

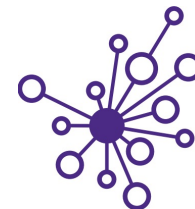


Knowledge and
solutions for a
changing world



Be boundless

Advancing data-
intensive discovery
in all fields



DESCRIPTIVE STATISTICS

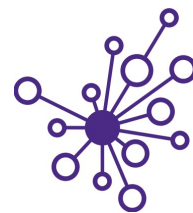
UW DIRECT

(Data Intensive Research Enabling Cutting-edge Tech)

<https://uwdirect.github.io>

Stéphanie Valteau

Chemical Engineering

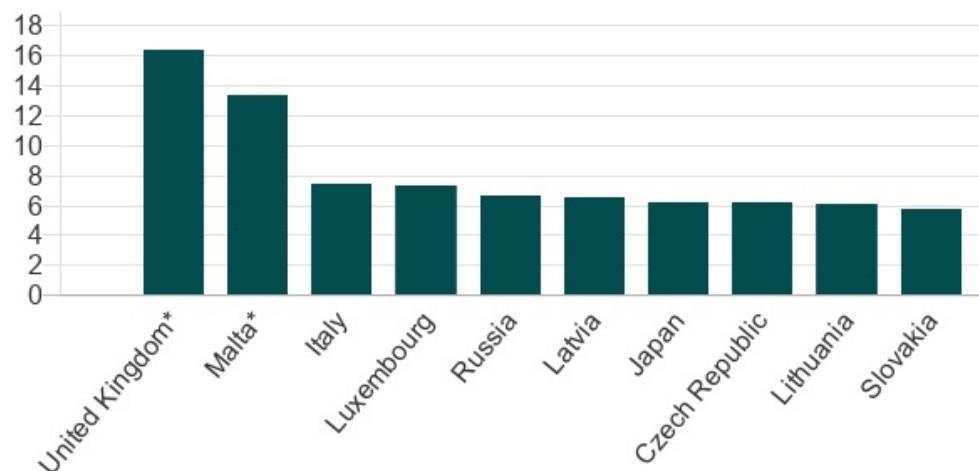


What is statistics?

The practice or science of **collecting** and **analyzing** numerical data in **large quantities**, especially for the purpose of **inferring proportions** in a whole from those in a **representative sample**.

Retail prices of roasted coffee

In USD (\$) per pound, 2016

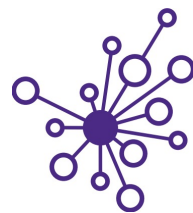


Source: International Coffee Organization. *Soluble coffee

BBC

W

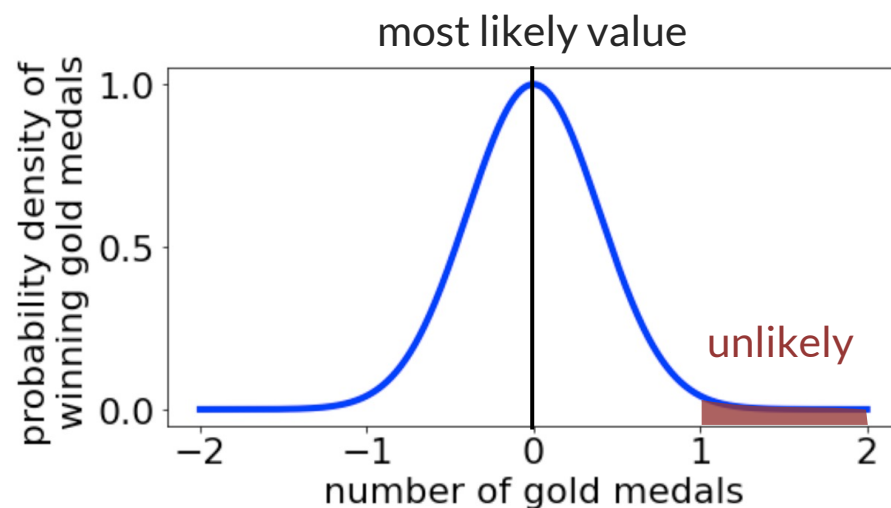
What can we learn from Stats?

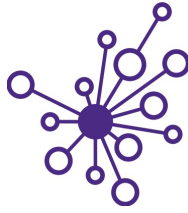


Statistics **does not tell us whether we are right** in coming to a conclusion (not absolute).



