Introduction to Statistics

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IMPERIAL

Overview

Session 1

- Describe your data
  - Descriptive statistics - summarising the data
  - Visualisation (plots and figures)
- Inferential analysis:
  - Inference about population from sample
- Given two groups of data
  - Test the differences between groups (**hypothesis testing**)
  - Test the relationship between two variables (correlation)

Session 2

- Lab Practice using SPSS

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Introduction to Statistics

Part 1: Describe your Data

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Describe your data

Describe your data

- Types of Variables
- Descriptive Statistics:
  - Average: Mean, Mode, Median, Frequency distribution
  - Spread/variability: Range, Percentile, Standard deviation
  - Skewness, Outliers
- Visualization: Plots

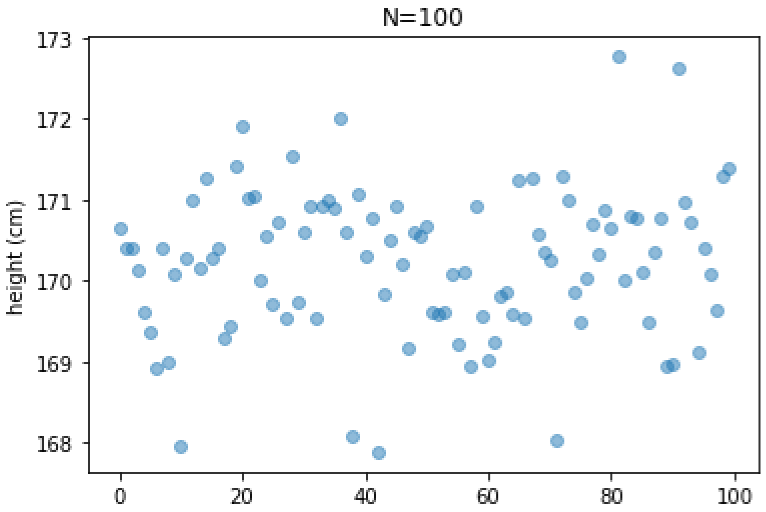
How would you describe it?

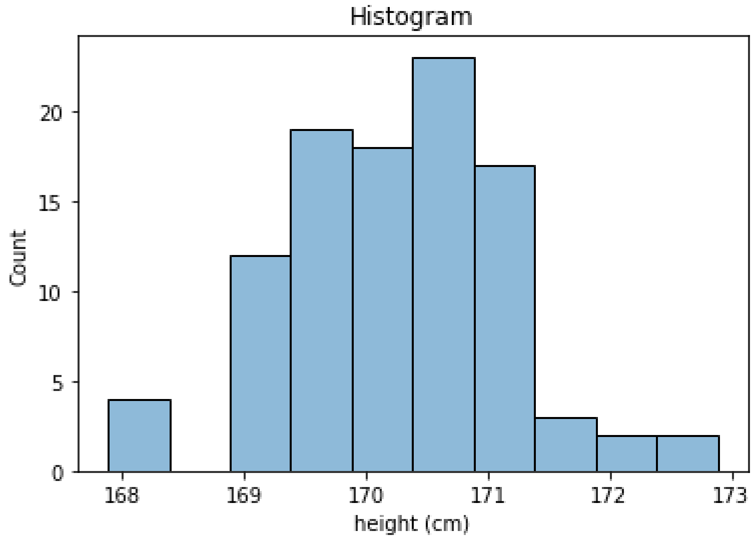
N=9

Height (cm)
167
168
160
170
171
160
162
165
167

N=100

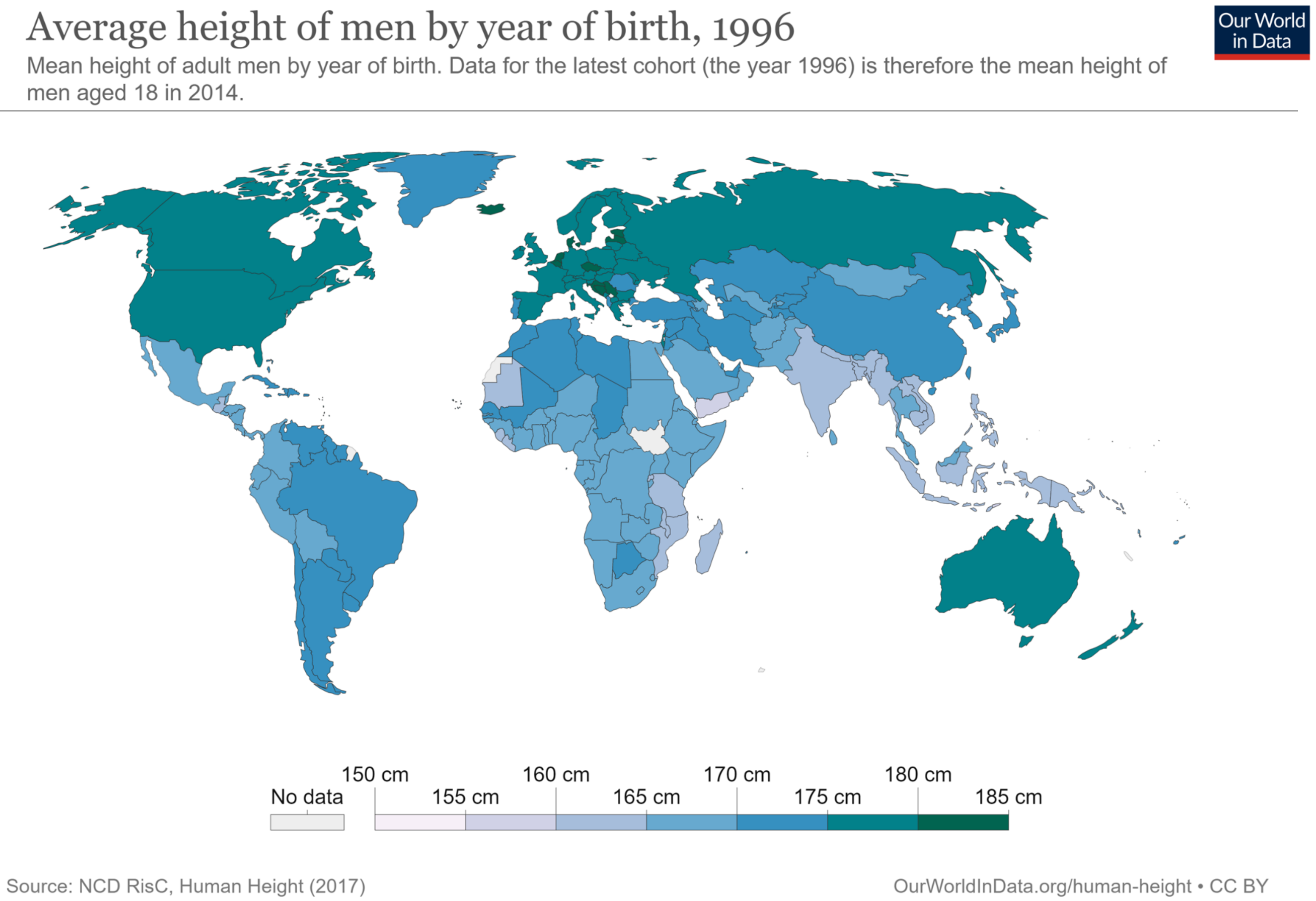
Height (cm)
167
168
160
170
171
.
.
.
.
.
.
.





Average – centre tendency of data
Variability – spread of data
Nice plots!

Example:



Source: https://ourworldindata.org/grapher/average-height-of-men

Types of Variable

Numerical

- Quantitative: blood pressure, sugar level, no of cells, height, BMI

  Continues, Discrete

Categorical

- Qualitative: ethnicity, disease or not? , sex?

  binary (2 categories), nominal (>2 cat.)

- Ordinal: satisfaction-rating, age-group

Operators (where?)

- +. -, x
- >, <
- =, ≠

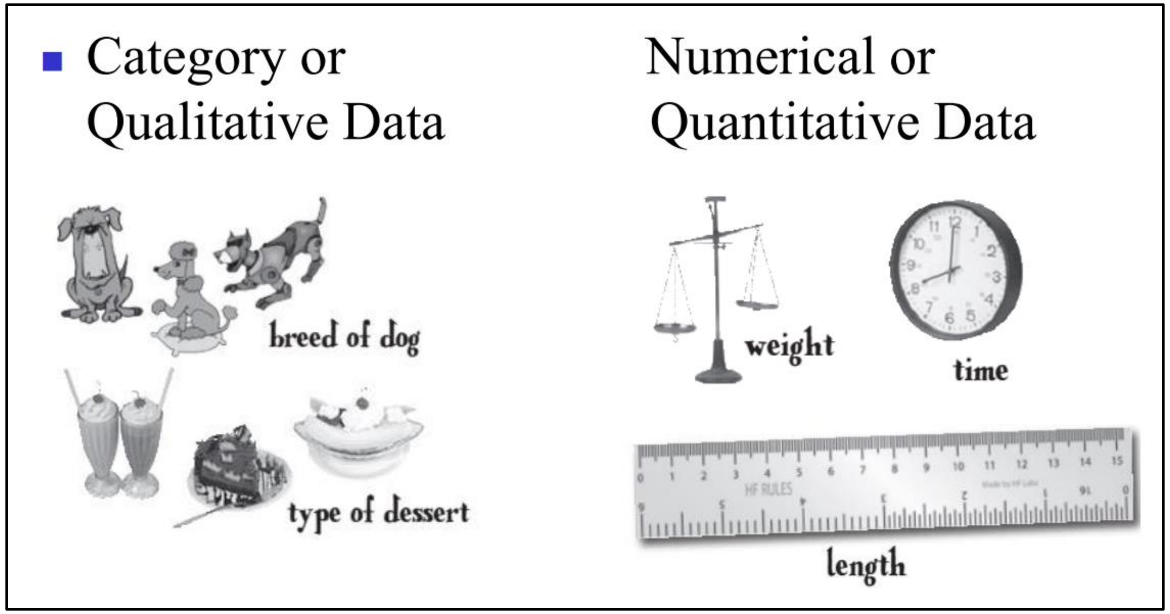
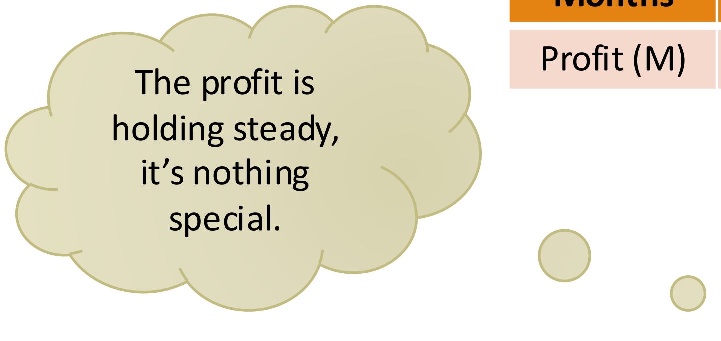


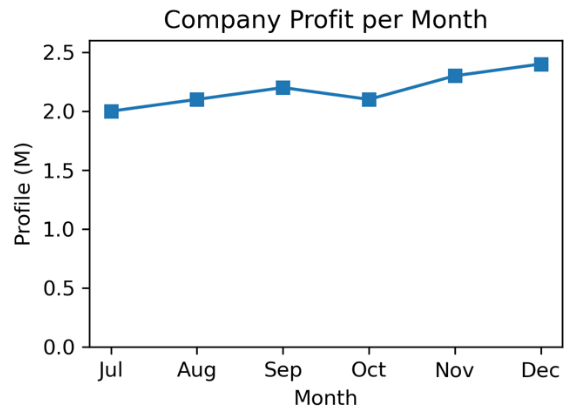
Image Ref: Book: First Head “Statistics” O’Relley

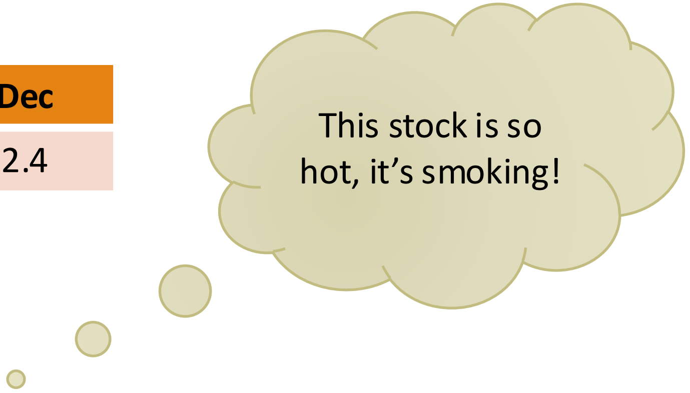
Visualisation – just plotting the data

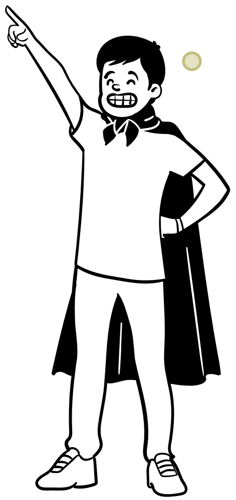
Impression of visualization: Profit of company

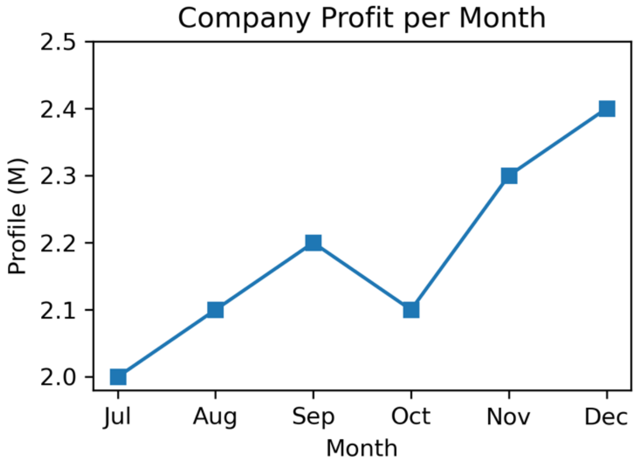
Months	Jul	Aug	Sep	Oct	Nov	Dec
Profit (M)	2.0	2.1	2.2	2.1	2.3	2.4











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Lab

Plot & type of variable

Numerical

- Quantitative: (height)

