

Dealing with Numbers

“Surviving the post p -values era”

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- I. Errors & Uncertainty**
- II. Statistical measures to characterize uncertainty**
- III. Statistical sampling**
- IV. Sample size**

Surviving the post p-values era:

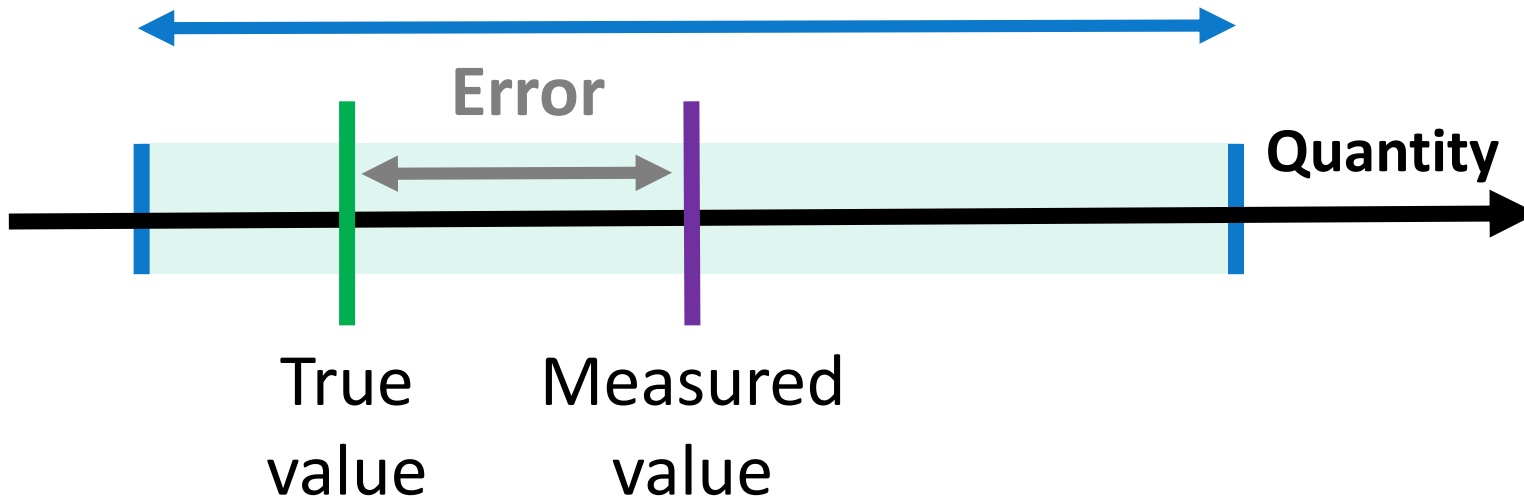
- Use of the p-value as a size-dependent function: model and applications**

Measurement is a process of experimentally obtaining the value of a quantity.

Optimal measurement procedure to be as close as possible to the **true value**

Uncertainty range

Measured value $\pm U$



<https://sisu.ut.ee/measurement/uncertainty>

ERROR

most times cannot be used in practice

UNCERTAINTY

Interval around the measured value. It provides the quality of a measurement

Resolution

Estimation method

Calibration

Repeatability uncertainty
(human factor)

Reproducibility

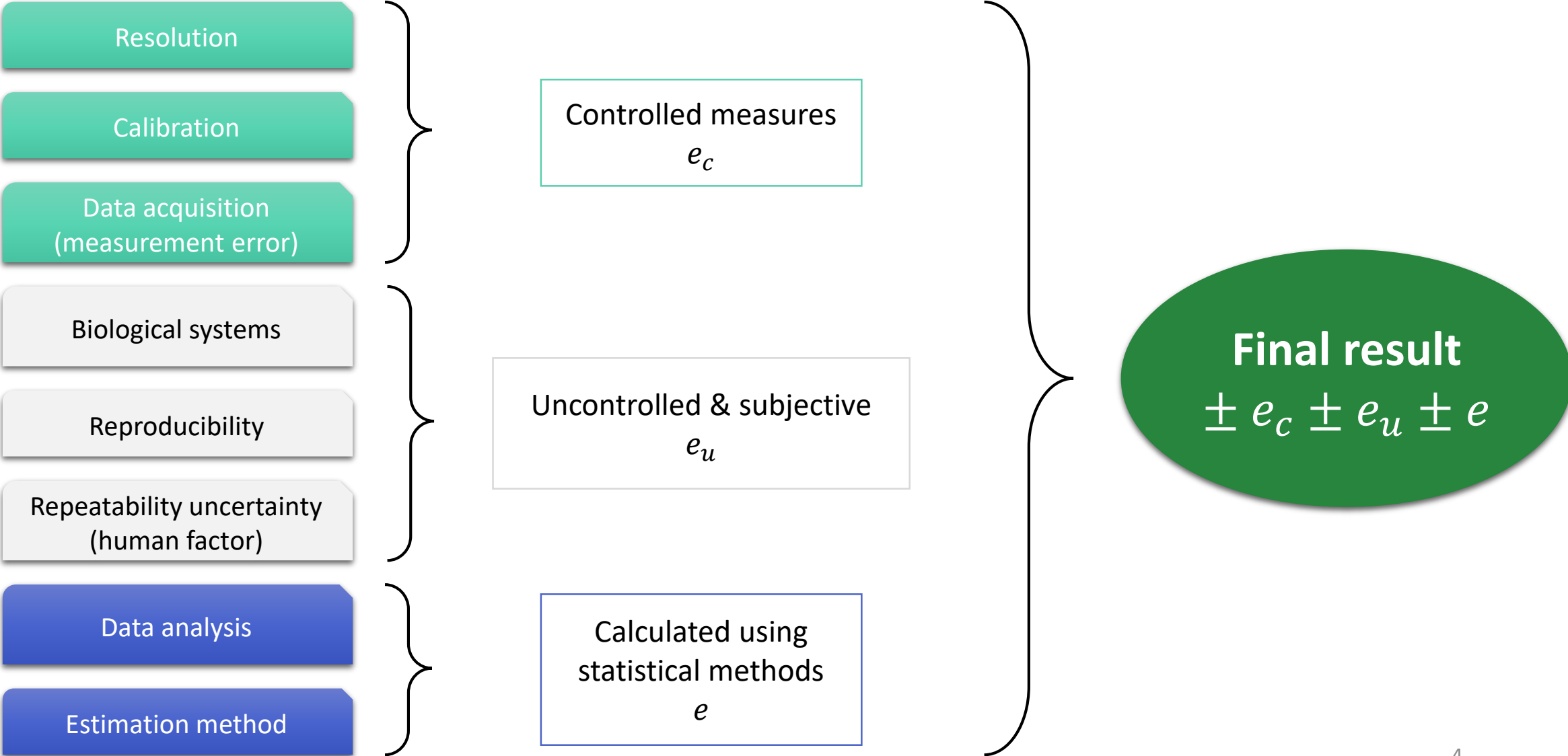
Data acquisition
(measurement error)

