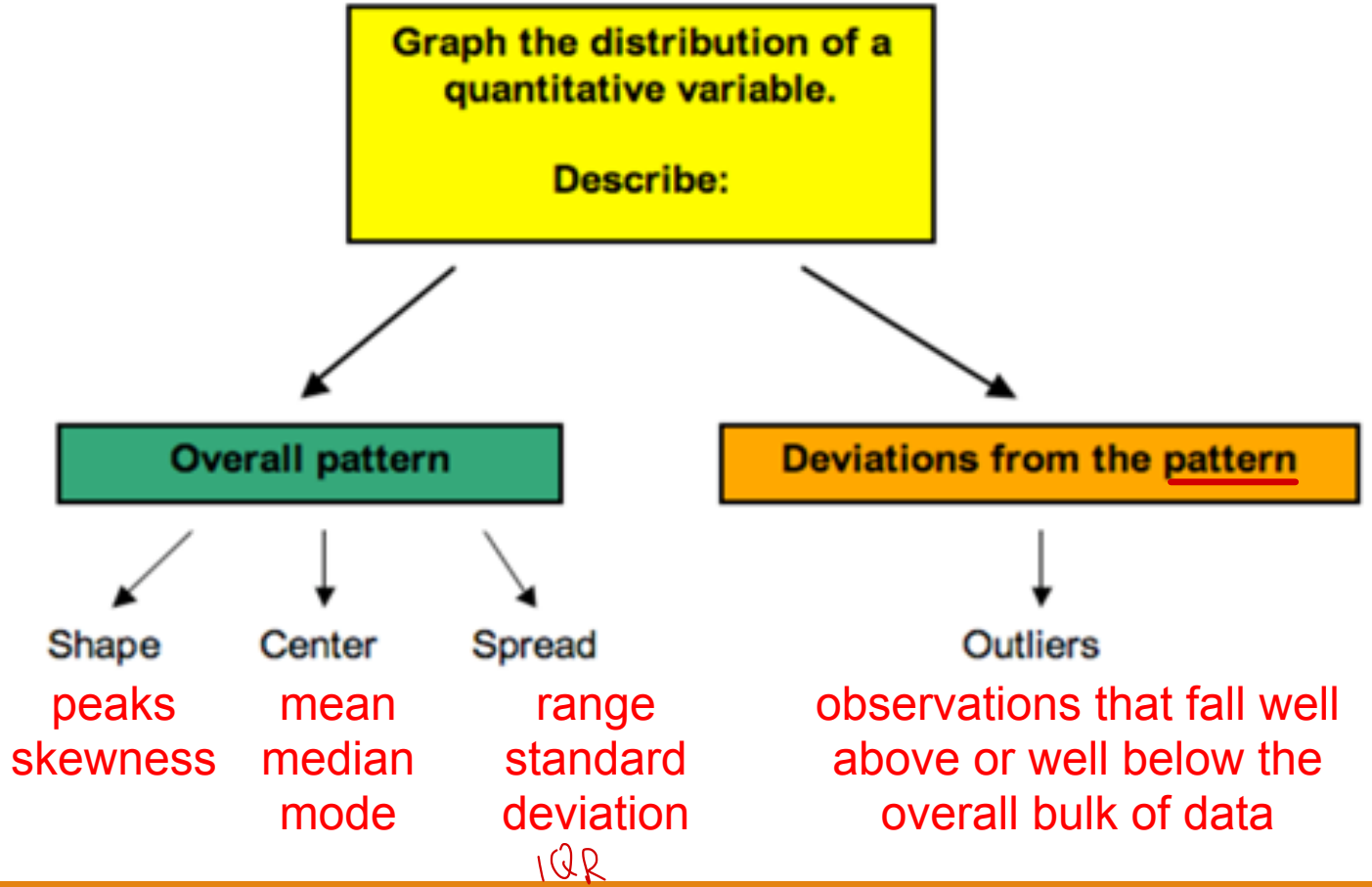


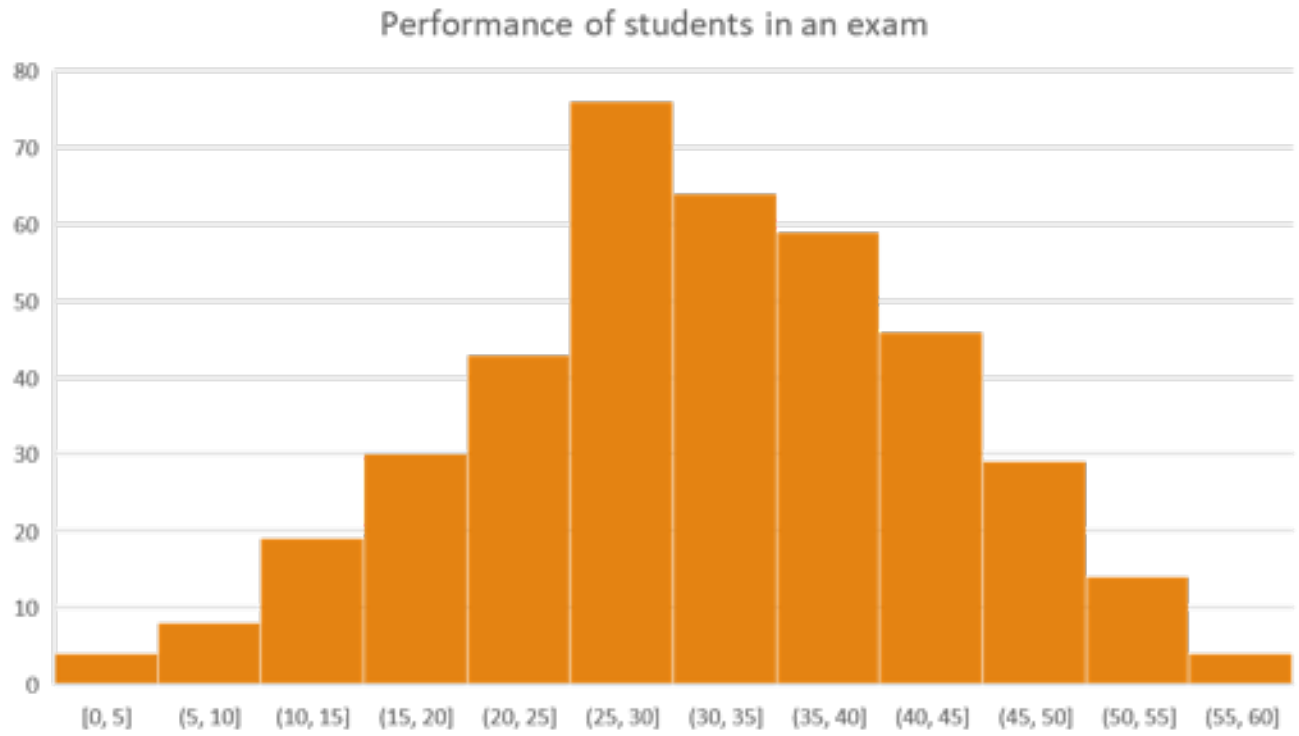

Outlier is defined as any point that lies above Q3 + 1.5 × IQR, or below Q1 - 1.5 × IQR,