It tells us the **likelihood** of an outcome / the chances of being **wrong**





Two key concepts

Population

All possible values of an experimental variable (e.g. all stars in the milky way, all types of enzymes etc.)

Sample

A set of data drawn from a population

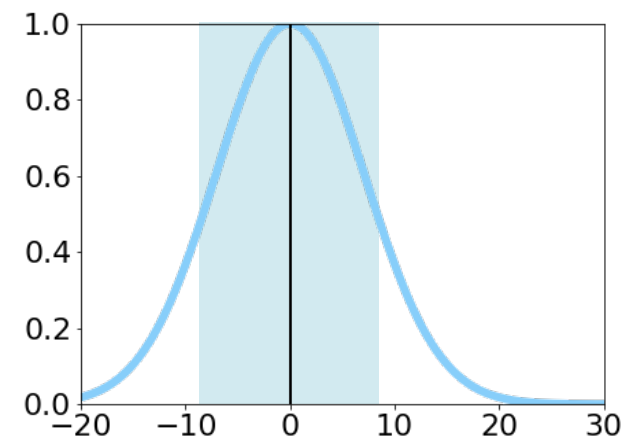
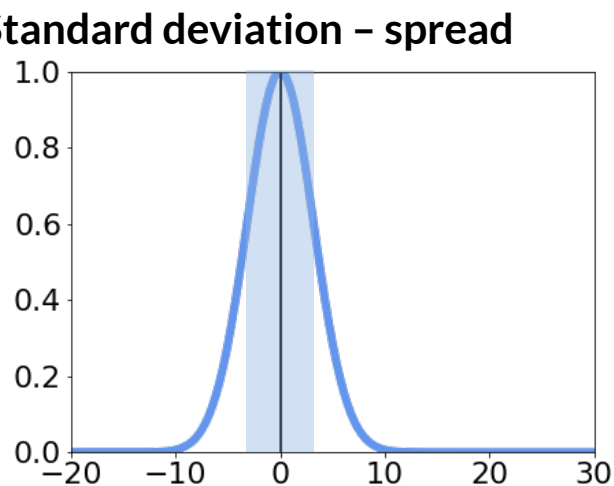
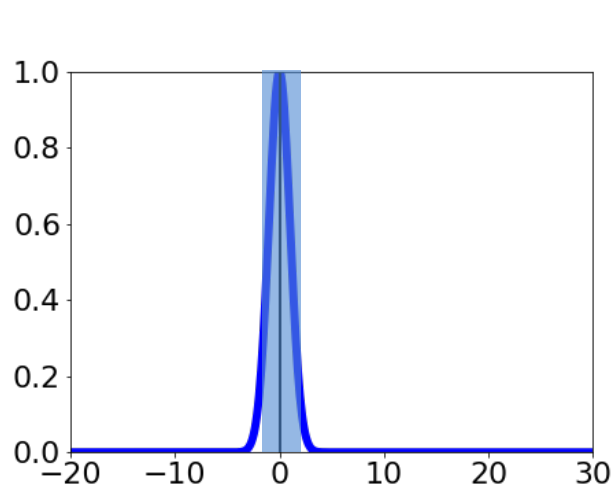
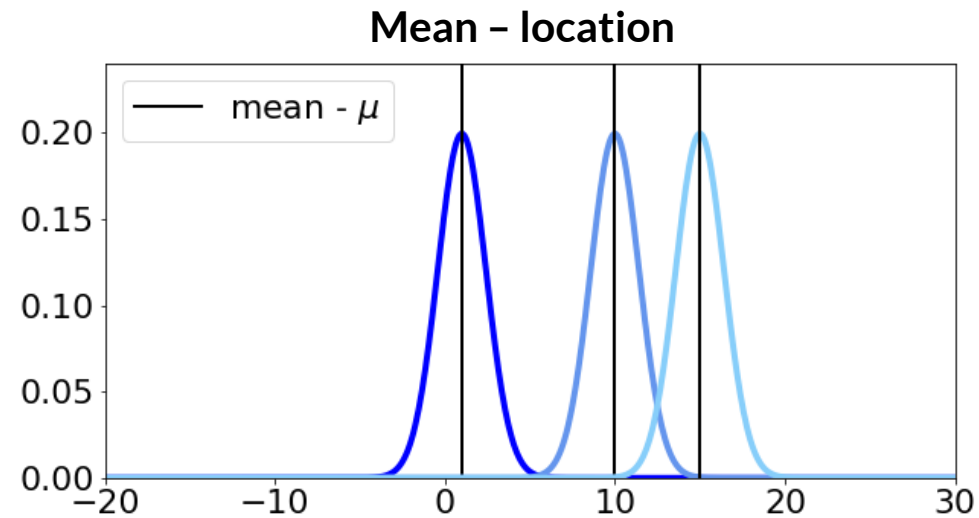
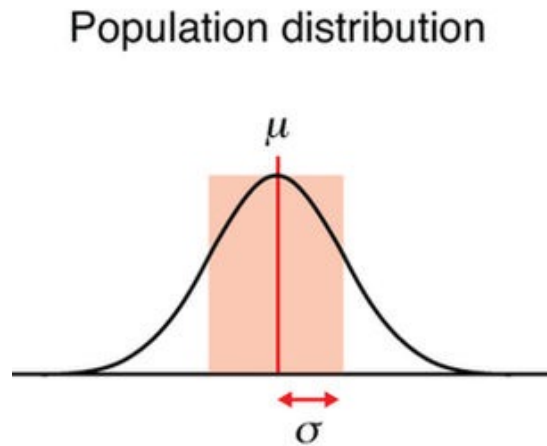
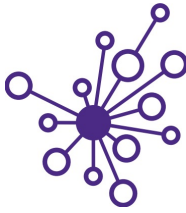
Often want to know the **mean** (μ) and **standard deviation** (σ) of a population