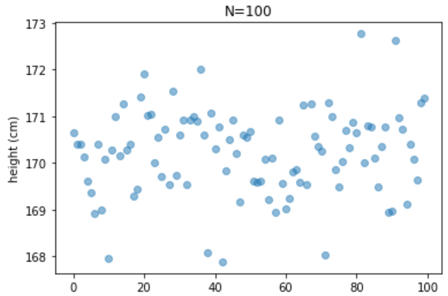
Categorical

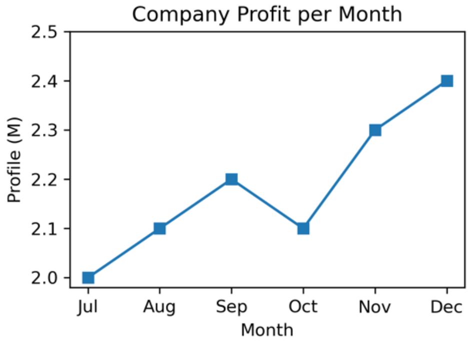
- Qualitative:

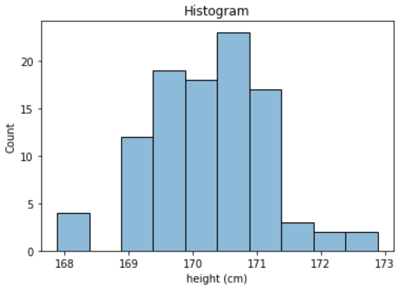
Genre	Unit Sold
Sports	27,500
Strategy	11,500
Action	6,000
Shooter	3,500
Other	1,500

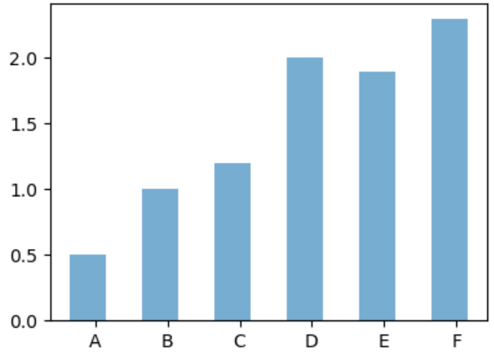
- Ordinal:

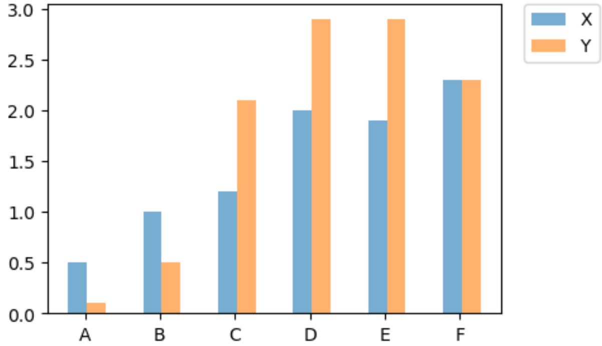
Hours	Frequency
0 --1	4,300
1--3	6,900
3--5	2,000
5--10	1,000
10--24	3,000











Descriptive Statistics

Summarizing the data

- Average: Mean, Mode, Median
- Frequency distribution
- Spread/variability: Range, Percentile, Standard deviation
- Skewness, Outliers

- What?, When?, Which?

Average: mean, mode, median

Most representative value of data

- Height in class

  [4, 4, 5, 4, 4, 4, 5, 4, 3, 5.5, 6, 4.5, 4.2, 5.2, 5,5, 6,1]

- Preference of drink

  [tea, tea, coffee, coffee, tea, tea, milk, tea, coffee, coffee]

- Age-group

  [10-15, 10-15, 10-15, 15-20, 20-25, 20-25, 20-25]

Mean: sum of all values/ number of values

Median: middle value of sorted sequence

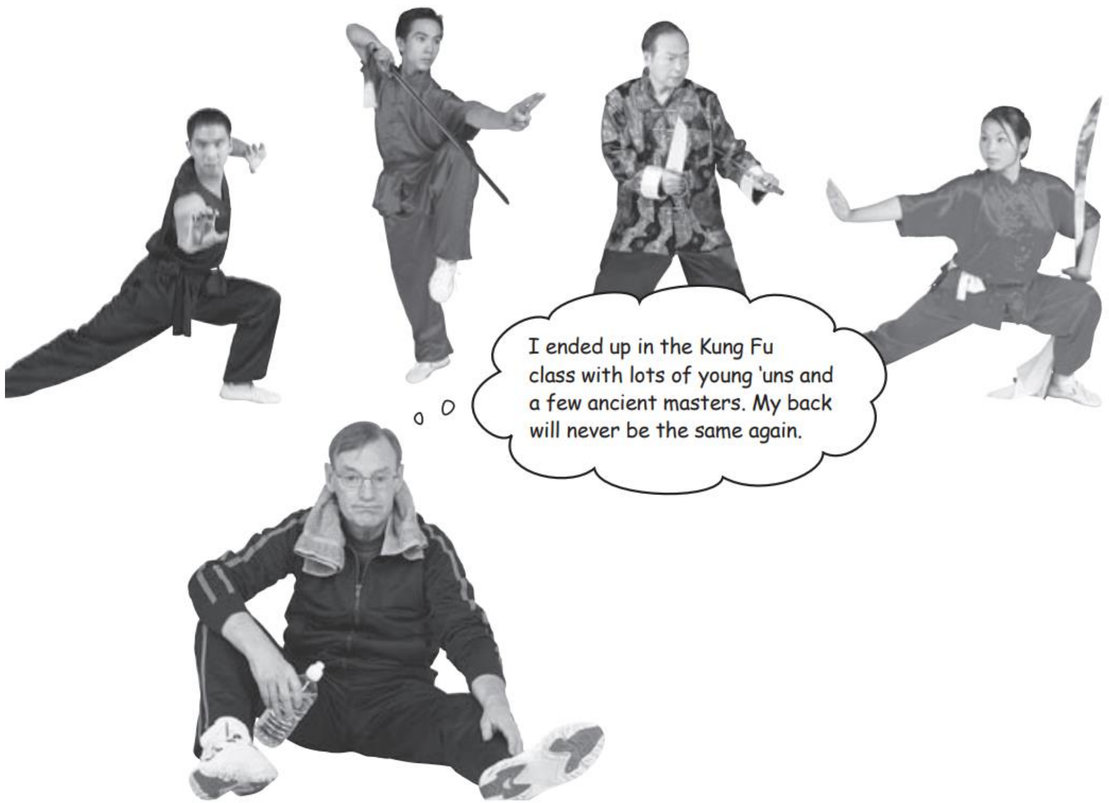
Mode : most frequent value

Let’s see a case:
A Health Club

The London Health Club
A perfect match for you

“London Health club proud to have class for everybody“

New customer in 50s



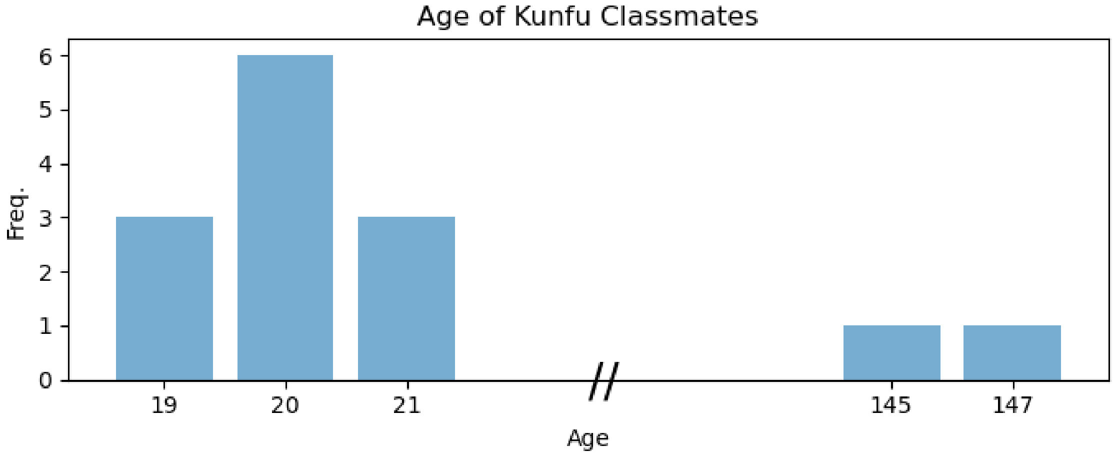
- Tuesday Evening

Class	**Mean age**
Class 1 :	17
Class 2 :	25
Class 3 :	38

Which class new customer should attend??

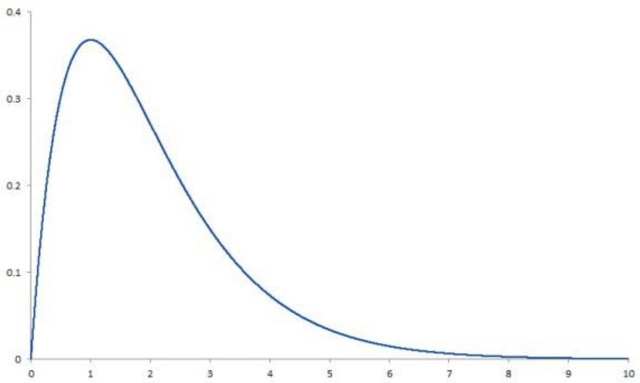
Health club

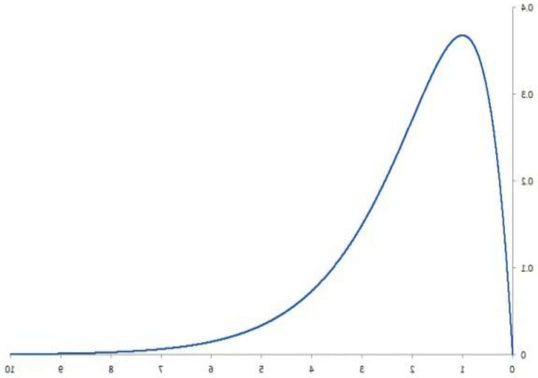
Age	19	20	21	145	147
Frequency	3	6	3	1	1



Median saved the day

Skewed distribution





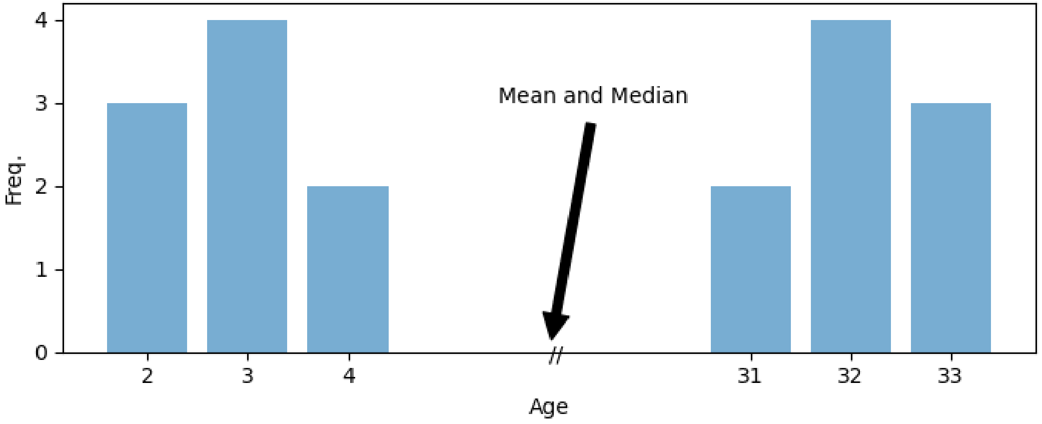
Median: 19 19 20 20 20 21 21 100 102

Here's the number in the middle. This is the median, 20.

Health club

teenager

Can anything go wrong?

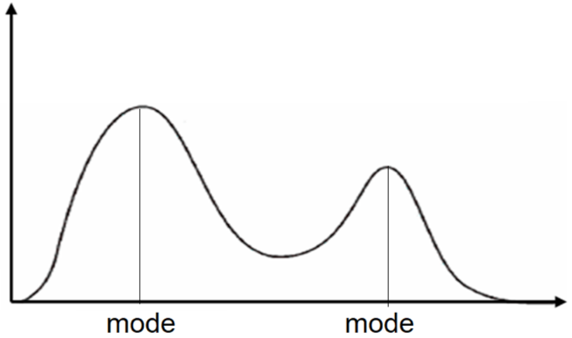


Class was for parents who bring their children to teach swimming?