Biological systems

Data analysis

Final result report

Sources of variance / errors / uncertainty / bias / noise



Statistical measures to characterize uncertainty

Standard deviation (σ)

- Variability or dispersion of the data w.r.t. the mean.


- $s = \sqrt{\frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n}}$; (s^2 : sample variance, biased)

- $S = \sqrt{\frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n-1}}$; (S^2 : quasi-variance non-biased)

Standard error of the mean (SEM)

- Standard deviation of the mean: how far the sample mean of the data is likely to be from the true population mean
- It can be used to validate the accuracy of a sample.
- $s.e.m_{\bar{x}} = \frac{s}{\sqrt{n}}$: standard error

Confidence Intervals (CI at 95%)

- “A range of values it can be 95% confident that it contains the true mean value” 
- If the variance of the population (σ^2) is known:

$$CI = \bar{x} \pm z_{(1-\alpha/2)} \frac{\sigma}{\sqrt{n}}$$

- If the σ is unknown but n is large:

$$CI = \bar{x} \pm z_{(1-\alpha/2)} \frac{s}{\sqrt{n}}$$

- If the σ is unknown but n is small:


$$CI = \bar{x} \pm t_{(n-1, 1-\alpha/2)} \frac{s}{\sqrt{n}}$$

Statistical measures to characterize uncertainty

Standard deviation (σ)

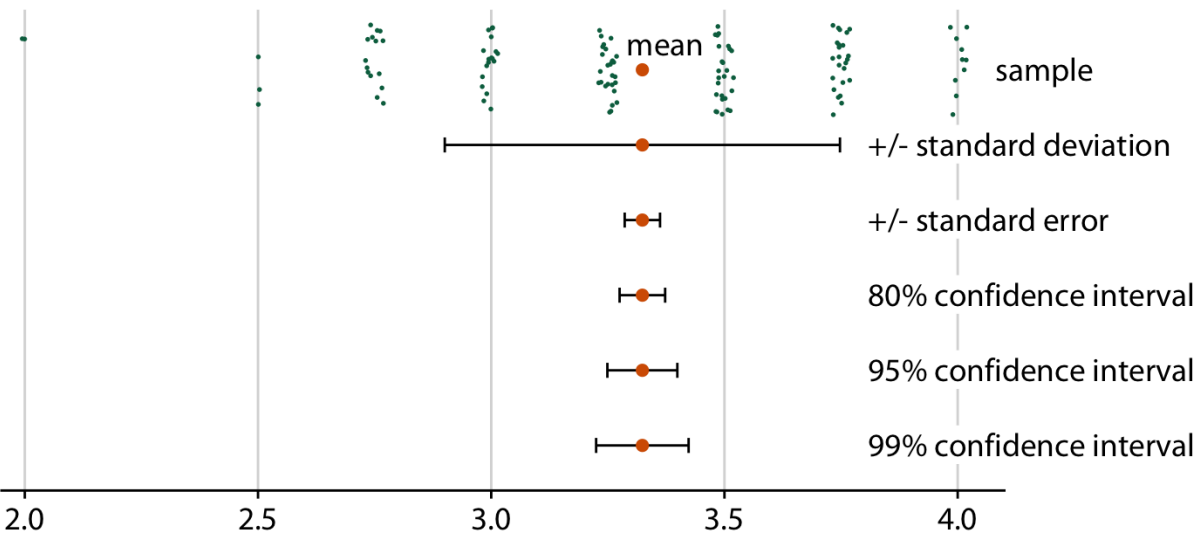
- Variability or dispersion of the data w.r.t. the mean.