$$\mu = \sum_{i=1}^N \frac{X_i}{N}$$

$$\sigma = \sqrt{\frac{\sum_{i=1}^N (X_i - \mu)^2}{N}}$$

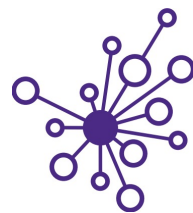


Mean & standard deviation





Standard dev. & Variance



Variance is square of std. dev.

$$\sigma = \sqrt{\frac{\sum_i^N (X_i - \mu)^2}{N}}$$

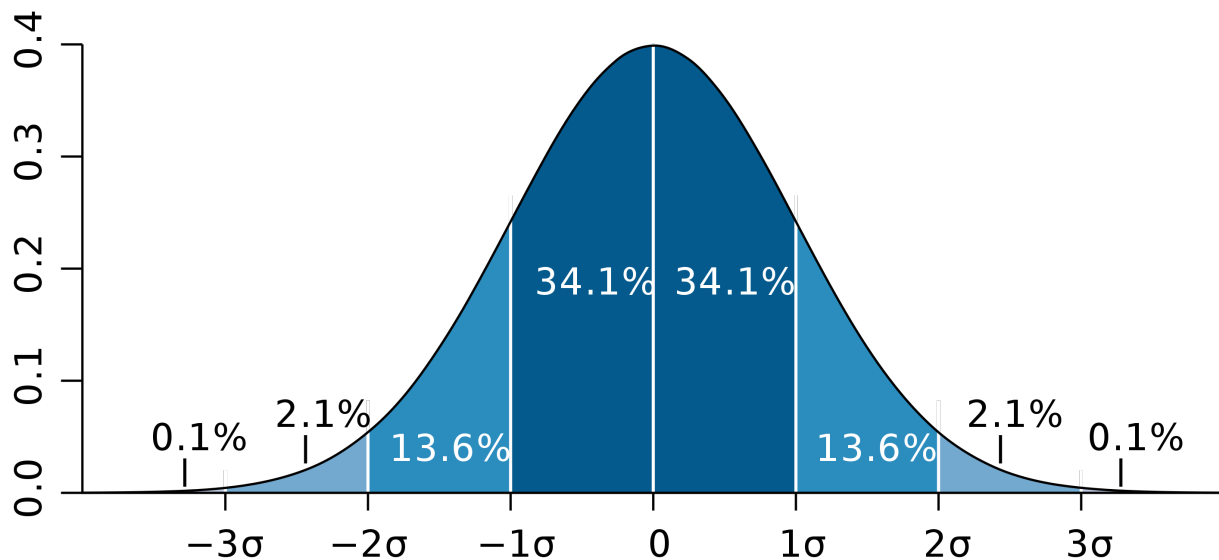
$$\sigma^2 = \frac{\sum_i^N (X_i - \mu)^2}{N}$$