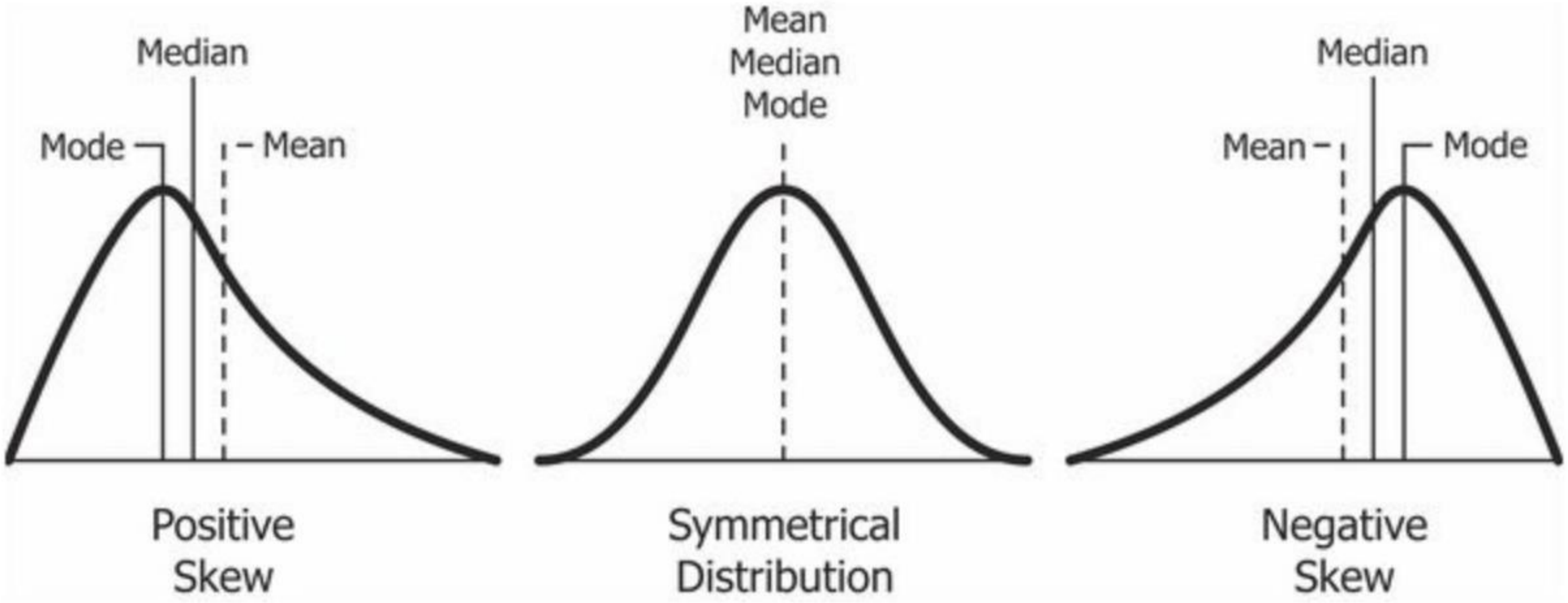
Mode is our solution

Mean = 17

Median = 17



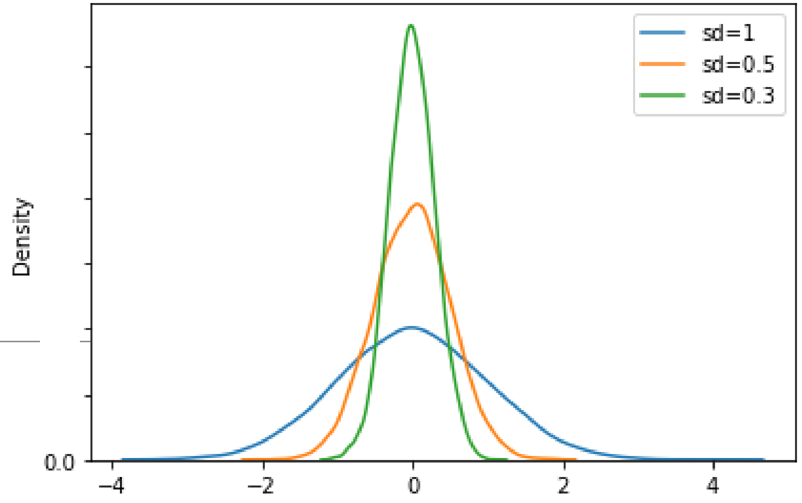
Skewness

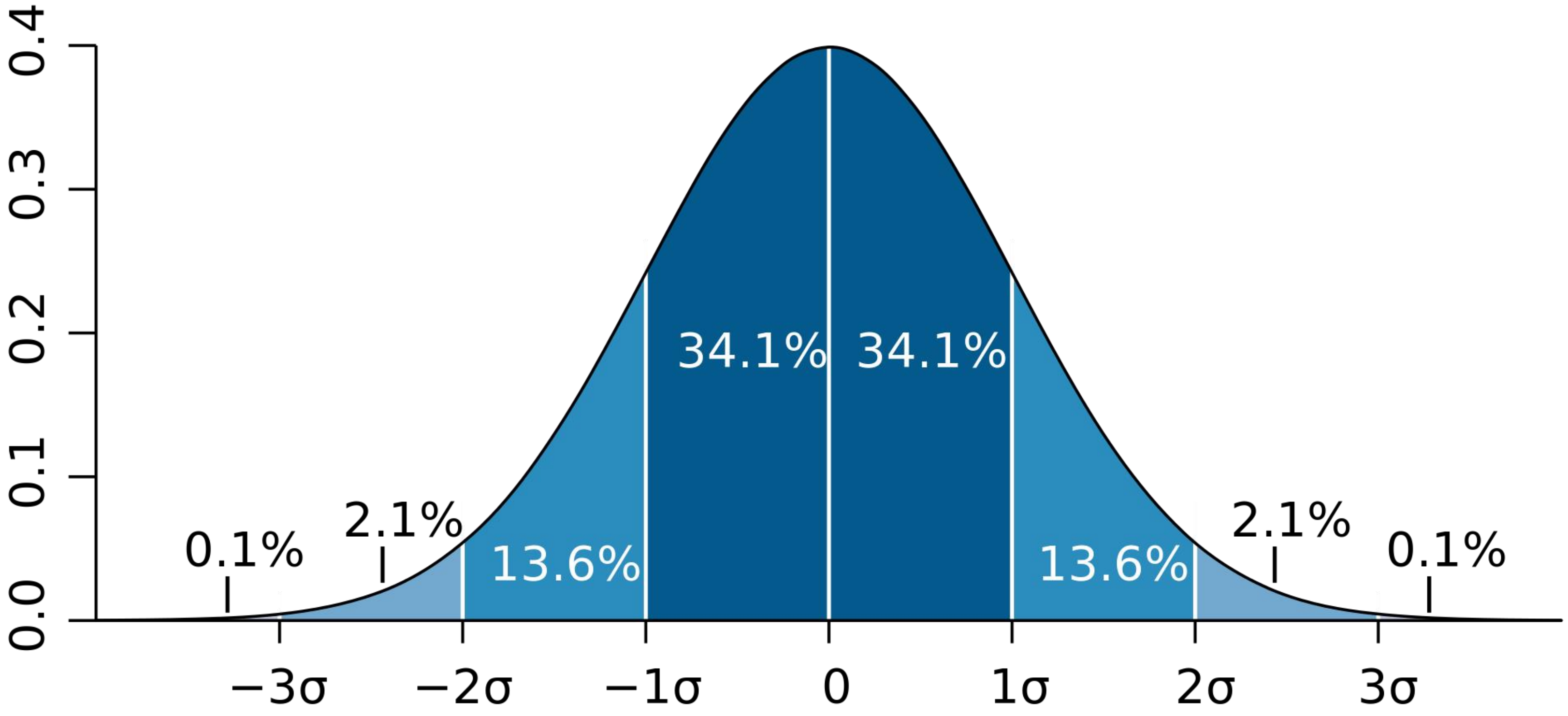


https://en.wikipedia.org/wiki/Skewness

Spread/Variability

Normal distribution





- Standard deviation σ *(sd)*

- Variance σ^2 *(var)*

- Range

- Interquartile Range (IQR)

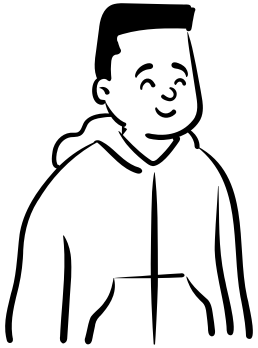
https://en.wikipedia.org/wiki/Normal_distribution

Question

You are a coach for a cricket (or football) team, need to hire a new player

- Player 1, **mean score** of run rate (or goal rate) is 8, with **standard deviation** of 4
- Player 2, mean score of run rate (or goal rate) is 8, with **standard deviation** of 2

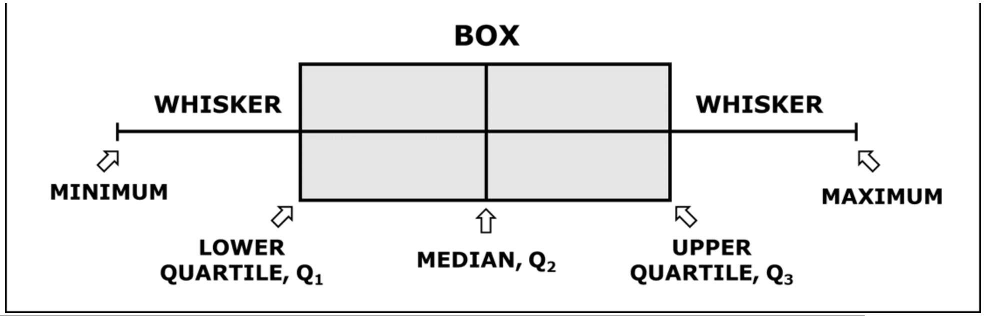
Who would you hire?

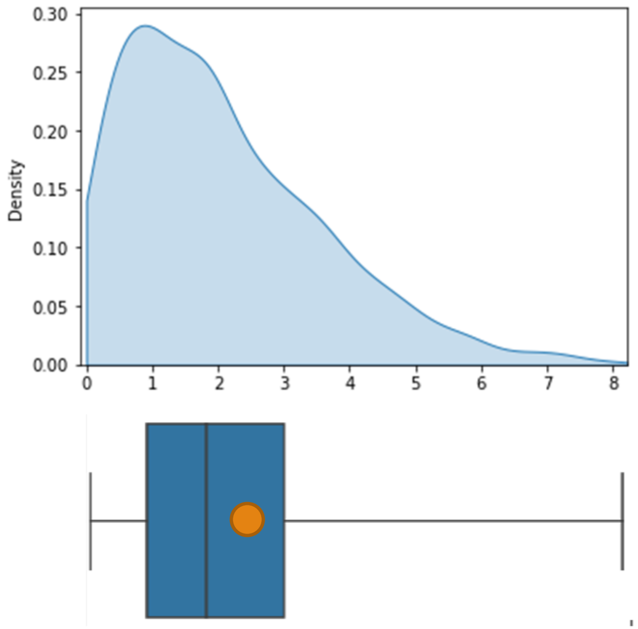


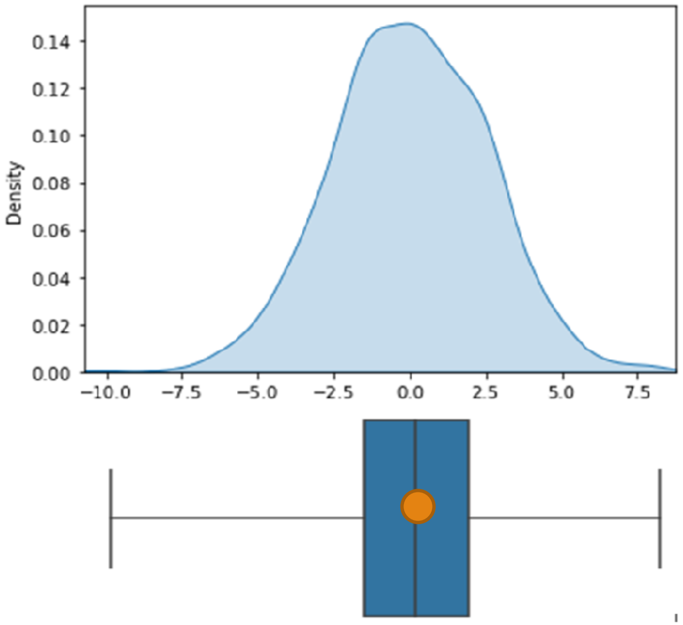
Player 1

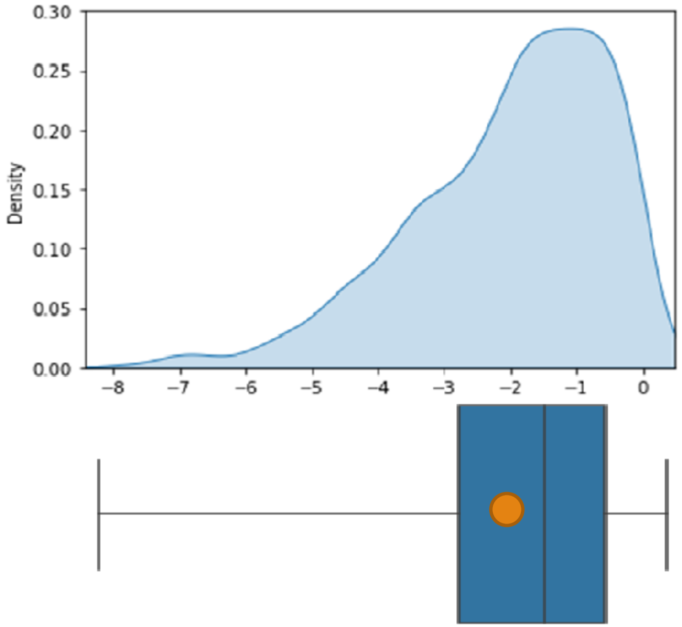
Player 2

Box-whisker plot





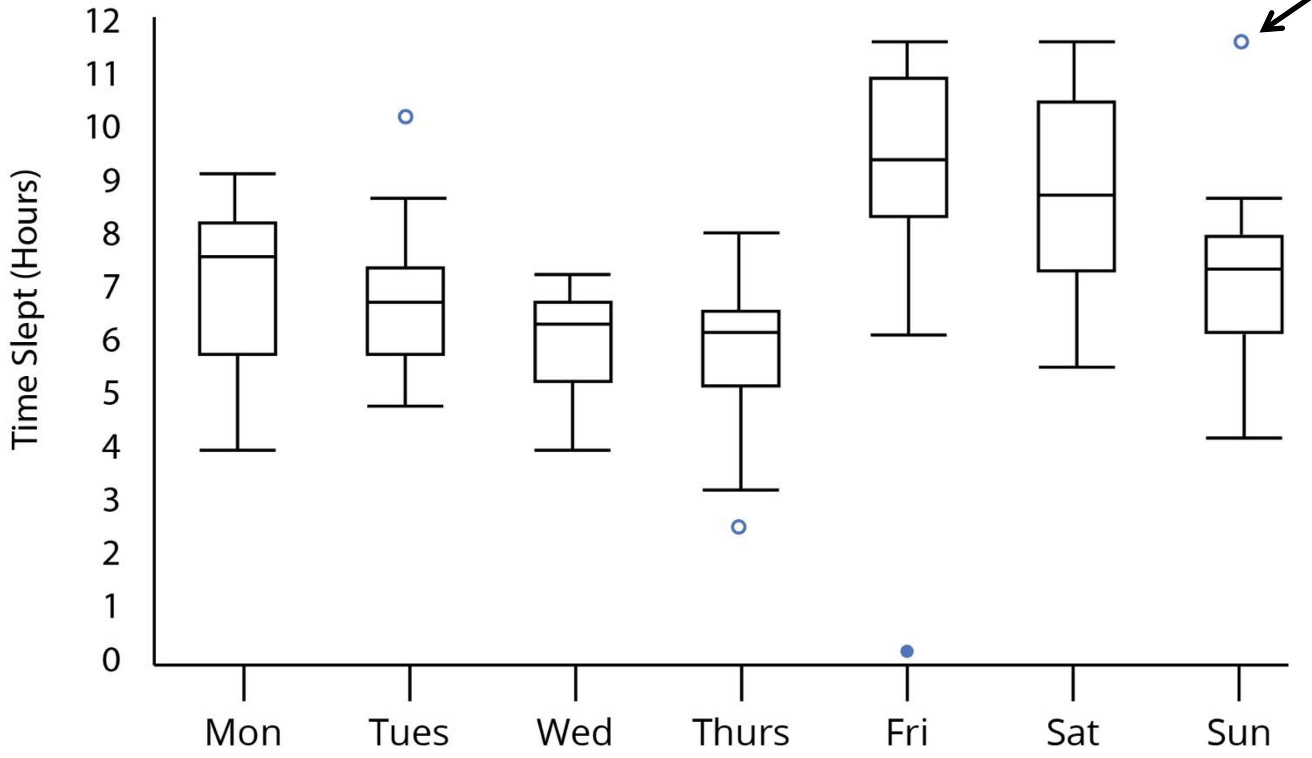




Visual comparison with boxplot

<u>Outlier</u>:
> Q3 + 1.5xIQR
< Q1 - 1.5xIQR

For a year, number of hours person P sleeps on different days



Questions:

1. On which day, P sleeps very less/ a lot?
2. Day with most (in)consistent hours of sleep

Lab

Can we categorise cont. data?