# How to describe distribution?



# Getting information out of histograms

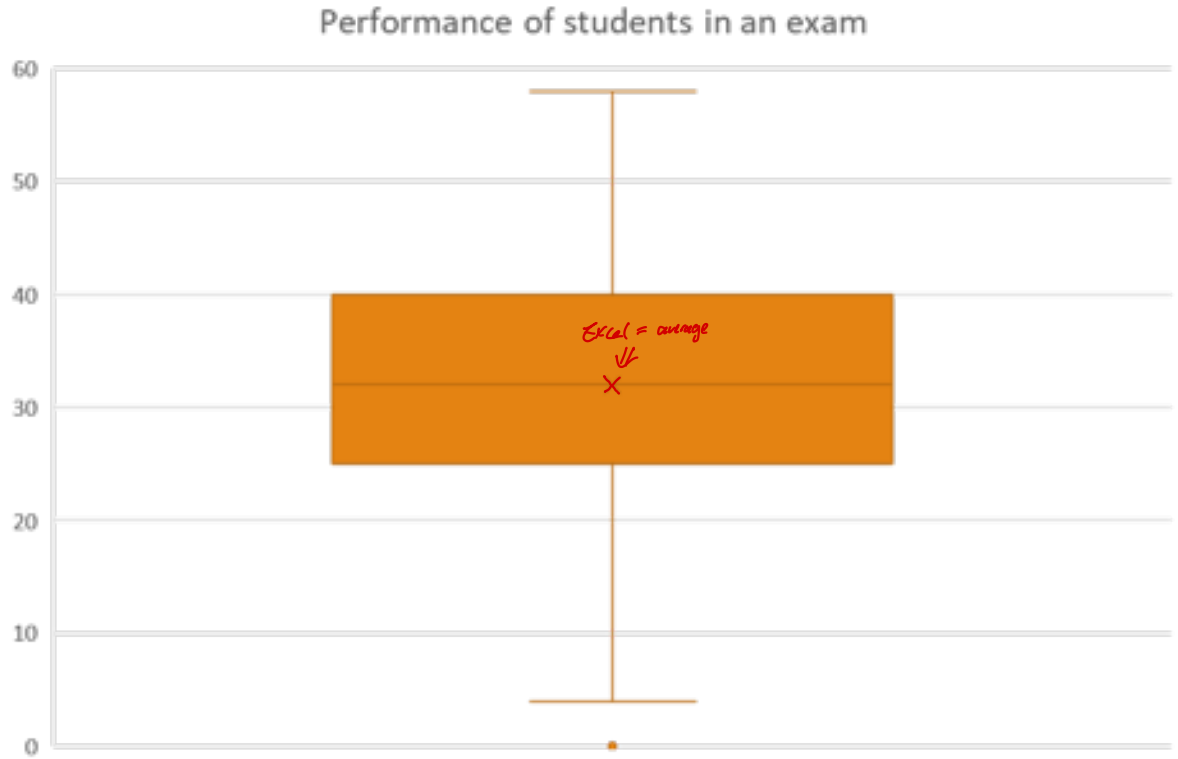



Histogram showing the distribution of scores for an exam taken by students in a school. The maximum mark for the exam is 60.

- What can you say about the performance of the students in the school based on the histogram?
  - Failure rate ✓
  - Borderline failures ✓
  - Distinction rate (those who scored 75% and above) ✓

- Is it possible to determine the following information based on the histogram?
  - Mean ✗ → score of every single student
  - Median ✗
  - Standard deviation ✗
  - Q1 and Q3 ✗
  - Outliers ✗

# What box-plots can tell us



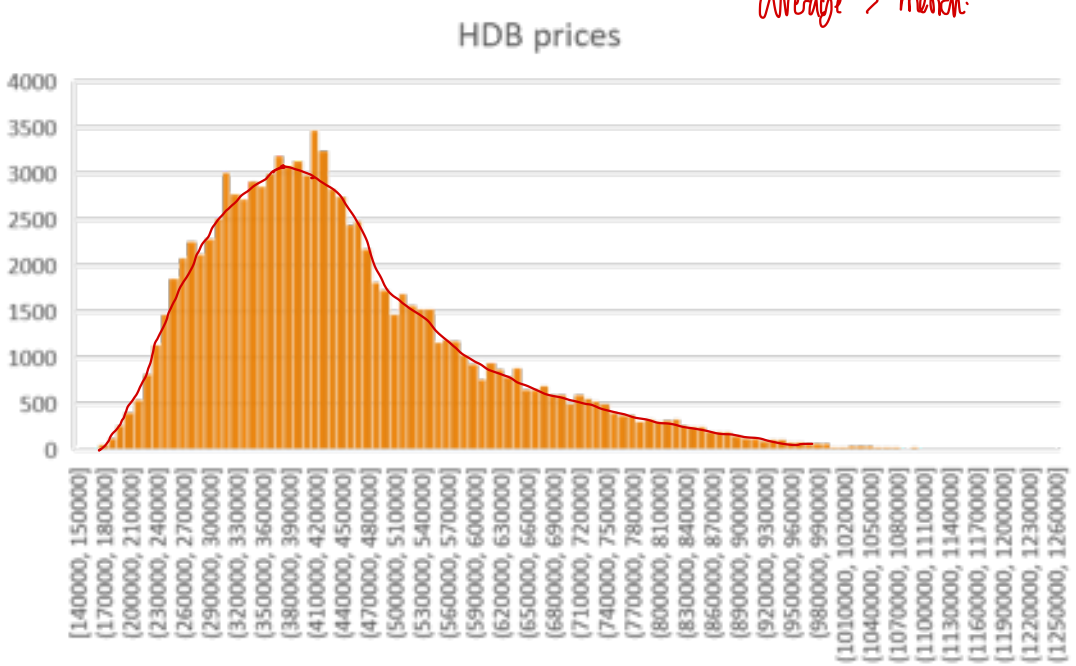

The same data shown as a histogram in the previous slide is now plotted using a boxplot.