$\pm 0.5 \sigma$ contains 39% of possible values

$\pm 1 \sigma$ contains 68%

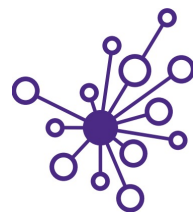
$\pm 2 \sigma$ contains 95%

$\pm 3 \sigma$ contains 99.7%



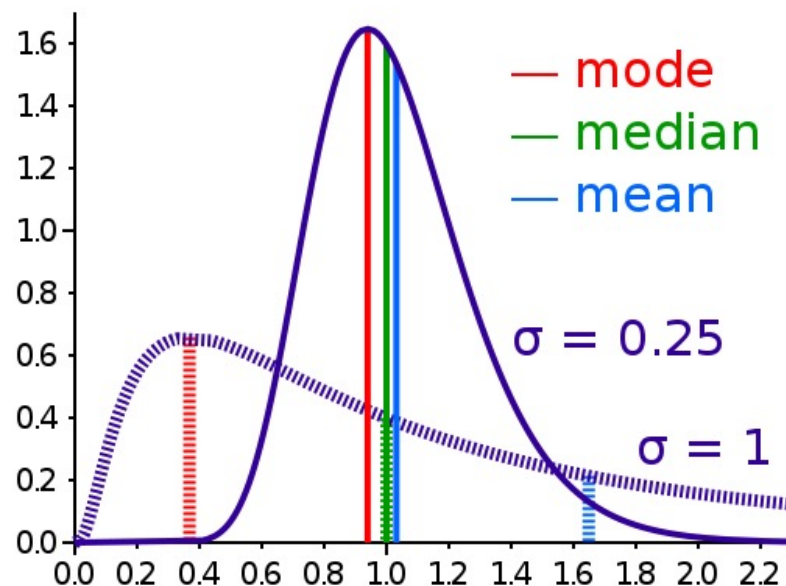


Median vs mean

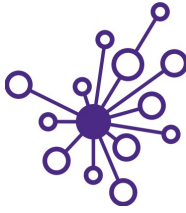


The **median** is a value separating the higher half from the lower half of a data sample, a population, or a probability distribution

Median is often **more robust** than the mean in the face of **skewed distributions** and outliers



Measuring μ, σ ?



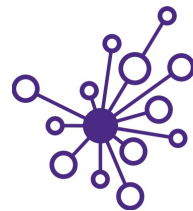
Is it possible to obtain the population for an experimental variable?

Is it possible to directly measure the mean (μ) and standard deviation (σ) of a population





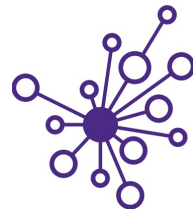
If we had a population



Open the `L3_Descriptive_Statistics.ipynb` notebook

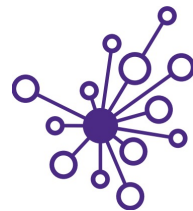


Back to reality



We could estimate the correct mean and standard deviation
from samples ...

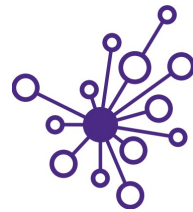
Population samples



Samples are **sets of data drawn from the population**

- Described by their size n (number of data points)
- Notation: X indexed by sample subscript, e.g. X_1

How to choose n ?



Sampling Bias

Do all values in a population have
the **same chance of being selected?**

If not, we have **bias**.

What is an example of bias in sampling?

Example: Assess average level of knowledge of US population based on survey responses from high school students

Population samples

Population: mean (μ) and s.d. (σ)