- To improve the interpretation

- Example : BMI into 2 or 3 categories - high or low BMI

- Implications?
  - Loss of information – loss of statistical power to detect the differences
  - Impact of choosing –where to cut
  - Splitting at median (dichotomising) - reduces statistical power

- Worst for binary than 4 or more categories

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Introduction to Statistics

Part 2: Inferential Statistics

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Inferential Statistics

Inferential Statistics

- Sample and Population

- Estimate population parameters from sample, and its accuracy (standard error)

- Standard error and standard deviation

- Confidence interval

- Size of data

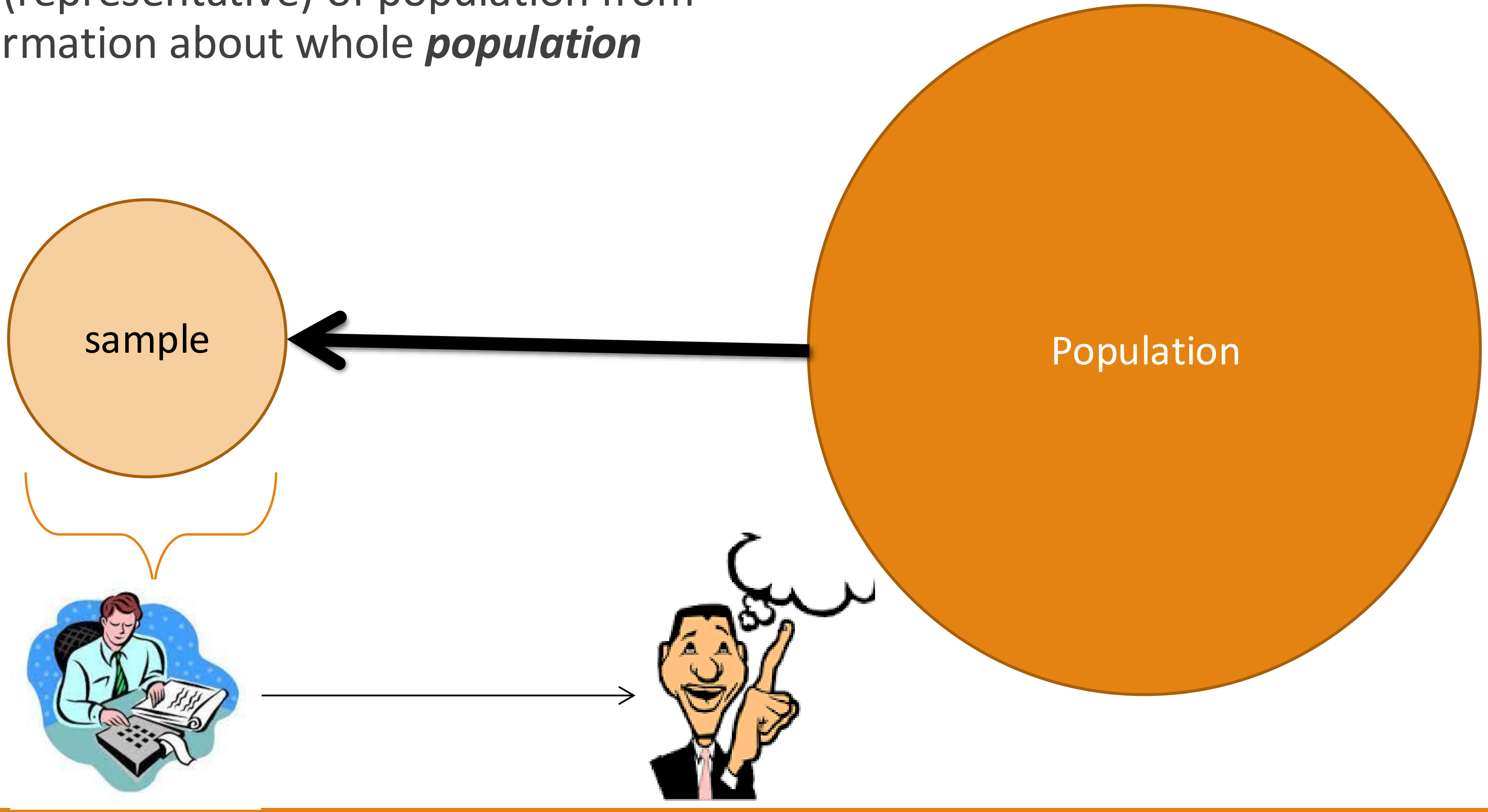
Population: A complete set of individual (objects)

Sample aka sample-data
One point in sample-data aka item, point, ~~sample~~

Size of sample : N

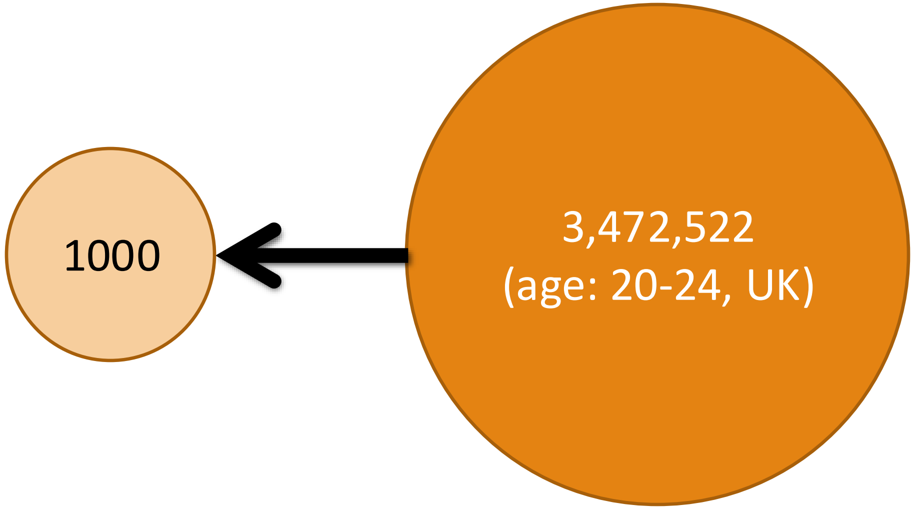
Population & Sample

Sample is a subset (representative) of population from which we infer information about whole ***population***



Examples:

1. Average weight of individual in UK of age 20-24.
2. Average marks of med. student in first year at Imperial.
3. Average Heart Rate of patients with High Blood Pressure in UK
4. Average height of Italian men

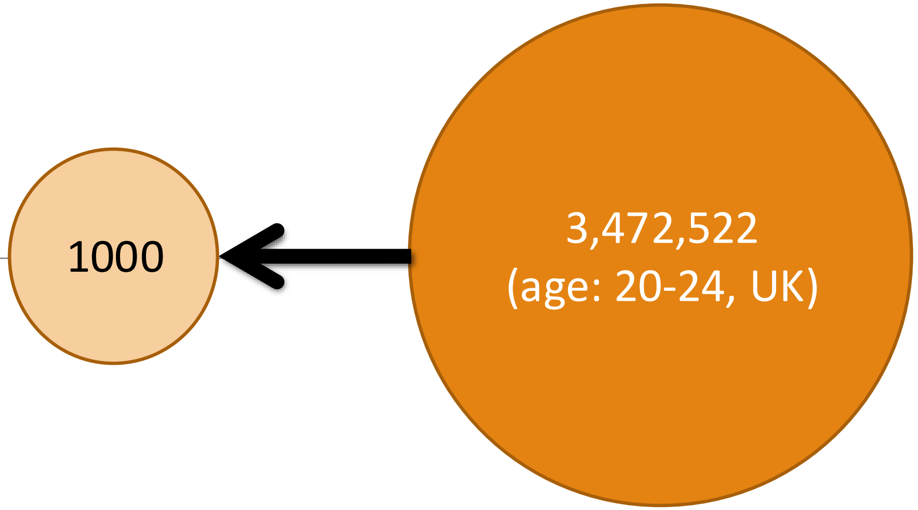


Q: Why take sample, why not entire population?

Q: How to sample (extract points/items from population)

Q: How many points (samples, items) are required in a sample

Estimation of parameter



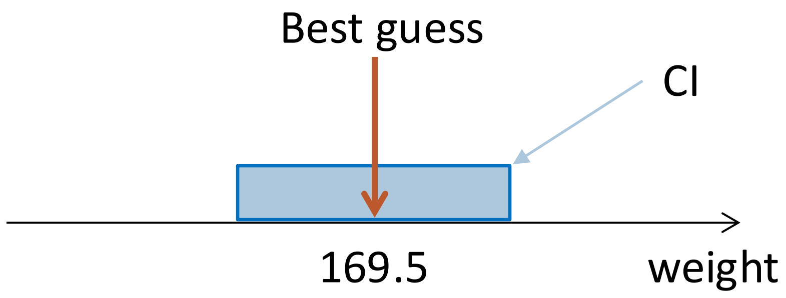
Mean weight of population

Best guess:

- Sample mean is estimation of population mean
- Uncertainty of this estimation (not exact value)

Accuracy of best guess

- Standard Error (SE)

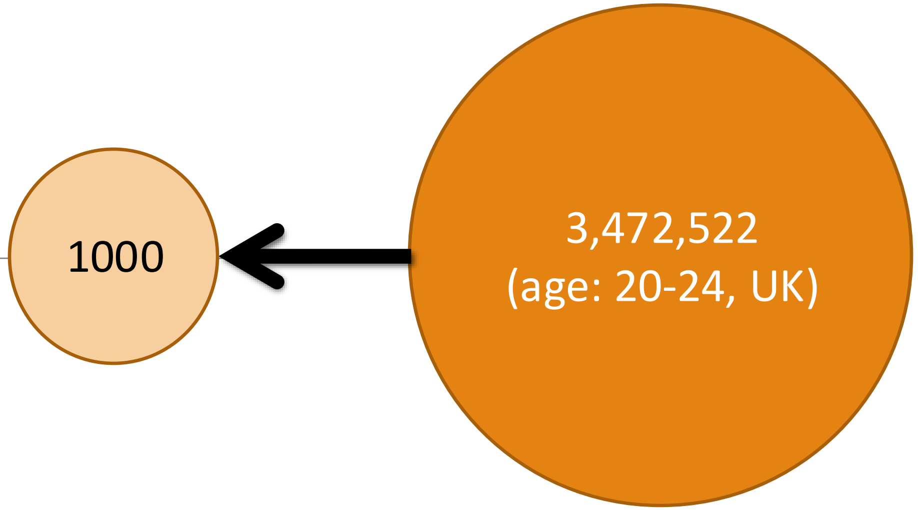


CI for the mean weight	
Mean	95%CI
169.5	163.7 – 174.2

Plausible values for true unknown

- Confidence Interval (CI)

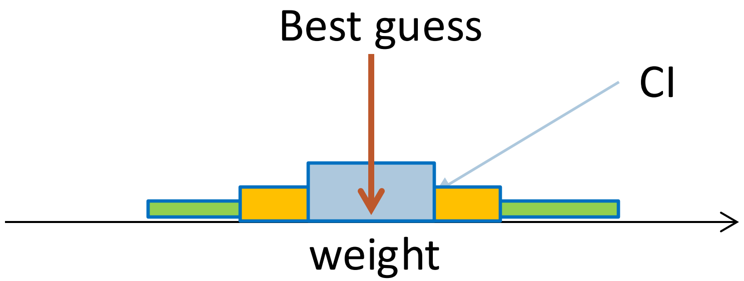
Estimation of parameter



<u>Mean weight of population</u>

Best guess:

- Sample mean weight of is estimation of population mean weight
- Uncertainty of this estimation (not exact value)



Accuracy of best guess

- Standard Error (SE)

Plausible values for true unknown

- Confidence Interval (CI)

CI for the mean weight			
Mean	90%CI	95%CI	99%CI
169.5	165.5-173.4	163.7 – 174.2	163.0-175.9

Standard Error and CI

Standard Error:

$$SE = \frac{SD\ of\ population}{\sqrt{sample\ size}} \approx \frac{SD\ of\ sample}{\sqrt{sample\ size}}$$

95% Confidence Interval

$$95\%\ CI = sample\ mean\ \pm 1.96 \times SE$$

Means of all (hypothetical) samples follow normal distribution and 95% of them lie within mean ± 1.96xSE

Standard Deviation Vs Standard Error

SD:

- measure of spread/variability of data
- descriptive statistics
- for normally distributed data, 2SD includes 95% of observed values

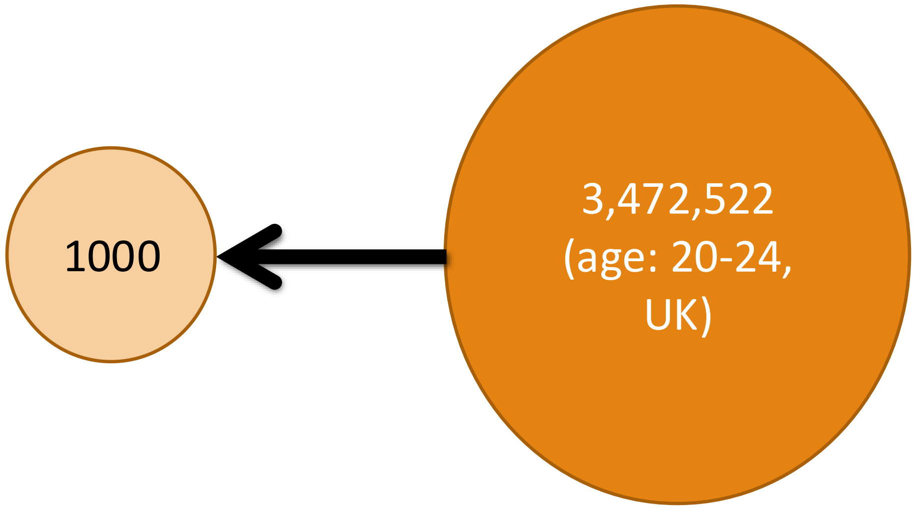
SE:

- accuracy of estimation of population
- inferential statistics
- for 95% CI, hypothesis testing etc
- range of values likely to include true population parameter

When it can be misleading?