Confidence Intervals (CI at 95%)

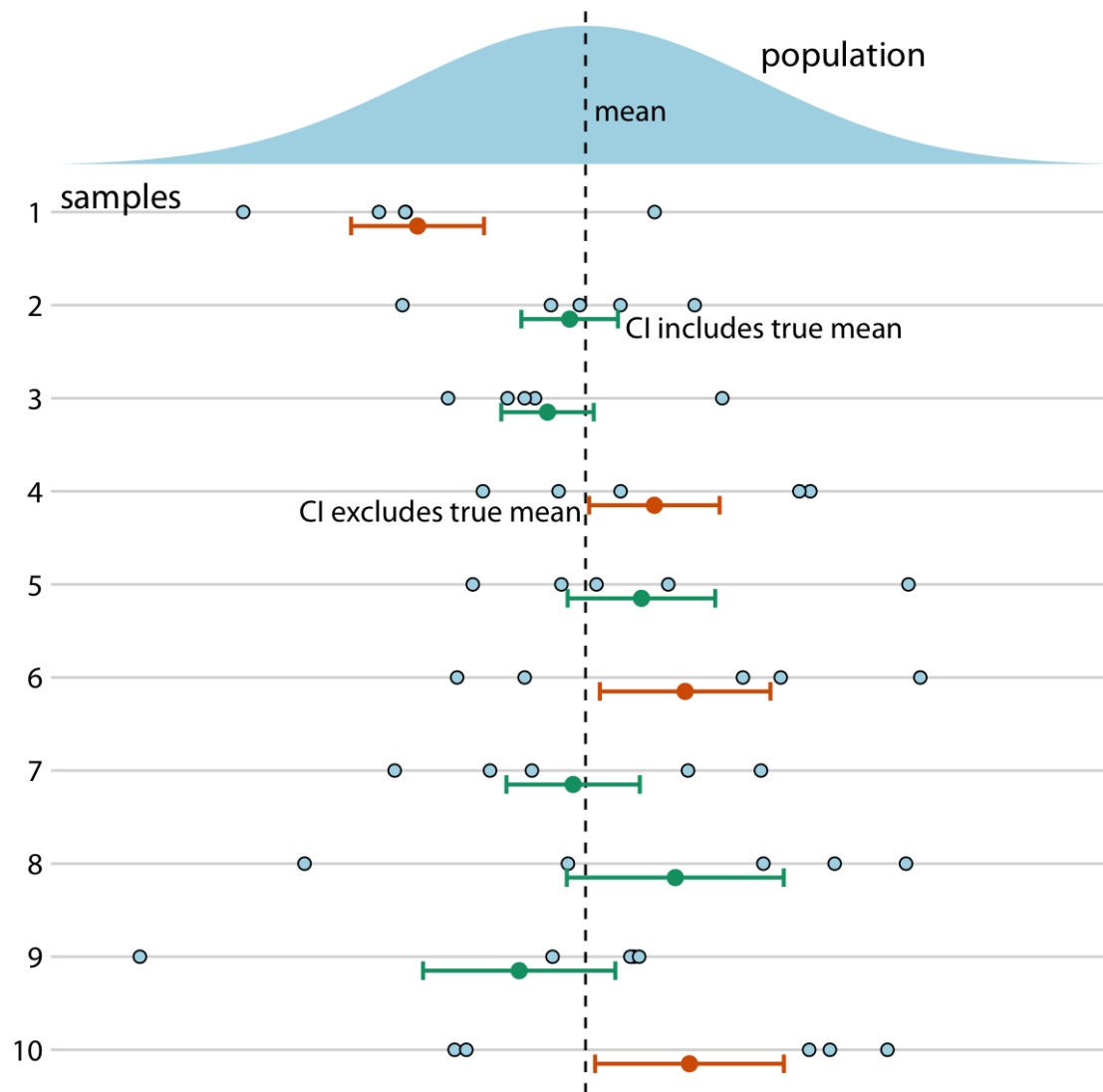
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Standard error of the mean (SEM)