Sample: mean - \bar{X} and standard deviation - s

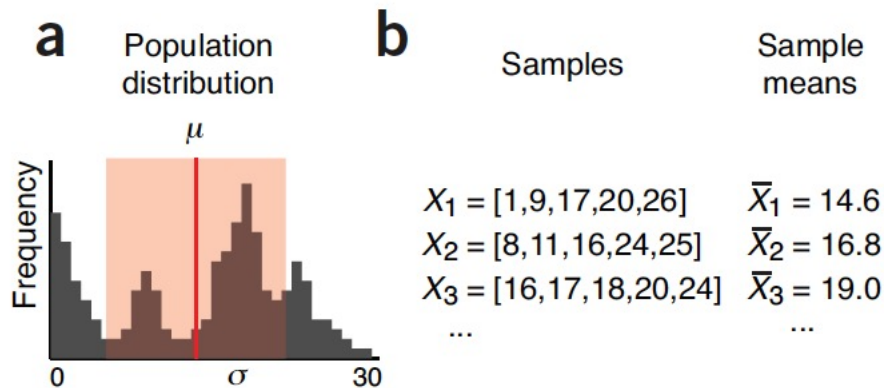
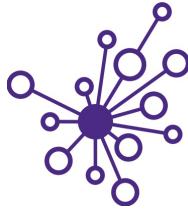


Figure 2 | Population parameters are estimated by sampling. (a) Frequency histogram of the values in a population. (b) Three representative samples taken from the population in (a), with their sample means. (c) Frequency histogram of means of all possible samples of size $n = 5$ taken from the population in (a).

Population samples



Sample parameters like \bar{X} have their own distributions – e.g. panel C

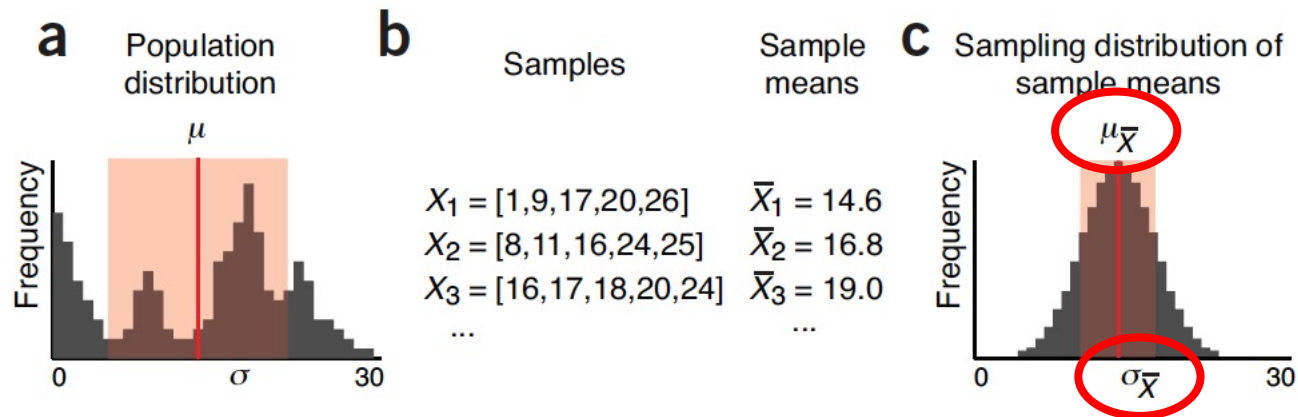
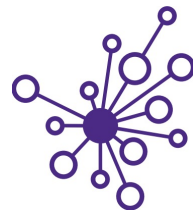


Figure 2 | Population parameters are estimated by sampling. (a) Frequency histogram of the values in a population. (b) Three representative samples taken from the population in a, with their sample means. (c) Frequency histogram of means of all possible samples of size $n = 5$ taken from the population in a.



Sampling **with** replacement



$N=7$ population values = [12, 13, 14, 15, 16, 17, 18]

Sample **with** replacement ($n=2$)

- First sample, each item has $1/7$ probability
- Second sample, each item has $1/7$ probability
- How many total possibilities (assuming order is important)?

$$N^n = 49$$

Each time we sample our choice is **independent** from the prior choice!



Sampling **without** replacement



Population values = [12, 13, 14, 15, 16, 17, 18]

Sample **without** replacement ($n=2$)

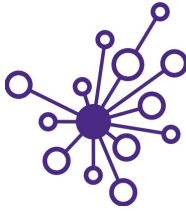
- First sample, each item has $1/7$ probability
- Second sample? $1/6$
- How many total possibilities (assuming order is important)?

$$\frac{N!}{(N - n)!} = \frac{7!}{(5 - 2)!} = \frac{7 \cdot 6 \cdot 5 \cdot \dots \cdot 1}{5 \cdot 4 \cdot 3 \cdot 2 \cdot 1} = 7 \cdot 6 = 42$$

Note: Each time we sample our choice **depends** on the previous choice!