- What can you say about the performance of the students in the school based on the boxplot?
  - Failure rate
  - Borderline failures
  - Distinction rate (those who scored 75% and above)

- Is it possible to determine the following information based on the boxplot?
  - Mean
  - Median
  - Standard deviation
  - Q1 and Q3
  - Outliers

# Outliers and skewness

average > median.

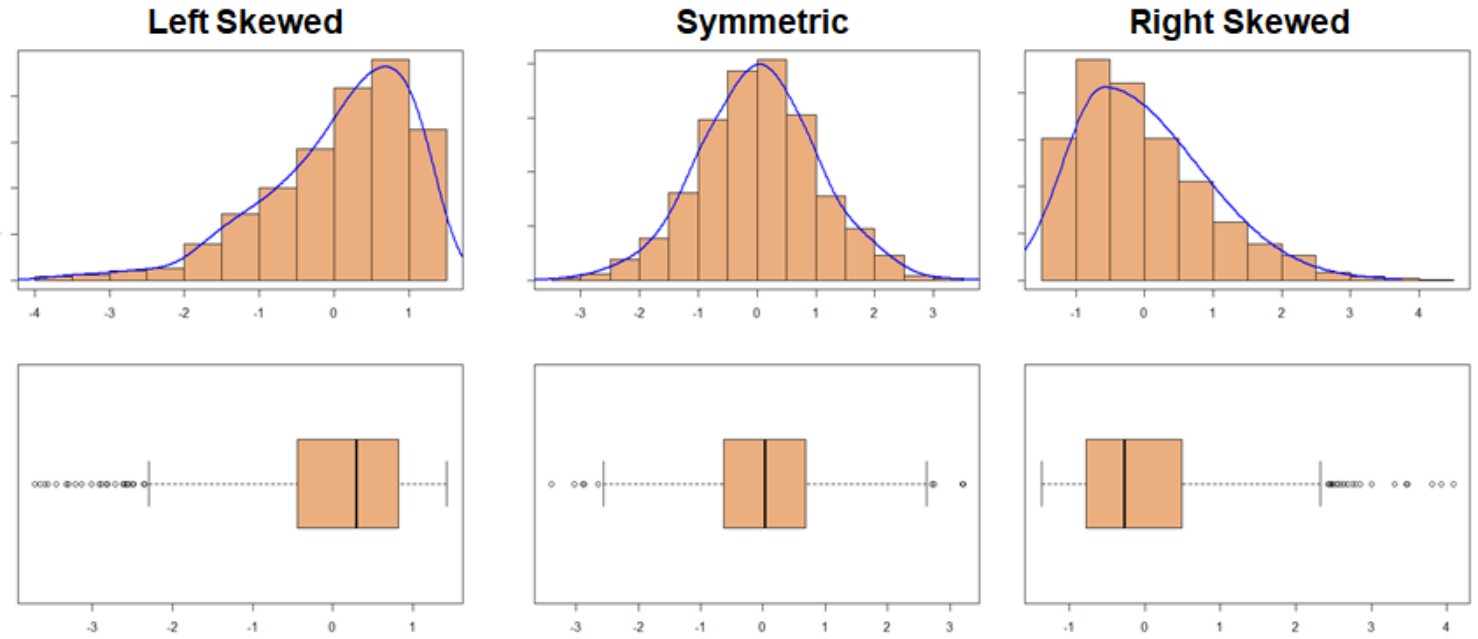




- Suppose the mean HDB price is $496 870 whilst the median HDB price is $468 000. Why do the mean and median differ significantly?
- Why do you think they prefer to talk about median HDB price as compared to mean HDB price?

↳ outliers affect average significantly, but don't really affect median.