- Sample is not **representative** of population (validity)

- **Not large enough** data (accuracy)



Lab

Let’s compute – Lab session

Data

Sample mean

Standard deviation

Standard error

Confidence interval

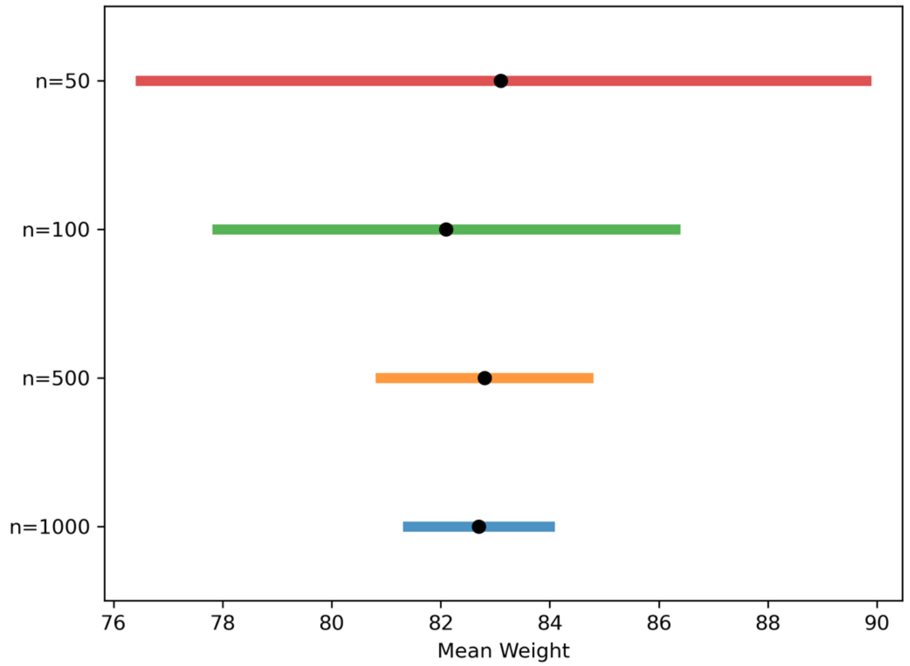
Effect of sample size: Example 1

Weight:

Sample mean = 81.4 kg, Standard Deviation = 21.4kg, n=1000

SE = ?, CI= ?

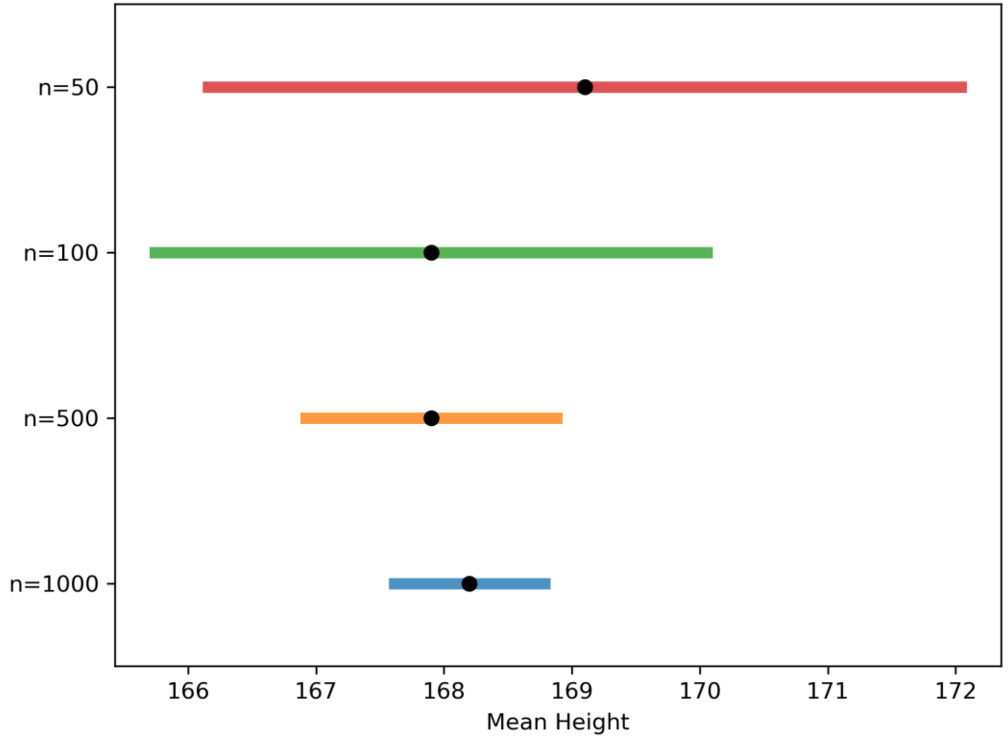
N of sample	Weight (mean)	Weight (SE)	95%CI
50	83.1	3.3	76.5-89.8
100	82.1	2.1	77.9-86.3
500	82.8	0.96	80.9-84.7
1000	82.7	0.68	81.4-84.0



Effect of sample size: Example 2

Height:

N	Mean	SE	CI
50	169.1	1.5	166.16-172.04
100	167.9	1.1	165.744-170.056
500	167.9	0.5	166.92-168.88
1000	168.2	0.3	167.612-168.788



Simulation: Sampling

https://nikeshbajaj.github.io/teaching/demos/Stats/sampling

Others

https://onlinestatbook.com/stat_sim/sampling_dist/index.html

https://onlinestatbook.com/2/index.html

Lab

Proportion and CI

Estimating proportion of population that have particulate condition A.

Find the proportion in sample p = #A/total

$$SE\ of\ the\ proportion\ p\ = \sqrt{\frac{p(1-p)}{n}}$$

$$95\%CI\ of\ the\ proportion\ p\ = p \pm 1.96 \times SE$$

**np>5 & n(1-p)>5*

Example:
Find proportion of obese people (BMI>30), given sample of 1000 people, among which 391 are obese.

p = 391/1000 = 0.391
SE = 0.0154
95%CI = 0.362 to 0.422

Question:
151 have asthma in 1000, compute..?

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Introduction to Statistics

Part 3: Hypothesis: Design & Testing

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Given two groups of data

TEST FOR DIFFERENCES

TEST FOR ASSOCIATIONS

Hypothesis Testing (p-value)

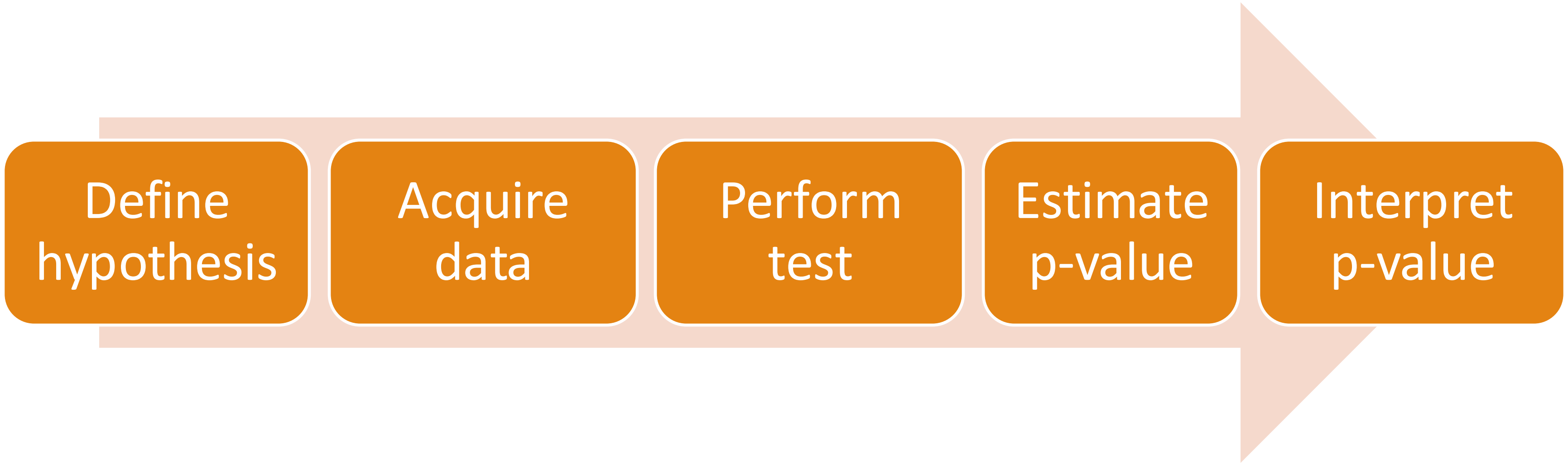
- Hypothesis? testing? P-value?

- Type I and type II error,

- Multiple testing and statistical power

Hypothesis and Testing

- Hypothesis: A statement about a true value of parameters and their relationship in a defined **population**
- Testing: The procedure, based on **sample,** to determine if the hypothesis is a ***reasonable statement****



You have a question/statement? → Define hypothesis

Define Hypothesis

In science to verify if a hypothesis is a reasonable statement, you need to test it against its contrary which is assumed to be true.

Null Hypothesis H0

Assumed to be true → No true difference or relationship between observed values in the sampled population

Alternative Hypothesis H1 { “Burden of Proof” }

To be proven → There **IS** true difference or relationship between ..

Example 1

You have a question/statement? → Define hypothesis

Is lung function different between genders?

H0: lung function in men = lung function in women

H1: lung function in men ≠ lung function in women

Alternative Hypothesis with sides

- H1: lung function in men ≠ lung function in women - two sided
- H1: lung function in men < lung function in women - one sided
- H1: lung function in men > lung function in women - one sided

You have a question/statement? → Define hypothesis

Example 2

Is height in a class A and class B different?

H0: ?

H1: ?

Question:

Can an alternative hypothesis be:

- **There is no difference between two samples?**
- **Two sample groups are same?**

Question:

You have a question/statement? → Define hypothesis

Example: Let's make some hypothesis

- *Your friend is **always** late and never agrees that he is.*

Statement?:

Acquire Data

Acquire Data

Acquiring Data:

- Conducting experiment in Lab
- Collecting data from other sources

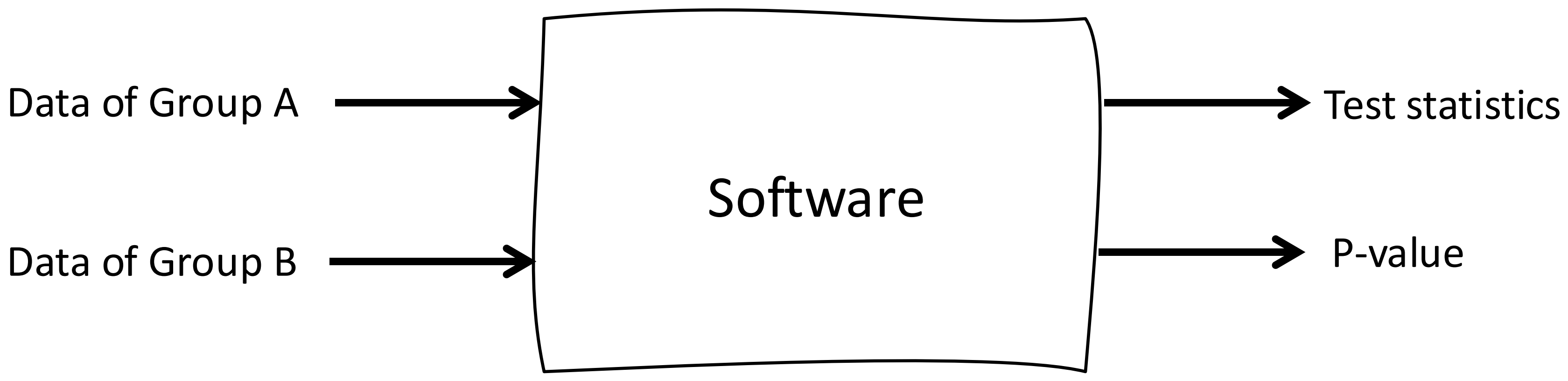
This is your sample data (think of population).