Will sampling work?



Let's go back to the notebook

Sampling distributions

We observed that $\sigma_{\bar{X}} < \sigma$
and also that $\mu \cong \mu_{\bar{X}}$

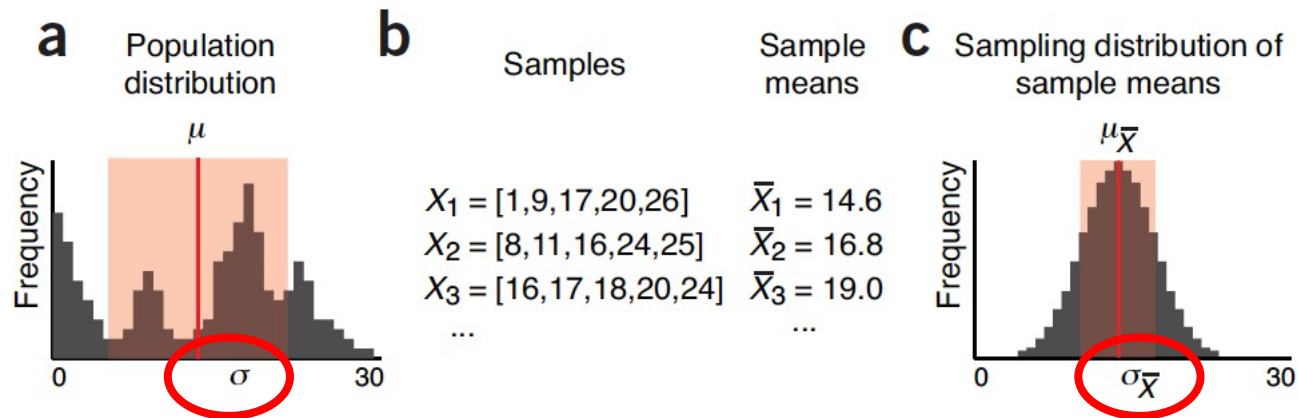
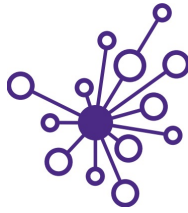
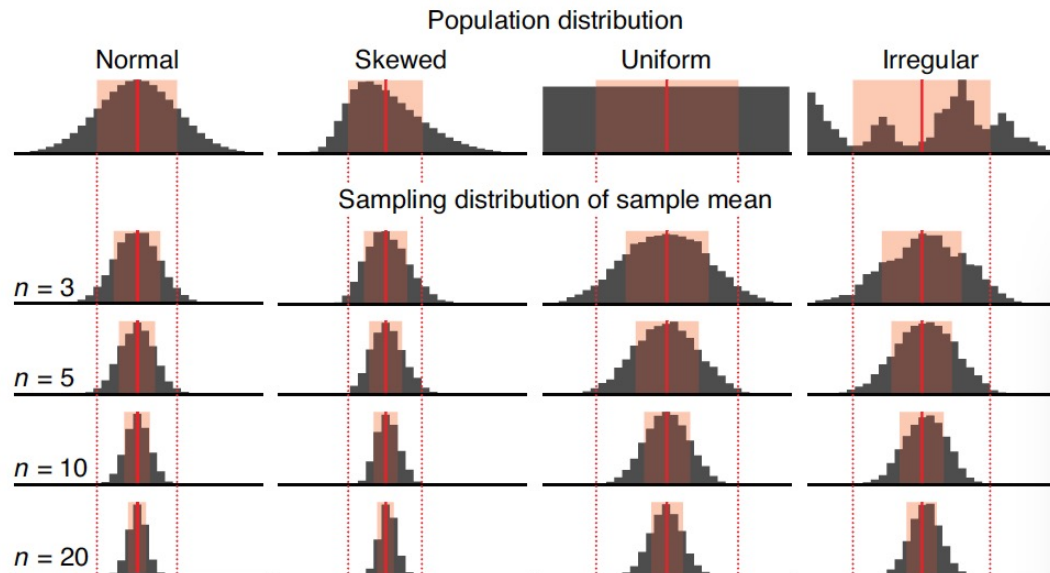