# Outliers, skewness and robust statistics




- How can outliers affect summary statistics such as
  - Mean
  - Median
  - Standard deviation
  - IQR

- How do means and medians compare in the above distributions?

mean < median < mode (in general)

mode ≈ median ≈ mean.

mode < median < mean. (in general)

# Histogram or Boxplot

- Histogram typically gives a better sense of the shape of the distribution of a variable compared to a boxplot.
- If we wish to compare the distributions of different data sets, putting the different boxplots side by side is more illustrative.
- Boxplot is better if we want to identify outliers

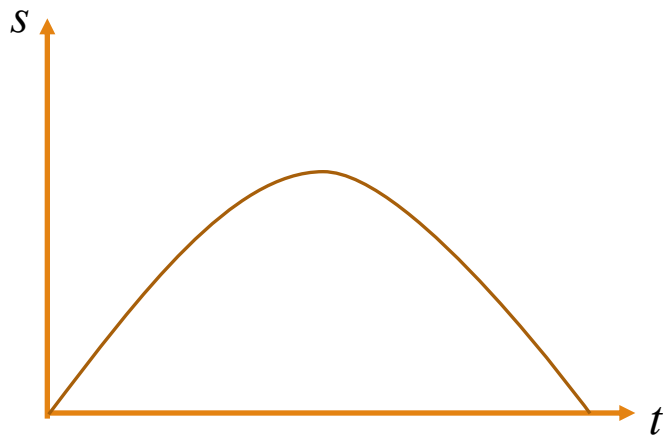
# Deterministic relationships (2 variables)

- A formula, for which given the value of one of the variables, you can calculate a **true** value for the other variable. E.g degrees Celsius to Fahrenheit.

$$T^0F = \frac{9}{5}(T^0C) + 32$$

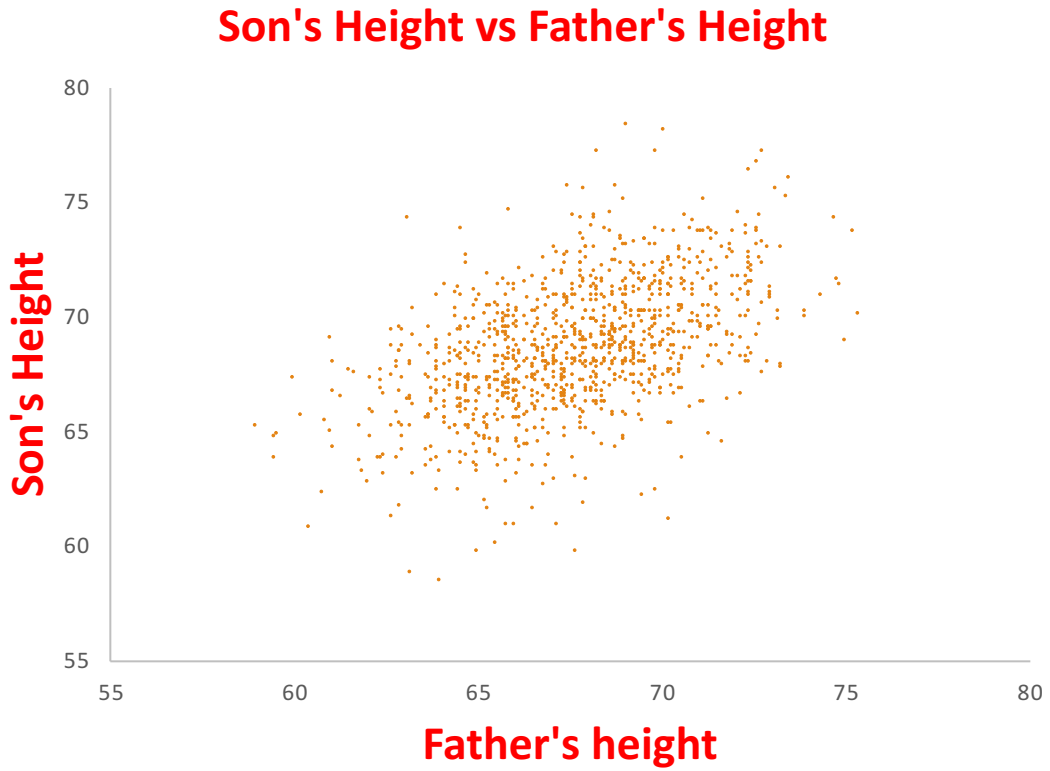
- Formula for calculating height of a stone thrown in the air (assuming ideal conditions)