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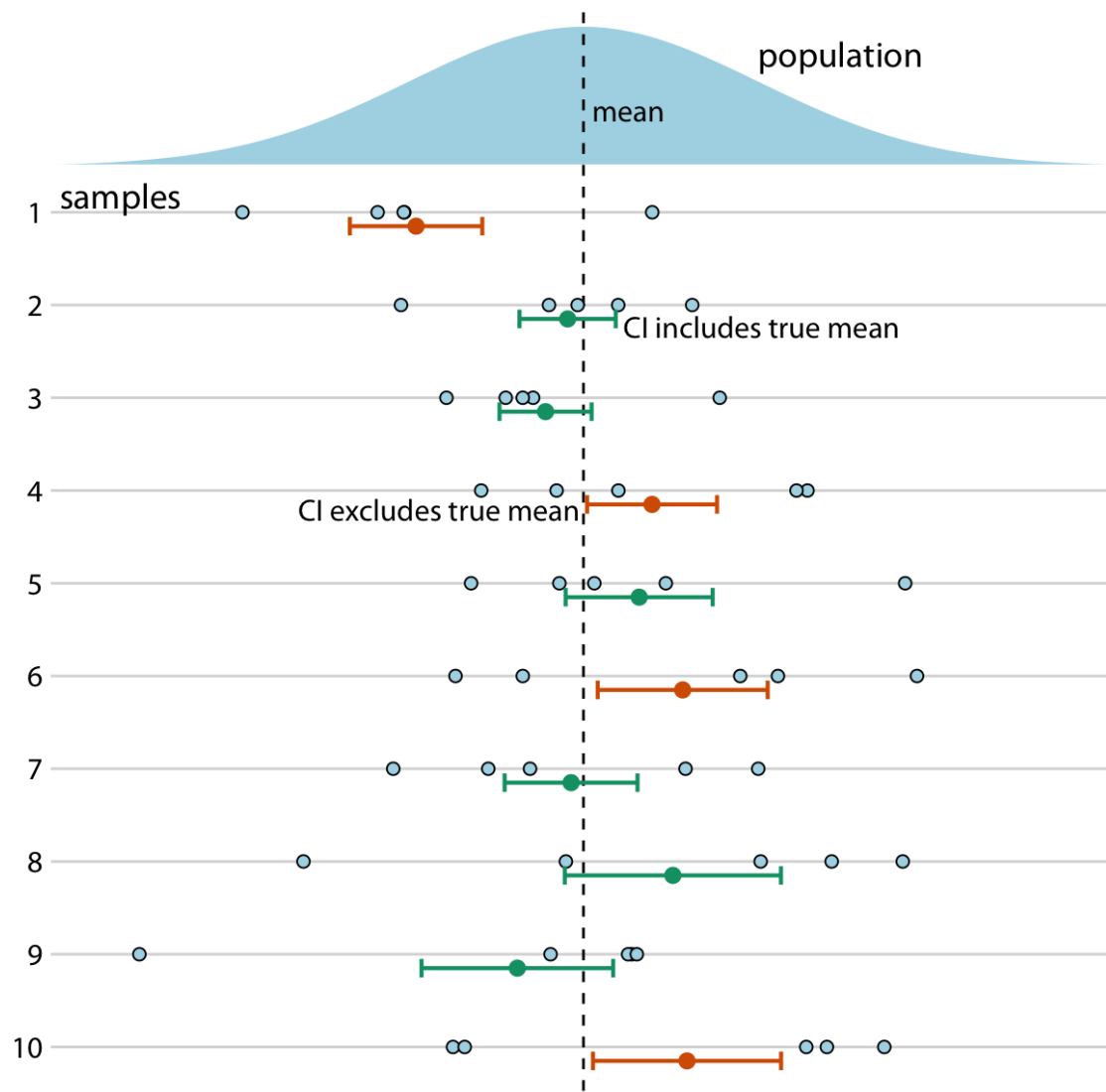
Previous measures come from experimental data → BIAS



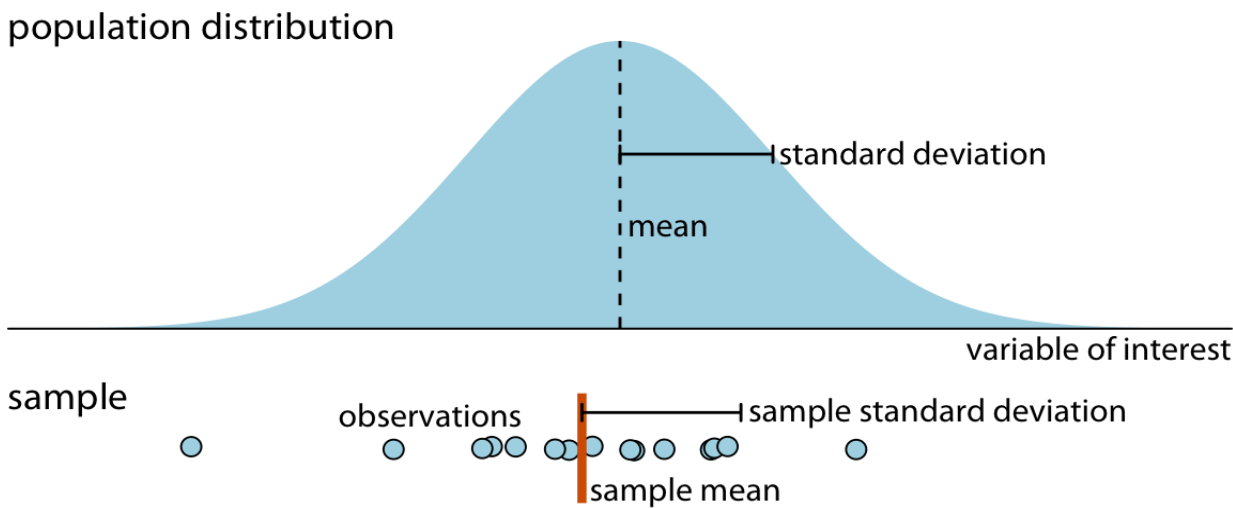
True meaning of a Confidence Interval

For each 100 intervals, we believe that at least in 95 of them we will find the expected value.