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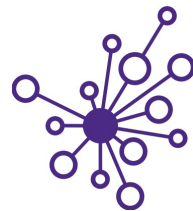
The Central limit theorem



As n increases, the distribution of sample means tends to become a **normal distribution**, regardless of population distribution shape



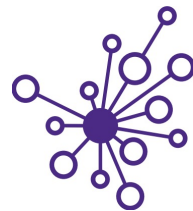
Central limit theorem



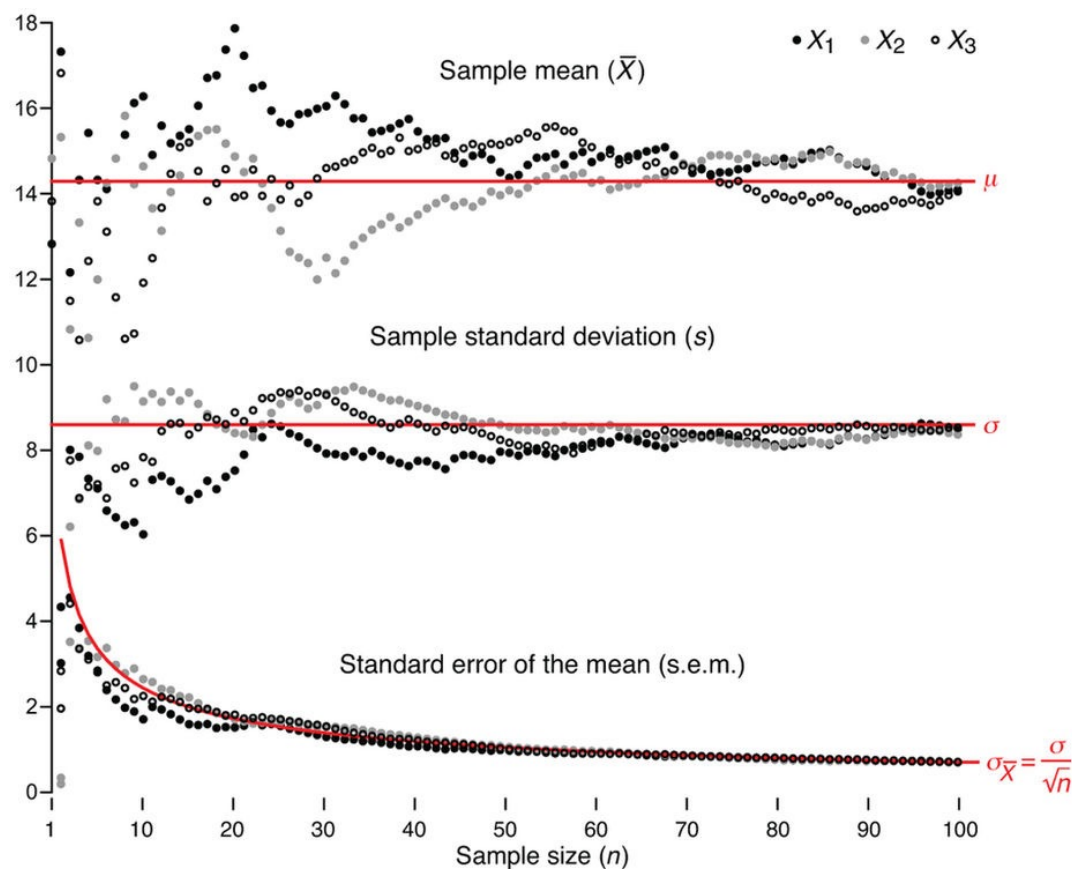
- As **n** increases, $\mu_{\bar{X}}$ decreases, i.e. we get **better and better estimates** of the population mean μ
- Thus big **n** makes $\mu \cong \mu_{\bar{X}}$
- As **n** increases, $\sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$

	$n = 50$	$n = 2000$	exact
mean	415.30	415.22	415.23

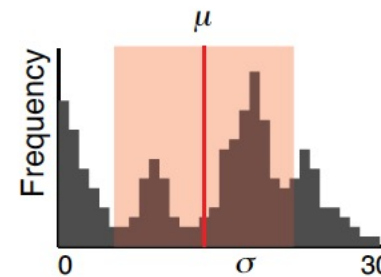
Central limit theorem



Increasing sample size improves estimates



a Population distribution





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That's all for today!