$$s = ut + \frac{1}{2}at^2$$

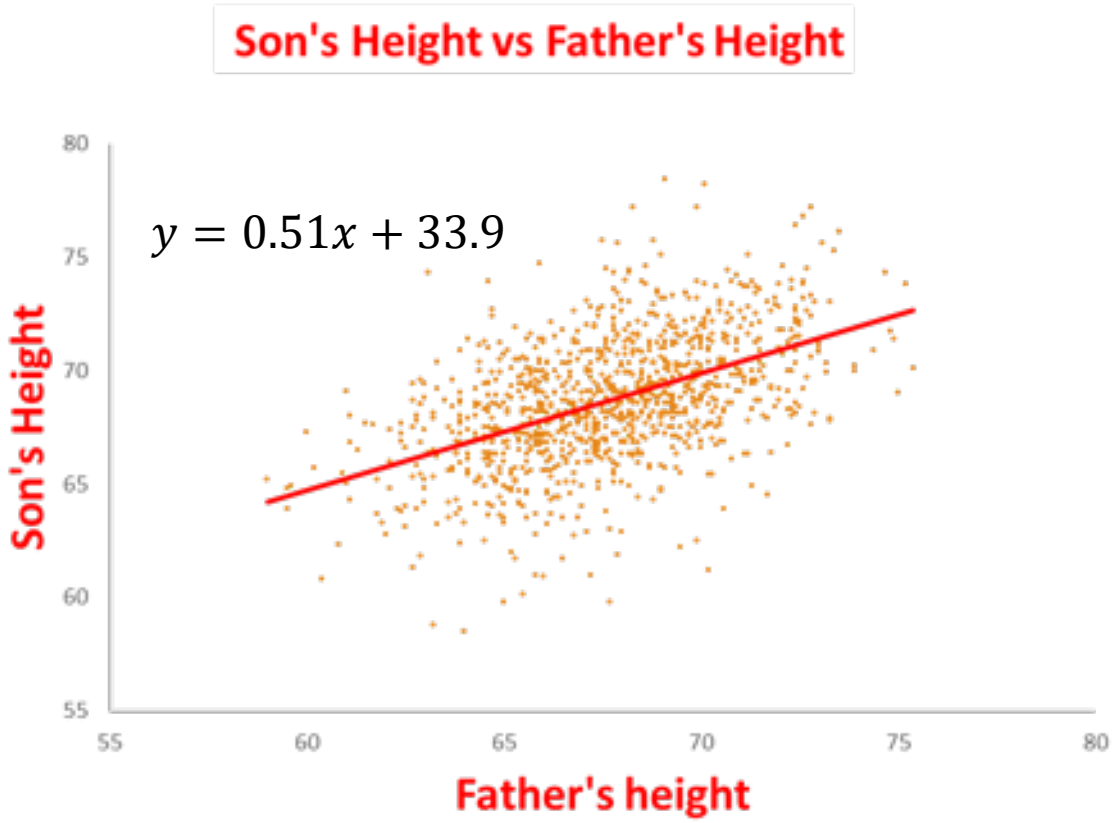


# Non-Deterministic relationships (2 variables)

- The relationship between numerical variables **cannot** be codified into a formula which gives us **true** values.
- We can only collect data and try to model the relationship and **we use the scatterplot to describe if there's any association between the 2 variables**.
- The scatter plot/model can comes with a "formula", but we need to be careful as to what we can and cannot do with that formula.



# Simple linear regression




- Suppose we wish to investigate the relationship between a father and his son's height. Let $x$ denote the father's height (independent variable) and $y$ denote the son's height (dependent variable).
- Based on the scatter plot, is there an association between the 2 variables?
- What is the predicted son's height for a father whose height is 67 inches? What about a father whose height is 80 inches?
- What are the limitations of our "formula" as compared to a deterministic relationship?

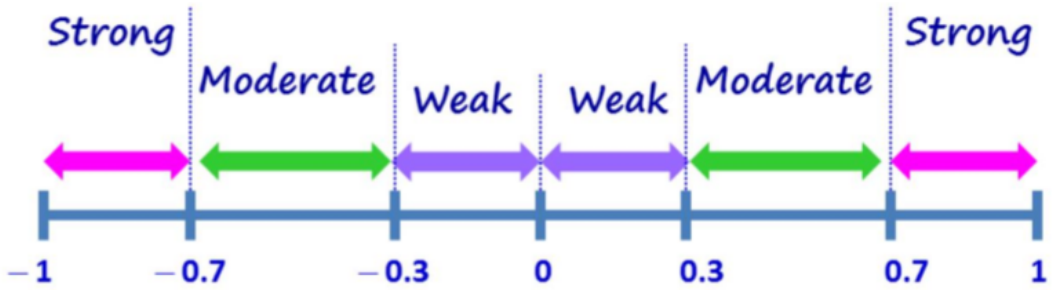
# Correlation coefficient *r*.