Ideally, this should be representative of population and large enough

Performing a test

Perform Test

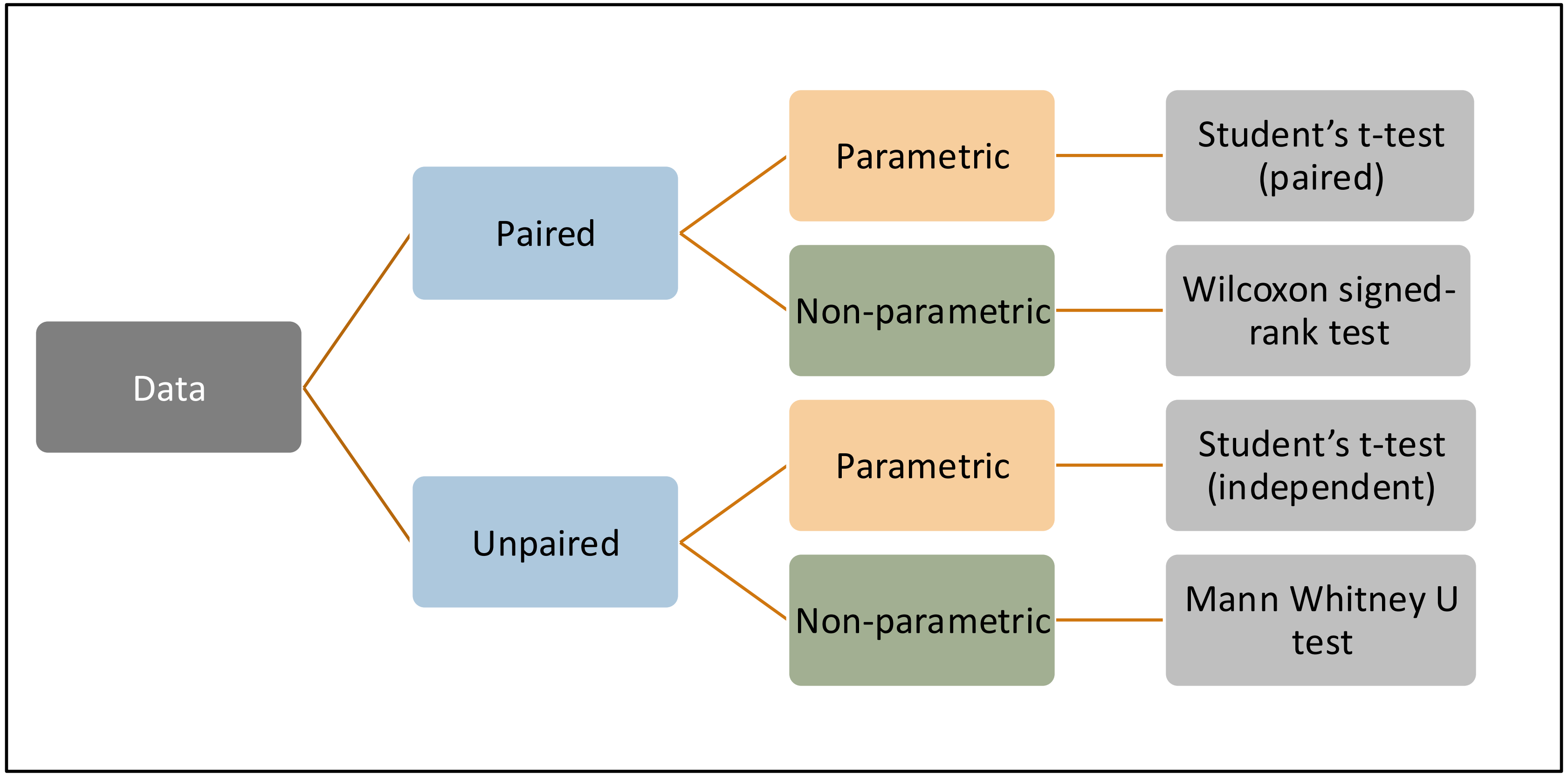
Performing a test on a two groups for establishing differences or association, we use software.



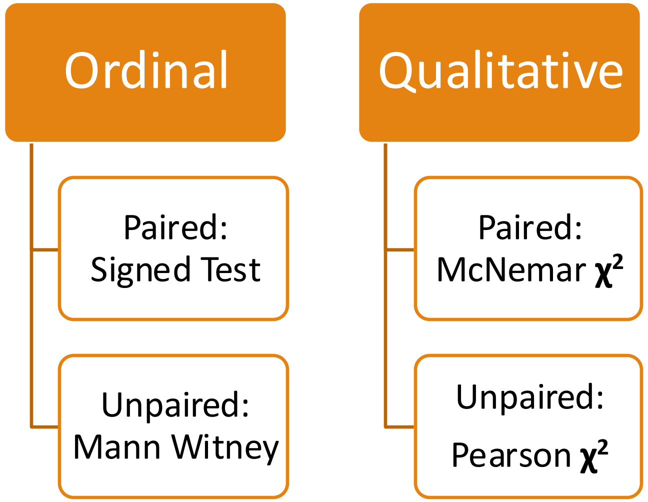
Good news! You don’t need to remember the formulas
But you need to know, which test to perform and how to the read results

Perform test
Choosing a right statistical test

Perform Test



We will see how to choose correct test



Test & P-value

Estimate p-value

With a right test, compute test statistics, that summarise the different/relationship in your sample.

- Use test statistics to compute p-value, that tells you either to ***accept or reject null hypothesis***

P-value: Probability of obtaining the difference/effect observed in given sample by pure effect of **chance**, when null hypothesis is true.

- If two samples comes from same population, how likely we see a difference between them

- Ranges from 0 to 1.
- To conclude a statistical significance , we need a cut-off value α (i.e. 0.05)
- NOT the probability of making a mistake!

P-value

Interpret p-value

Example:

- If p-value is <0.05, we are confident enough to reject the null hypothesis.
- α = 0.05 → 5% chance of rejecting null hypothesis, even if it is true.
- α = 0.05 → probability of committing a type I error

	Null hypothesis H0	
	True	Not True
Accept H0 (fail to reject H0, $p>\alpha$)	Right	Type 2 Error (False negative with probability β)
Reject H0 ($p<\alpha$)	Type 1 Error (False positive with probability α)	Right

https://www.slideshare.net/smulford/type-1-and-type-2-errors

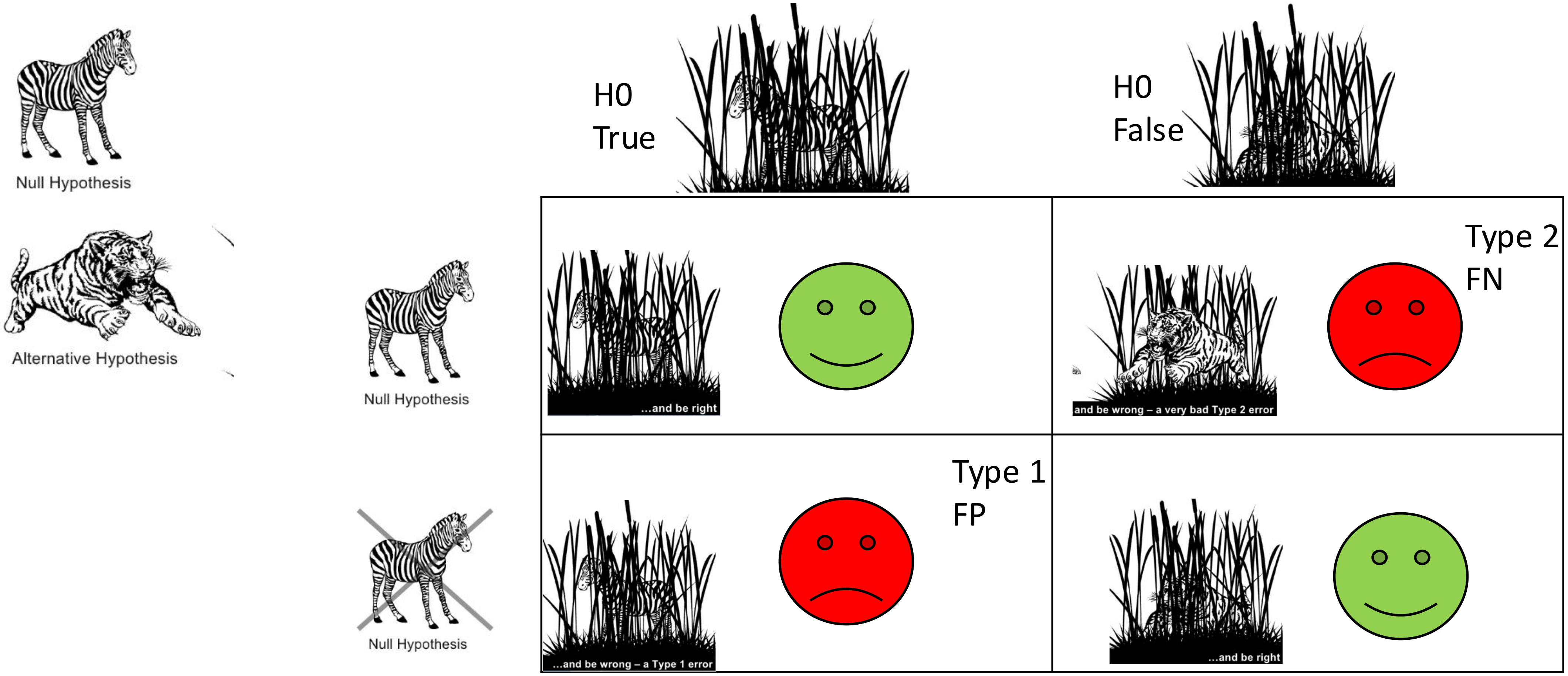
Interpret p-value

Type 1 and Type 2 Error

https://www.slideshare.net/smulford/type-1-and-type-2-errors

Interpret p-value

Type 1 and Type 2 Error



Type 1 and Type 2 Error

Interpret p-value

Type 1 Error (α):

- Rejecting TRUE null hypothesis
- False Positive
- α=0.05, 5% probability of false positive

Type 2 Error (β)

- Failing to reject FALSE null hypothesis
- False Negative
- Power is probability that we correctly rejects the Null Hypothesis
- Power = 0.8, β =0.20 (1-power), power 80%

A fun way to remember Type of Error

P-value: summary

Interpret p-value

Smaller the p-value, stronger the evidence against null hypothesis

If p-value is <0.05:

- It is unlikely that any difference found in samples are due to chance
- Reject the null hypothesis in favour of alternative hypothesis
- **Statistical significance**

If p-value is <0.001:

- Strong evidence of significant results

What is p=0.049 or p=0.051?

- p-value is a guideline to decide if results deserves second look
- Ref: Scientific method: Statistical errors

Lab

Multiple testing

Each test has a 5% chance of 1 false positive

So, running multiple tests increases the probability of false positive

On same dataset, testing for multiple outcomes, single outcomes in multiple sub-groups, or multiple effects.

- Before testing, limit your objectives and outcomes to be tested.
- If you have to apply for multiple testing, apply α correction methods (e.g. Bonferroni, α/n)

Significance and Meaningful

Statistically significant results does not always have meaningful relevance and vice-versa

Example:

- Two class groups A and B, have statistical significant ($p<0.001$) difference of 2 marks in given course.
- Men and women have a statistically significant difference of 0.5mL in lung function

Lab

Power and Sample size (N)

- Low power (due small sample size) increase the probability of False Negative

- You might find no difference between groups, and that might be False Negative, due to high β or low power (small sample size)

Example

Group	N	Mean	sd
Male	421	3555.20	909.75
Female	407	2500.77	625.99

Difference in mean = 1054.43
p-value <0.0001
T-statistics = 0.81

Group	N	Mean	sd
Male	12	3548.08	917.3
Female	8	2993.00	571.3

Difference in mean = 555.08
p-value = 0.4319
T-statistics = 0.81

Lab

Sample Size calculation

For two groups with mean m1 and m2, and standard deviation of sd, we need N samples in each group to be able to reject a null hypothesis with probability of False positive as 5% and probability of False negative of 80%

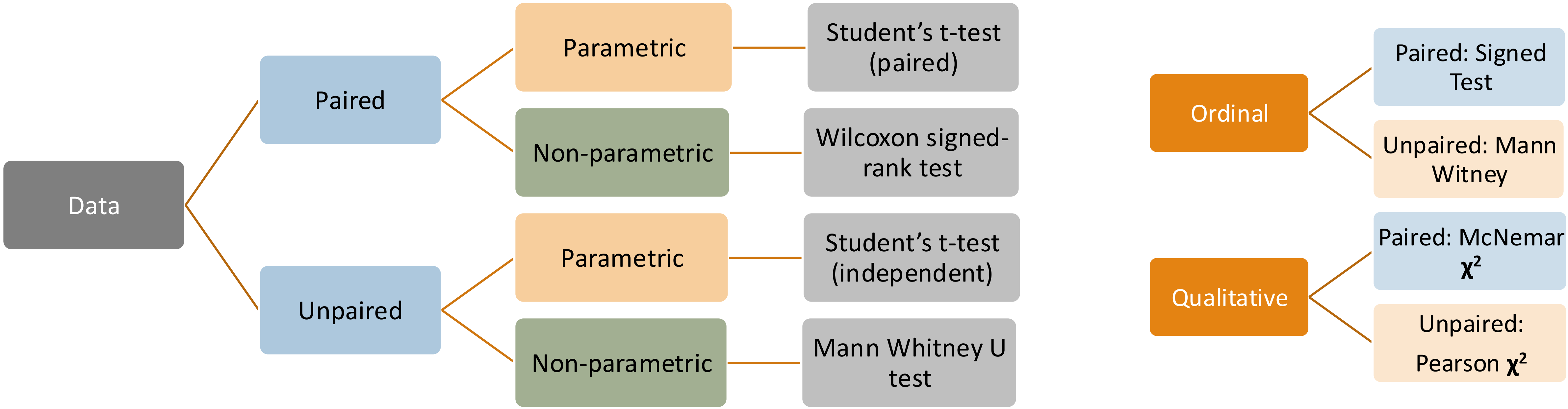
$$N\ in\ each\ group = f(\alpha, \beta) \times \frac{2(sd^2)}{(m_2 - m_1)^2}$$

$$f(\alpha, \beta) = f(0.05, 0.20) = 7.85$$

Don’t worry about the formula, it is available in all the software.

Important thing to notice → smaller the difference you like to detect, more samples you need, smaller the sd is less sample you need

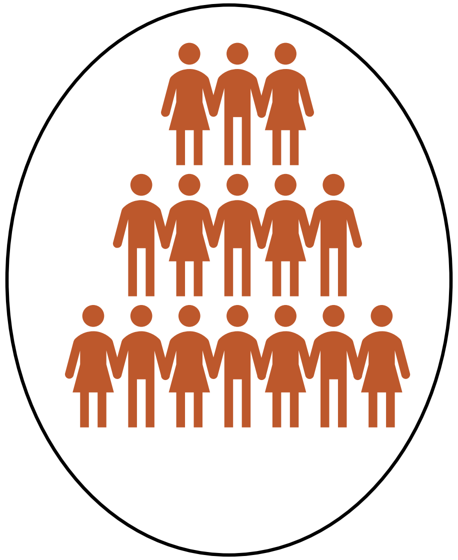
Choosing the correct test



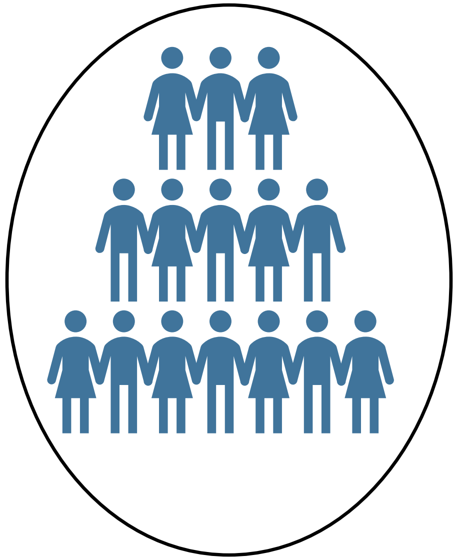
Examples?