Previous measures come from experimental data → BIAS

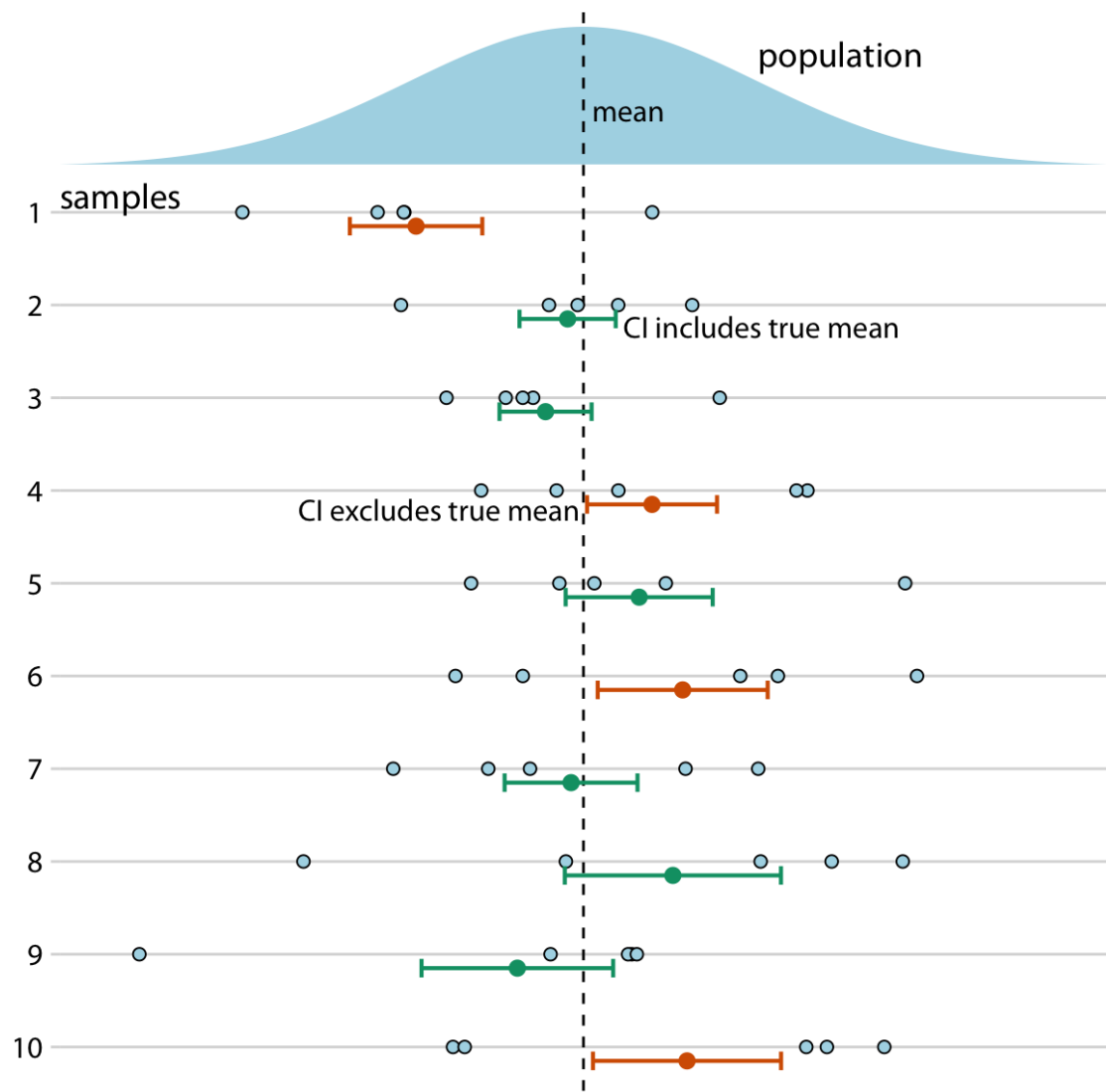


Statistical sampling



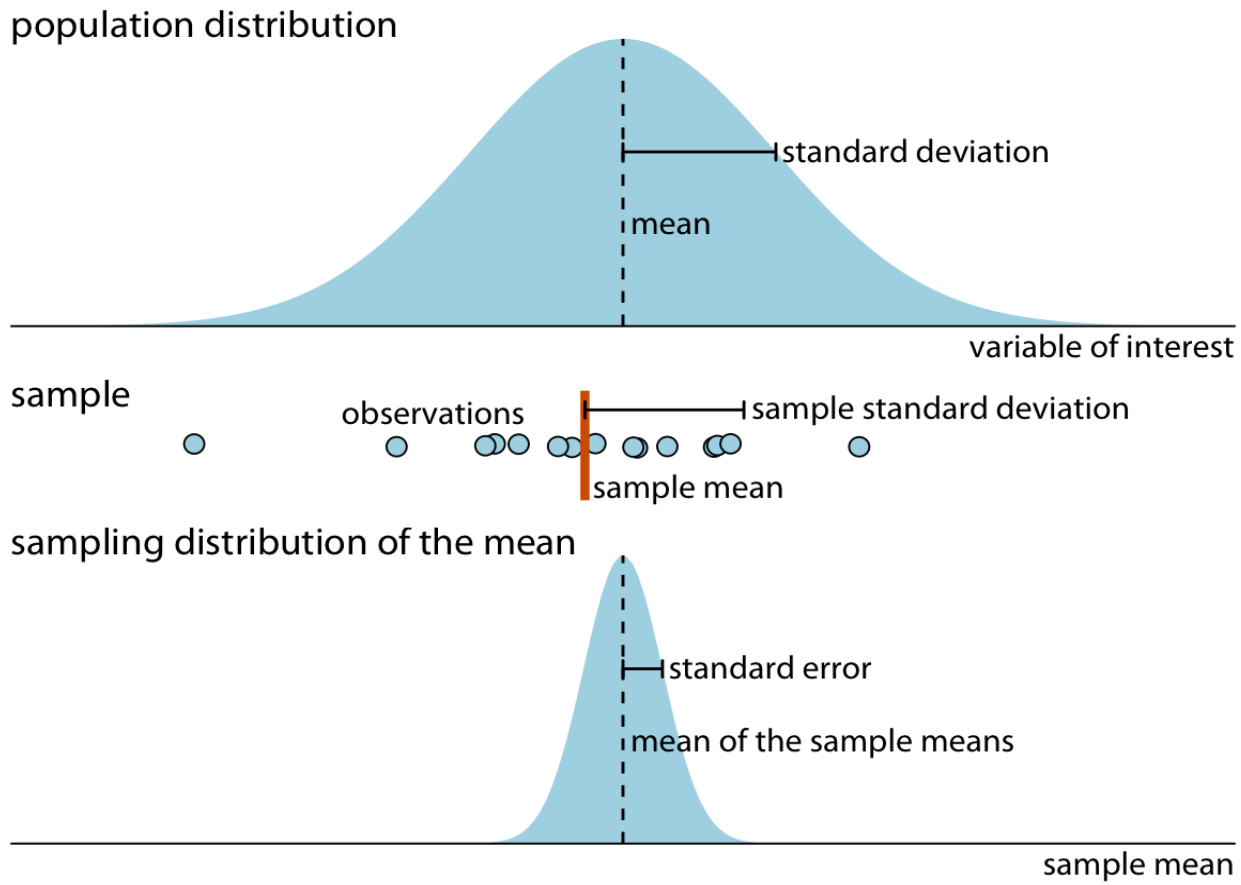
- 1. Any finite sample of that variable will have a different sample mean and standard deviation.
- 2. Repeat the sampling and calculate a mean each time.