The correlation coefficient is a way to quantify the degree of **linear relationship** (which is one type of association) between numerical variables.

- measures ***linear*** association between 2 variables (NOT causation!)
- ranges between **-1 and 1** (no units)
- $r > 0 \rightarrow$ **positive** *linear* association

  $r < 0 \rightarrow$ **negative** *linear* association

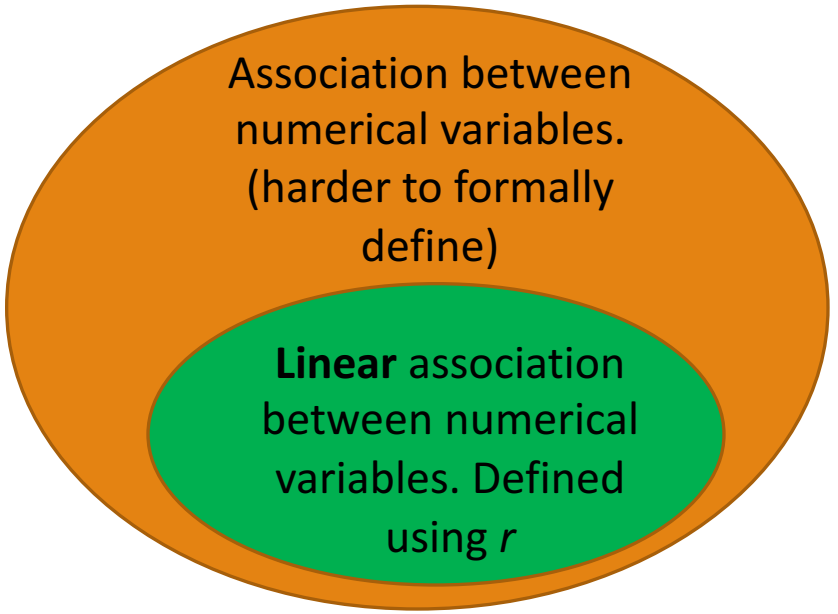
  $r = 0 \rightarrow$ **no** *linear* association



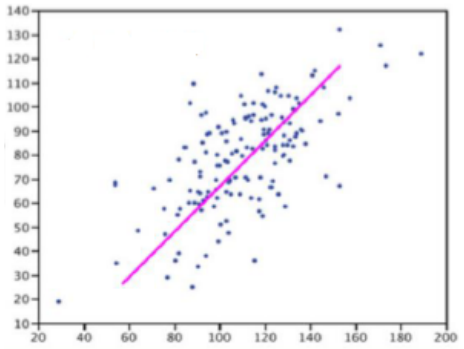
# Correlation coefficient and gradient

- $r$ is not affected by the following
  - Adding and subtracting constants to either variable
  - Multiplying and dividing *positive* constants to either variable.
  - Interchanging the *x* and *y* axis.
- $r \neq m$ in general.
  - $r = m \times \frac{s_x}{s_y}$

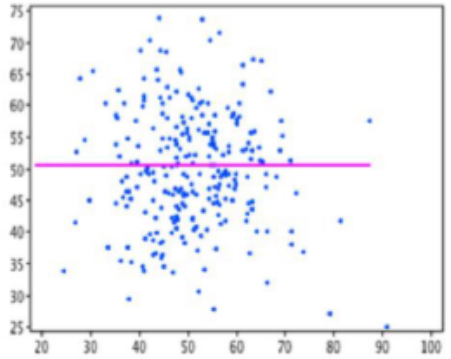
Using this you can also figure out when is *r* going to be the same as *m*



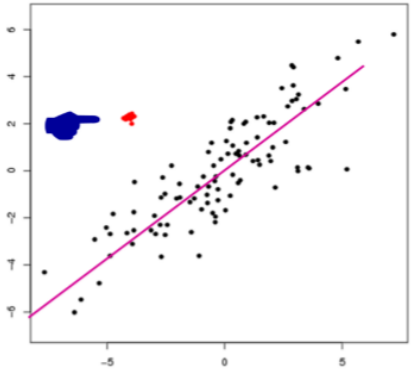
# *r* value alone doesn't tell you the whole story



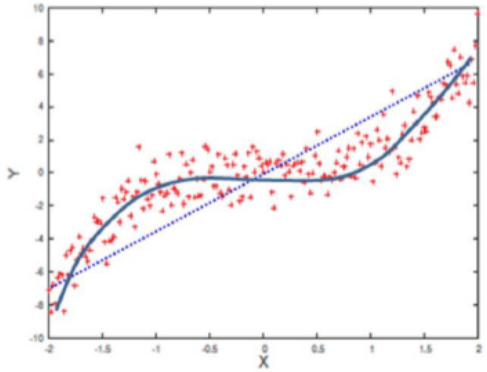
$r = 0.85$ → might not be a straight line.