Paired & Unpaired

Unpaired

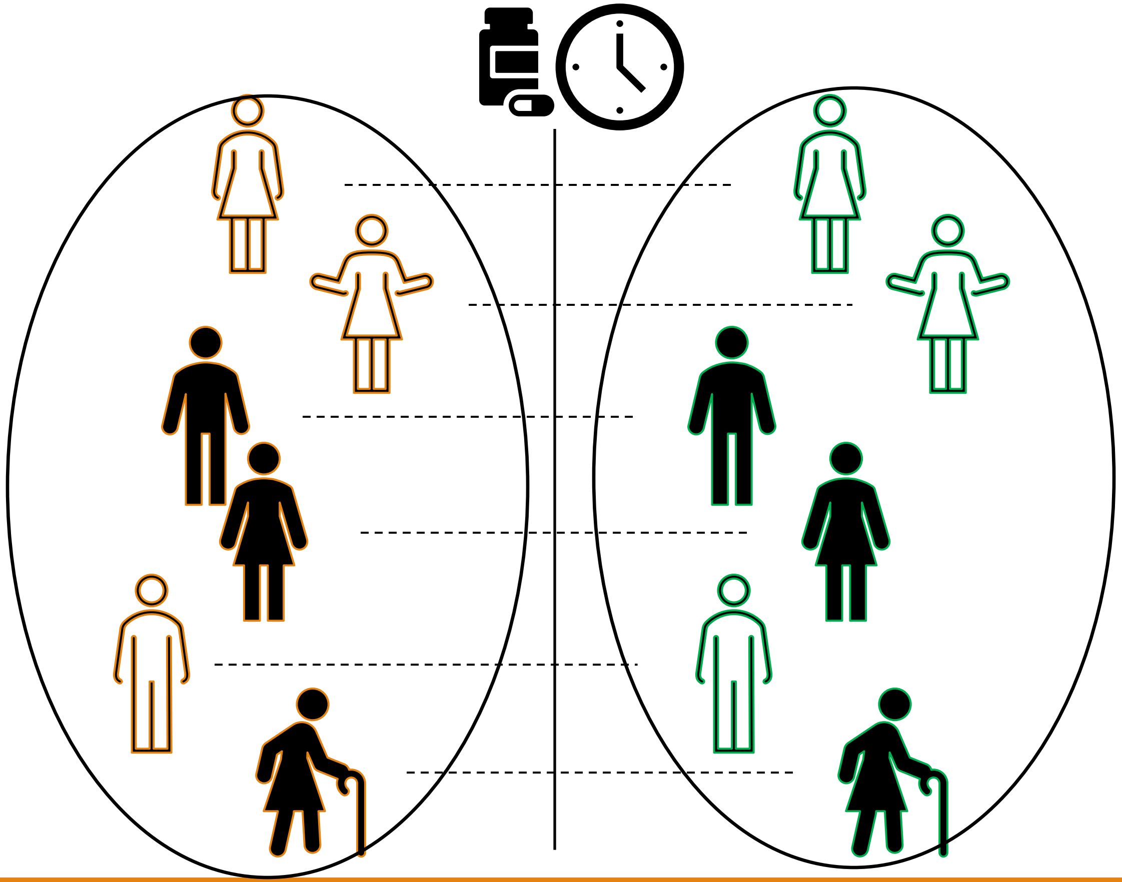


A



B

Paired



Paired & Unpaired

Unpaired (independent)	**Paired (dependent)**
- Data collected from each sample is independents of time (usually collected once)	- Data collected from same subjects at different time (before and after treatment)
- Different subjects in different groups	- Same subjects in different groups
- BMI, age, height of subject	- BMI, before and after a treatment

Parametric & Non-parametric

Parametric Tests

- Relies on the underlaying statistical distribution
- Normally distributed data (**normality test**)

Non-parametric Tests

- Do not depend on any distribution

For Categorical Data

Numerical

- Quantitative**:** blood pressure, sugar level, no of cells, height, BMI

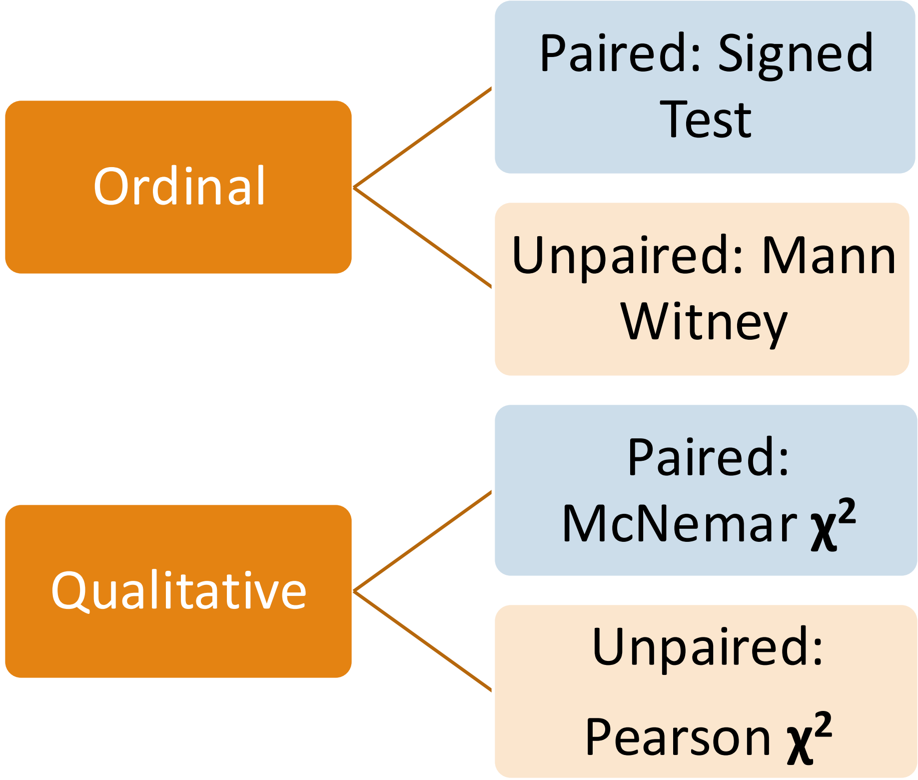
  \- Cont., Discrete

Categorical

- **Qualitative**: ethnicity, disease or not? , sex?

  \- Binary (2), nominal (>2)

- **Ordinal**: satisfaction-rating, age-group



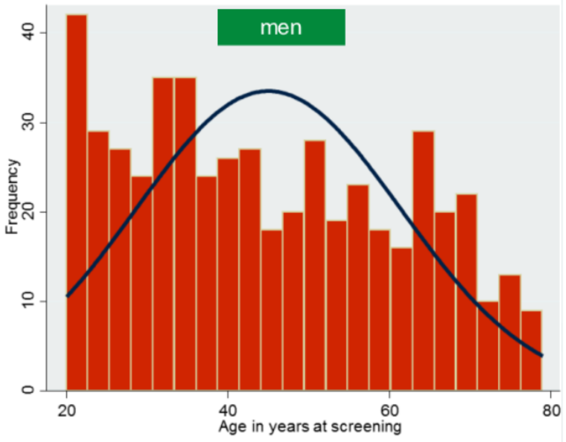
Example 1

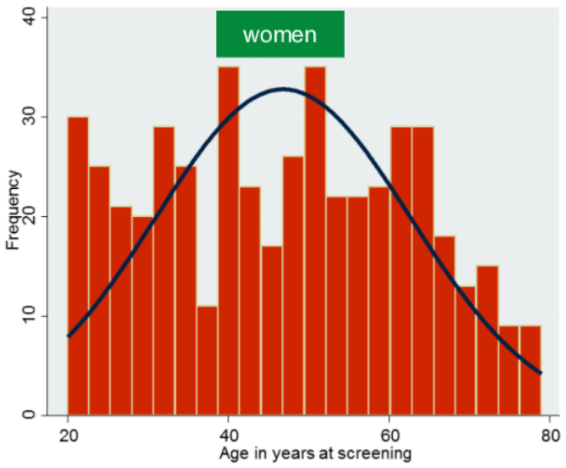
Is average age of men and women different in given sample (dataset)?*

H0: $\mu_{men} = \mu_{women}$

H1: $\mu_{men} \neq \mu_{women}$

Paired? Normality? Equal variance?





Group	N	mean age	sd	sd^2
Men	514	45	16.41	269.38
Women	486	46.83	15.87	251.81

Men → Test? → Test statistics

Women → Test? → P-value

Test statistics: 1.79

P-value: 0.0741

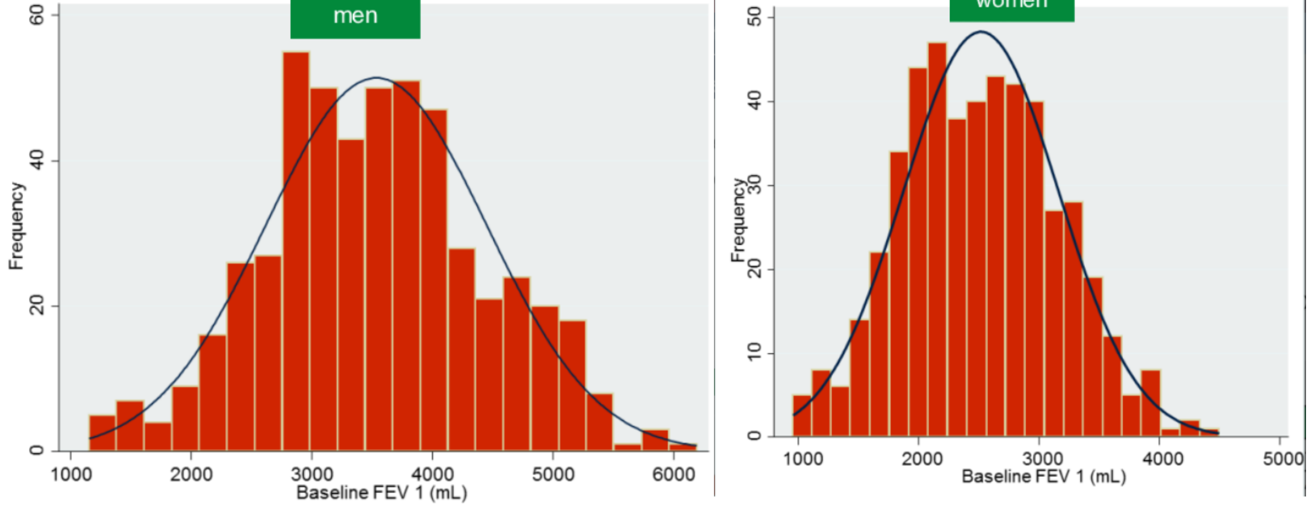
Age difference?

Example 2

Is the average lung function in men different to one in women in general population

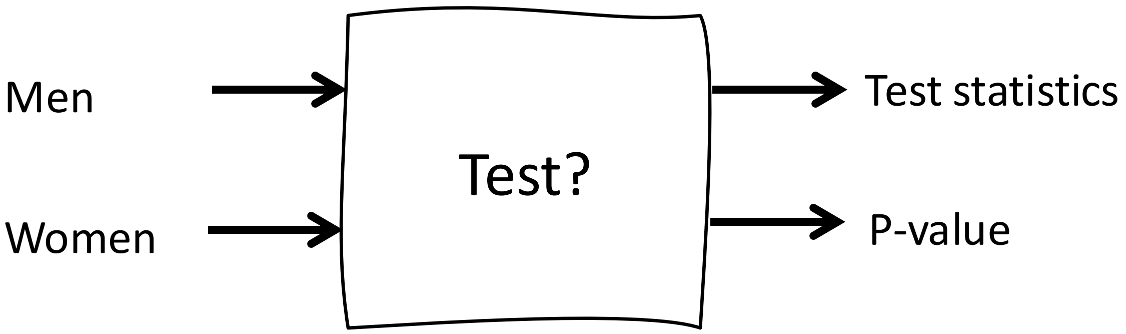
H0: $\mu_{men} = \mu_{women}$

H1: $\mu_{men} \neq \mu_{women}$



Group	N	mean fev1(mL)	s d
Men	514	3535.08	915.08
Women	486	2515.17	646.08

Paired? Normality? Equal variance?



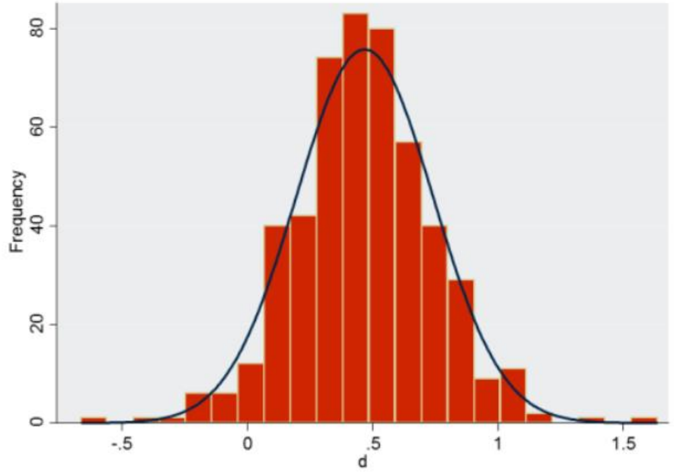
Test statistics: 20.26

P-value: 0.001

difference?

Example 3

Has the lung function changed after intense exercise in the study



H0: $\mu_{before} = \mu_{after}$: $\boldsymbol{\mu_d = 0}$

H1: $\mu_{before} \neq \mu_{after}$: $\boldsymbol{\mu_d \neq 0}$

	N	mean	median	sd	sd
d	496	0.47	0.47	0.27	0.01

Paired? Normality? Equal variance?

Men → Test? → Test statistics

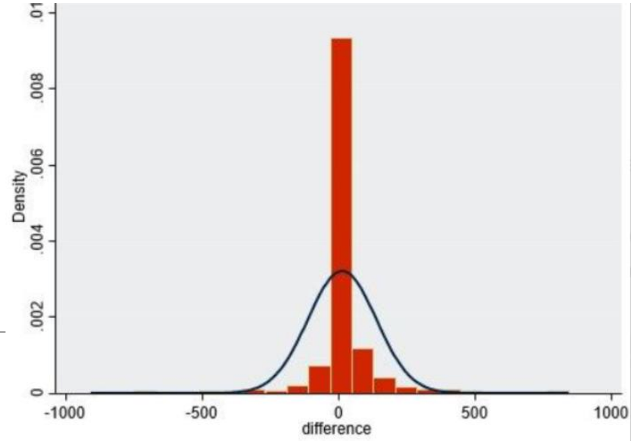
Women → Test? → P-value

Test statistics: 38.13

P-value<0.001

difference?