Previous measures come from experimental data → BIAS



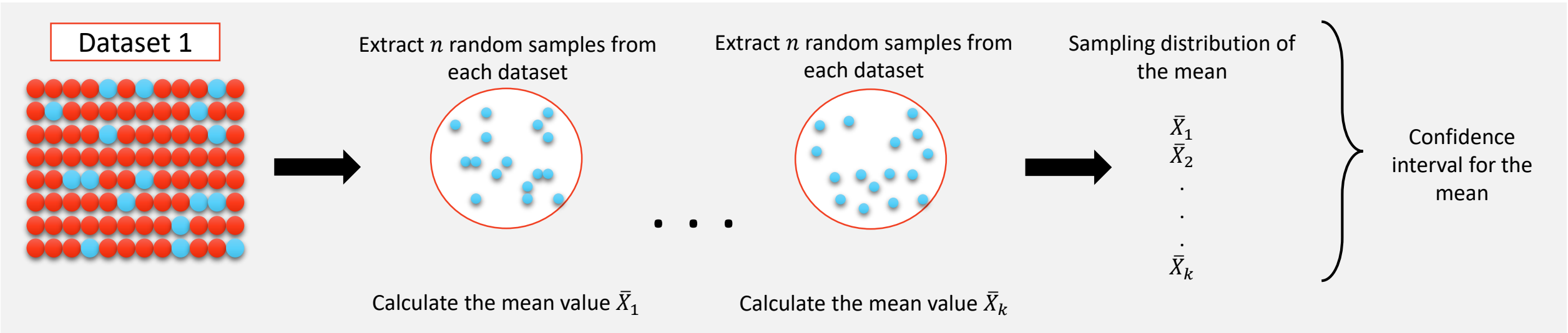
<https://serialmentor.com/dataviz/visualizing-uncertainty.html>

Statistical sampling

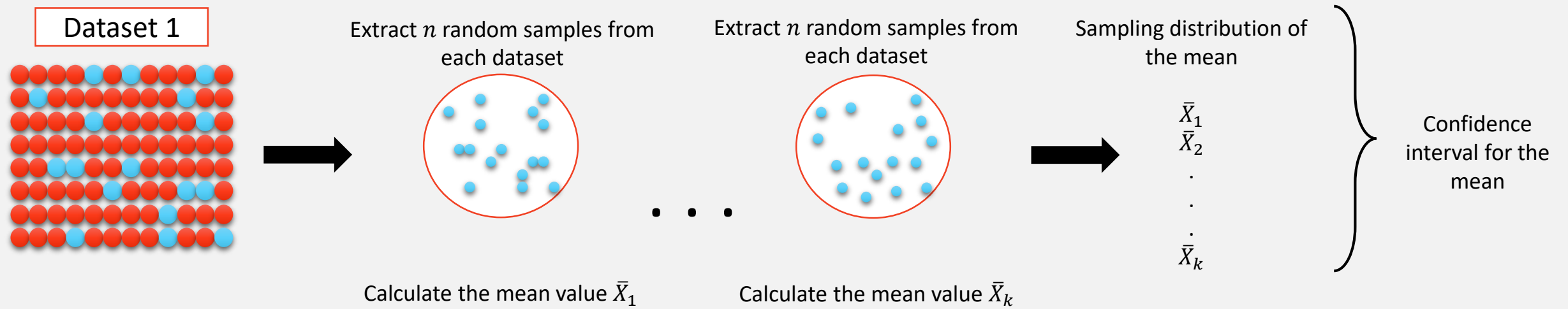


The resulting means would be distributed according to the sampling distribution of the mean.

Computation of (confidence) intervals of the sample mean



Computation of (confidence) intervals of the sample mean



The following conditions must be satisfied:

I. All the samples must be **randomly chosen**

II. Confidence intervals are designed for **normal distributions**.

➤ The mean follows a normal conditions (central limit theorem).