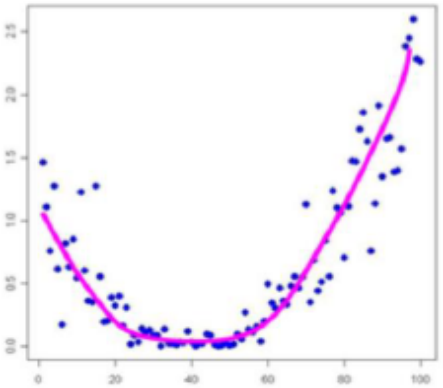


$r = 0$ → no linear association.



**Check scatter plot!**

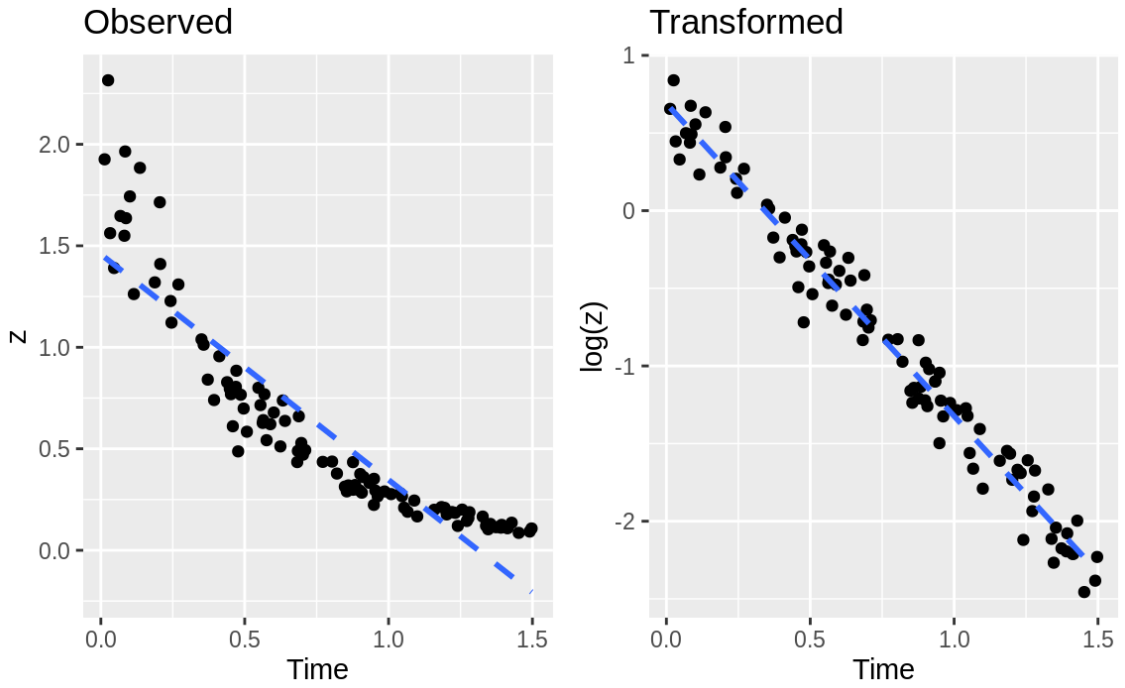




**Note:** When an outlier is removed, the $r$ value can increase, decrease or remain unchanged!

# Non-linear regression

- The goal is to use our understanding of **linear regression** to help us understand non-linear ones.




**Exponential decay of a population of some organism:**

1. Model using

$$y = cb^t$$

2. Take log on both sides

$$\ln y = \ln c + t \ln b$$

Convert to linear form

$$Y = mX + C$$

$$Y = \ln y,\ m = \ln b,\ X = t \text{ and } C = \ln c$$