Example 4

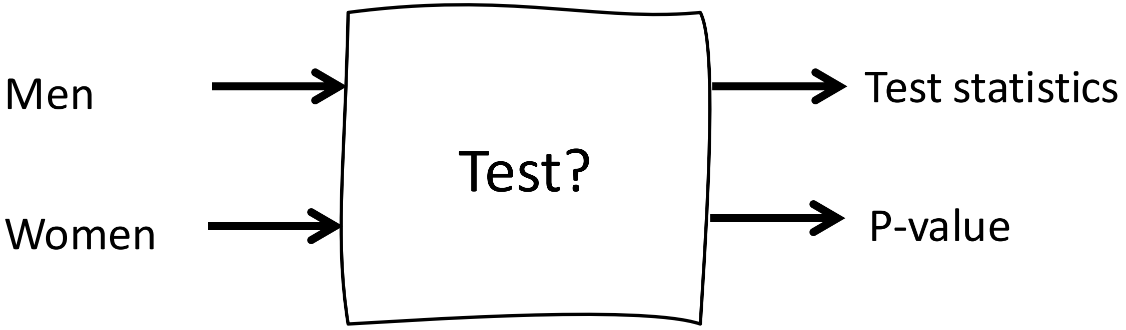


Has total immunoglobulin E (IgE) changed over time (over 10 years)

H0: $\mu_{before} = \mu_{after}$: $\boldsymbol{\mu_d = 0}$

H1: $\mu_{before} \neq \mu_{after}$: $\boldsymbol{\mu_d \neq 0}$

Paired? Normality? Equal variance?



Test statistics: 10

P-value>0.05

Let’s give it a try

From the statement from list, let’s define a hypothesis, testing approach:

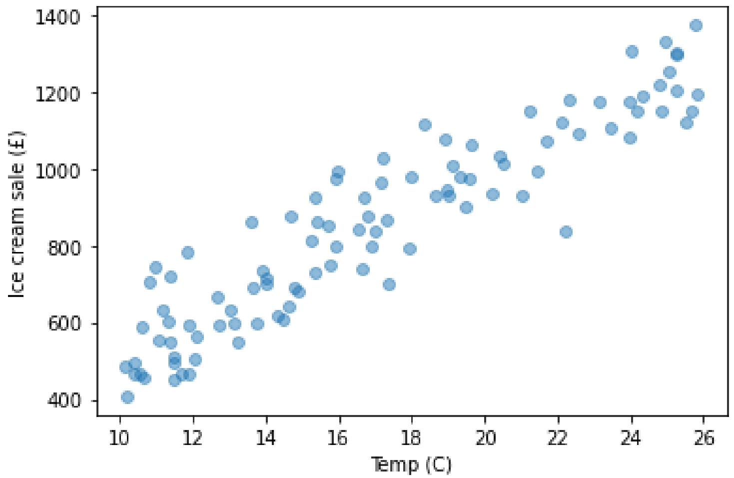
1. Effectiveness of a teaching method for a subject
2. Effectiveness of a drug on elderly >50 for lung function
3. Lung function of smokers and non-smokers

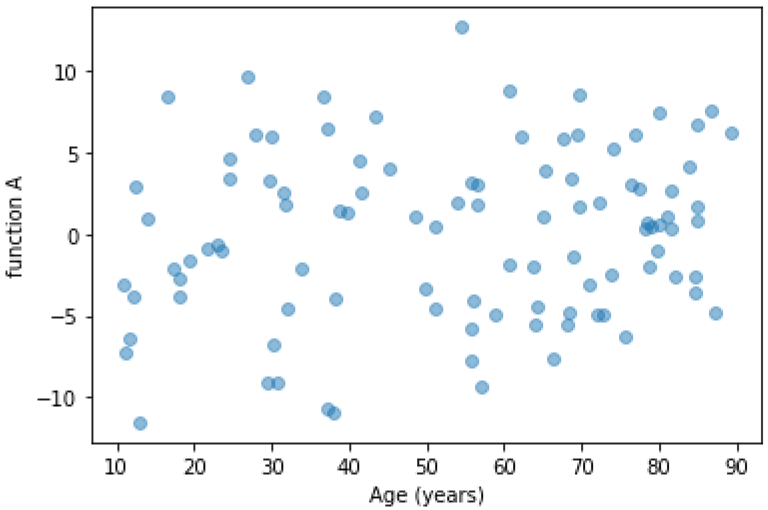
Association between two

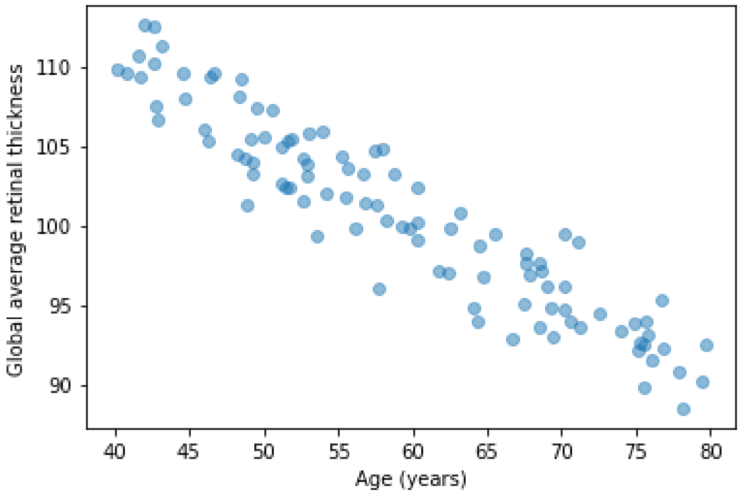
Correlation

- Investigate a relationship between two independent variables (i.e. x and y)
- Does x increase as y or vice-versa?
- Is relation linear?

One simple way is to plot scatter graph and see

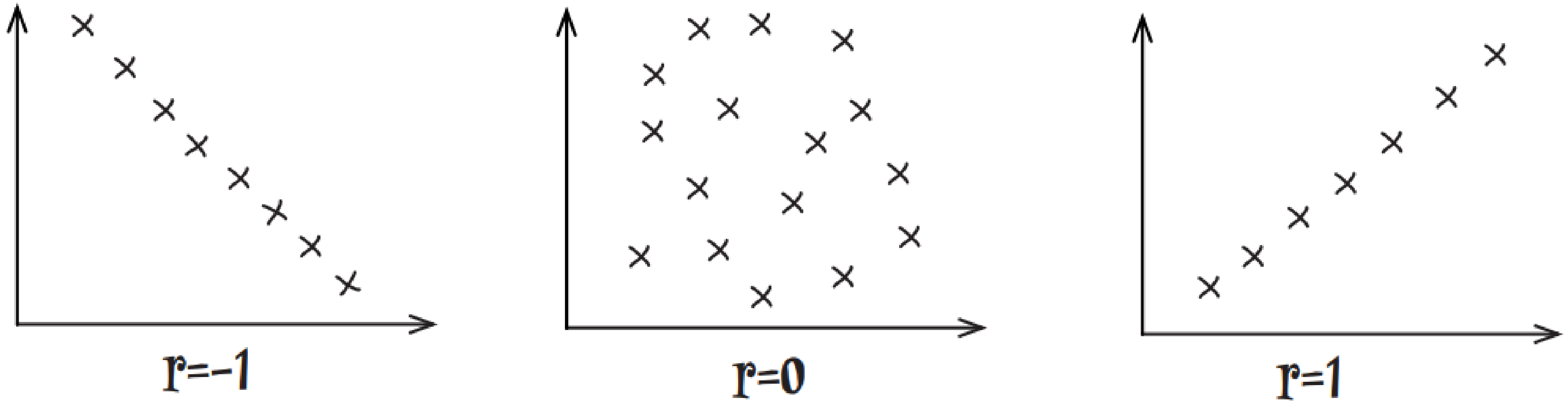




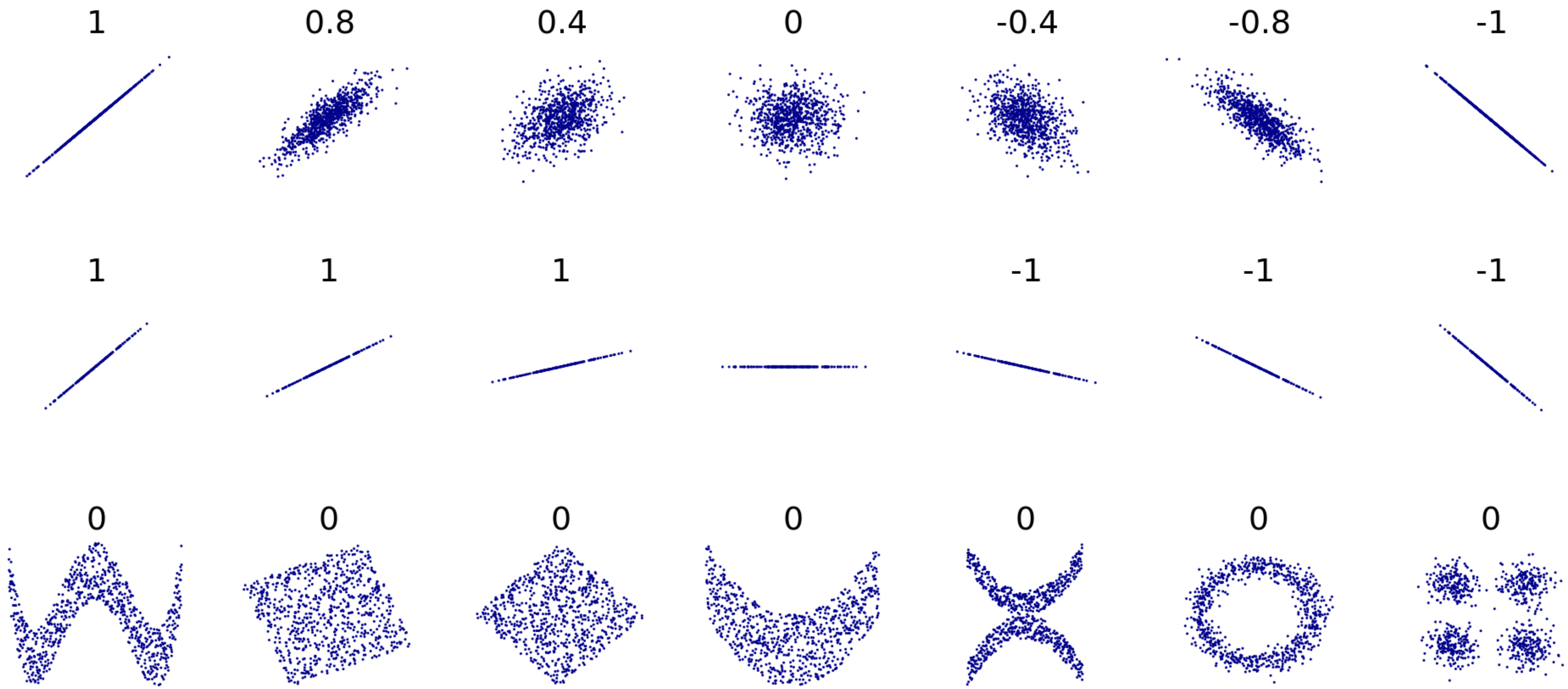


Quantifying Correlation

Pearson Correlation Coefficient r or ρ (rho)



Pearson Correlation Coefficient



https://en.wikipedia.org/wiki/Correlation

Correlation

Pearson Correlation Coefficient

Parametric test

- x, y: normally distributed
- linear relationship

Generally:

- $|r| < 0.4$ → weak
- $0.4 < |r| < 0.7$ → moderate
- $0.7 < |r|$ → strong

$r = 0$, no ***linear*** relationship

Spearman Rank correlation

Non-parametric test

Based on ranks rather than exact values

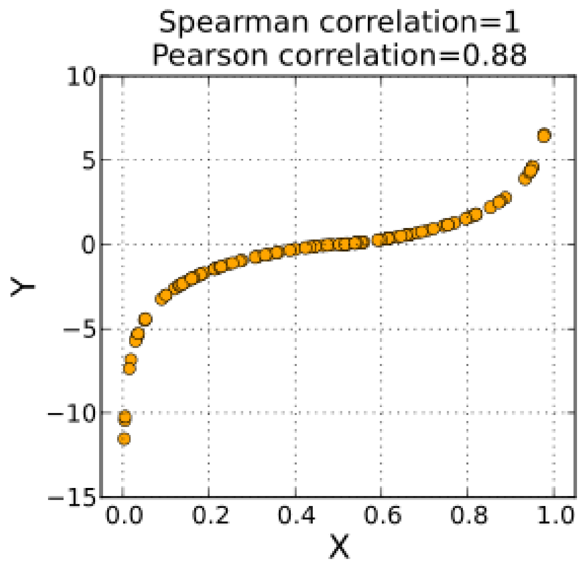


Image: https://en.wikipedia.org/wiki/Spearman%27s_rank_correlation_coefficient

Correlation and P-value

P-value can be obtained from correlation with

Null Hypothesis $H0 : r = 0$

Alternative Hypothesis $H1 : r \neq 0$

P-value tells us the probability of getting high correlation between x and y by pure chance

Examples

1. BMI vs Age, $r= 0.13$, p-value $=0.08$

2. BMI vs Age, $r= 0.13$, p-value $=0.04$

3. lung function before vs after exercise, $r= 0.93$, p-value $= 0.001$

Correlation

A strong correlation between x and y does **Not mean**

- x causes y : $X \rightarrow Y$
- y causes x : $Y \rightarrow X$
- x and y are caused by one or more other variables z: $Z \rightarrow X$, $Z \rightarrow Y$

Correlation is not causation

For single or more than 2 samples

One Sample t-Test

- Also known as single sample t-test
- When you have only one sample, to test if the mean of sample is equal to some well stablished/ hypothesized mean

- Null hypothesis H0 : $\boldsymbol{\mu_s = P}$
- Alternative hypothesis H1 : $\boldsymbol{\mu_s \neq P}$

ANOVA

- Analysis of Variance, generalised for more than 2 groups
- Tests weather mean of all the groups are similar or not

- Null hypothesis: There is no difference in means between groups
- Alternative hypothesis: At least one group mean is significantly different from the overall mean

Stats Demo links:
https://nikeshbajaj.github.io/teaching/demos/Stats/sampling
https://nikeshbajaj.github.io/teaching/demos/Stats/hypothesis

Resources/Video/Material:
https://nikeshbajaj.github.io/teaching/stats

If you have any question or doubt, please contact me via email

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