III. Sampling has to satisfy the **independence** condition.

- Sampling with replacement.
- Case without replacement: Questionnaires to people who is leaving a store, you cannot ask them to go back again.
 - Recommendation: n value smaller than 10% of the sample → Calculate first n .

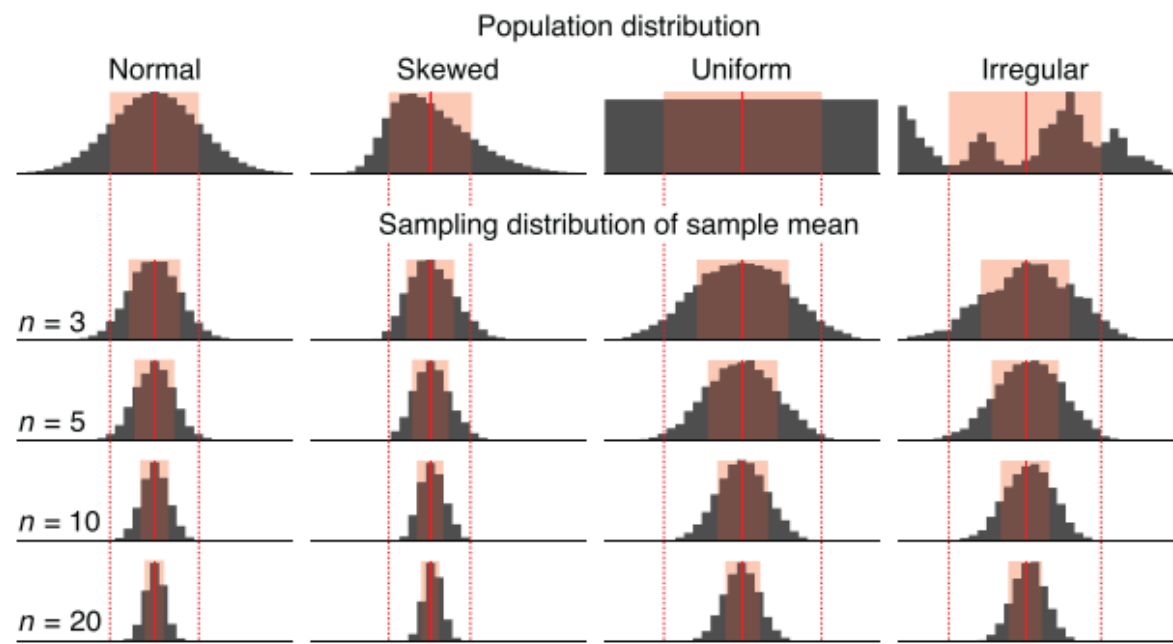
How robust are these measures?

What is the main problem with estimation
using experimental data?

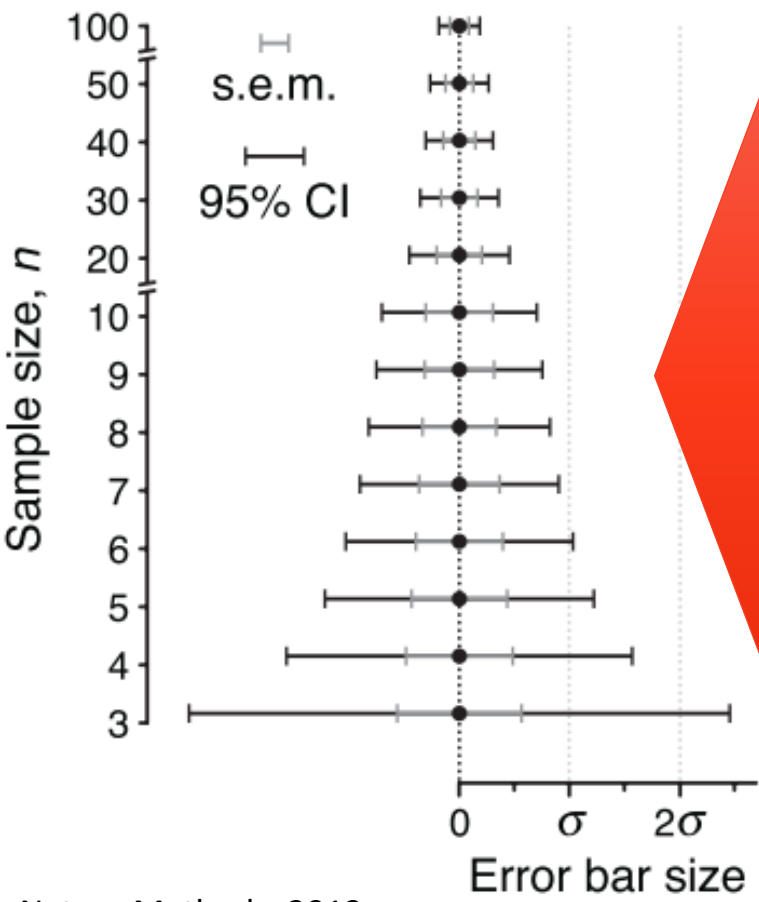
Previous measures come from experimental data → BIAS

- Pooled mean (mean of the sample means): $\mu_X = \frac{\sum_i N_{X_i} \mu_{X_i}}{\sum_i N_{X_i}}$
- Pooled standard deviation: $\sigma_X = \frac{\sum_i (N_{X_i} - 1) \sigma_{X_i}^2}{\sum_i (N_{X_i} - 1)}$

Sampling distributions of sample means for 10,000 samples for indicated sample sizes



Relationship between s.e.m. and CI at 95% error bars when increasing the sample size (n)



Intervals of the error bars decrease when n increases

Sample size (n)

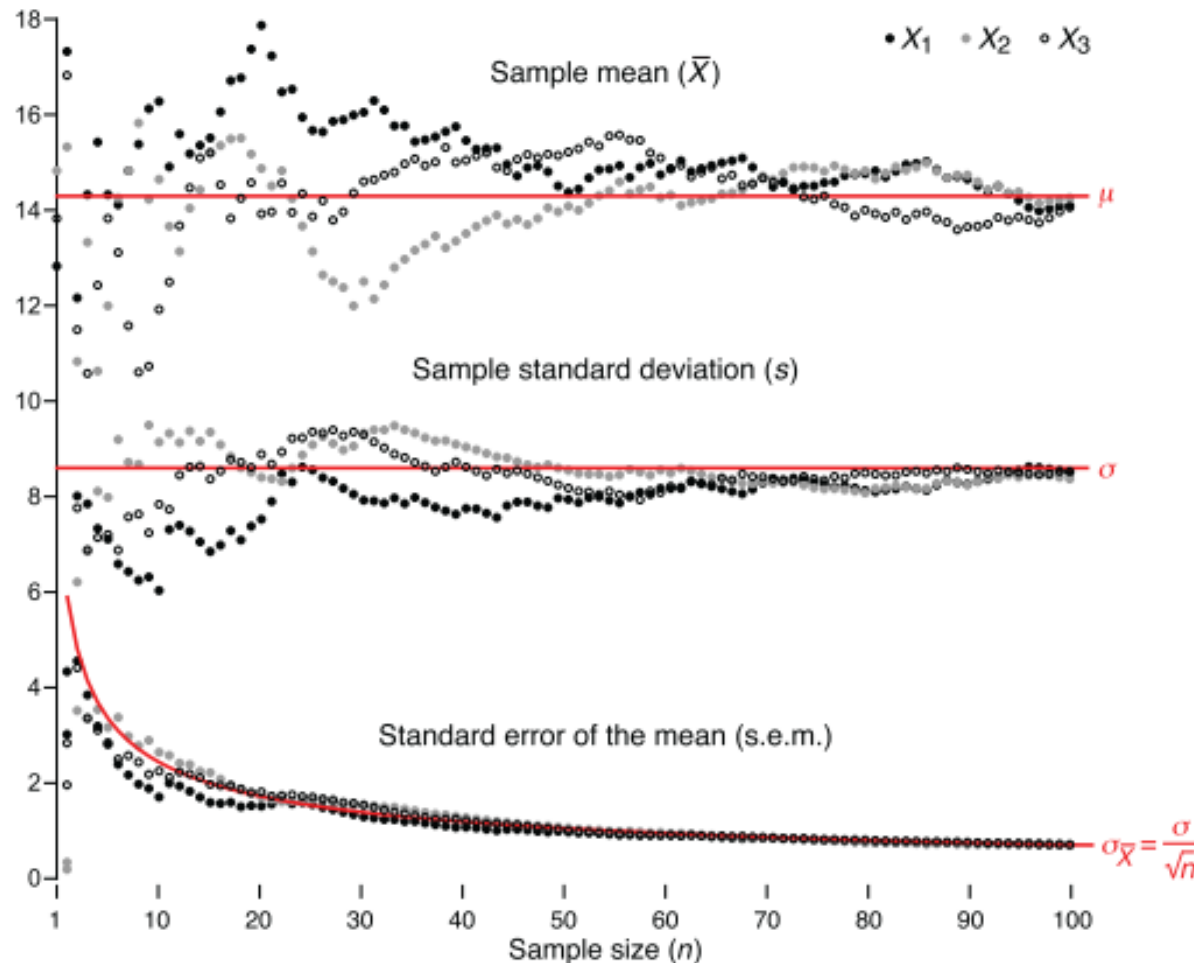


Figure 4 | The mean (\bar{X}), s.d. (s) and s.e.m. of three samples of increasing size drawn from the distribution in **Figure 2a**. As n is increased, \bar{X} and s more closely approximate μ and σ . The s.e.m. (s/\sqrt{n}) is an estimate of $\sigma_{\bar{X}}$ and measures how well the sample mean approximates the population mean.

There is a certain value of n for which
the population mean is well
approximated.

Sample size (n)

To calculate n it is necessary to set parameters that depend on the type of study:

- Standard deviation of the sample (σ)
- Confidence level ($1-\alpha$)
- Error range (e) / Length of the confidence interval (L)

If we are interested in the confidence interval of the sample mean (μ):

- If the standard deviation of the population (σ) is known:
$$n = \frac{(2z_{(1-\alpha/2)}\sigma)^2}{L^2}.$$
- If σ is unknown but n is large:
$$n = \frac{(2z_{(1-\alpha/2)}S)^2}{L^2}.$$
- If σ is unknown but n is small:
$$n = \frac{(2t_{(n-1,1-\alpha/2)}S)^2}{L^2},$$

where $z_{(1-\alpha/2)}$ and $t_{(n-1,1-\alpha/2)}$ are the corresponding quantiles in the Normal and t-Student distributions respectively.

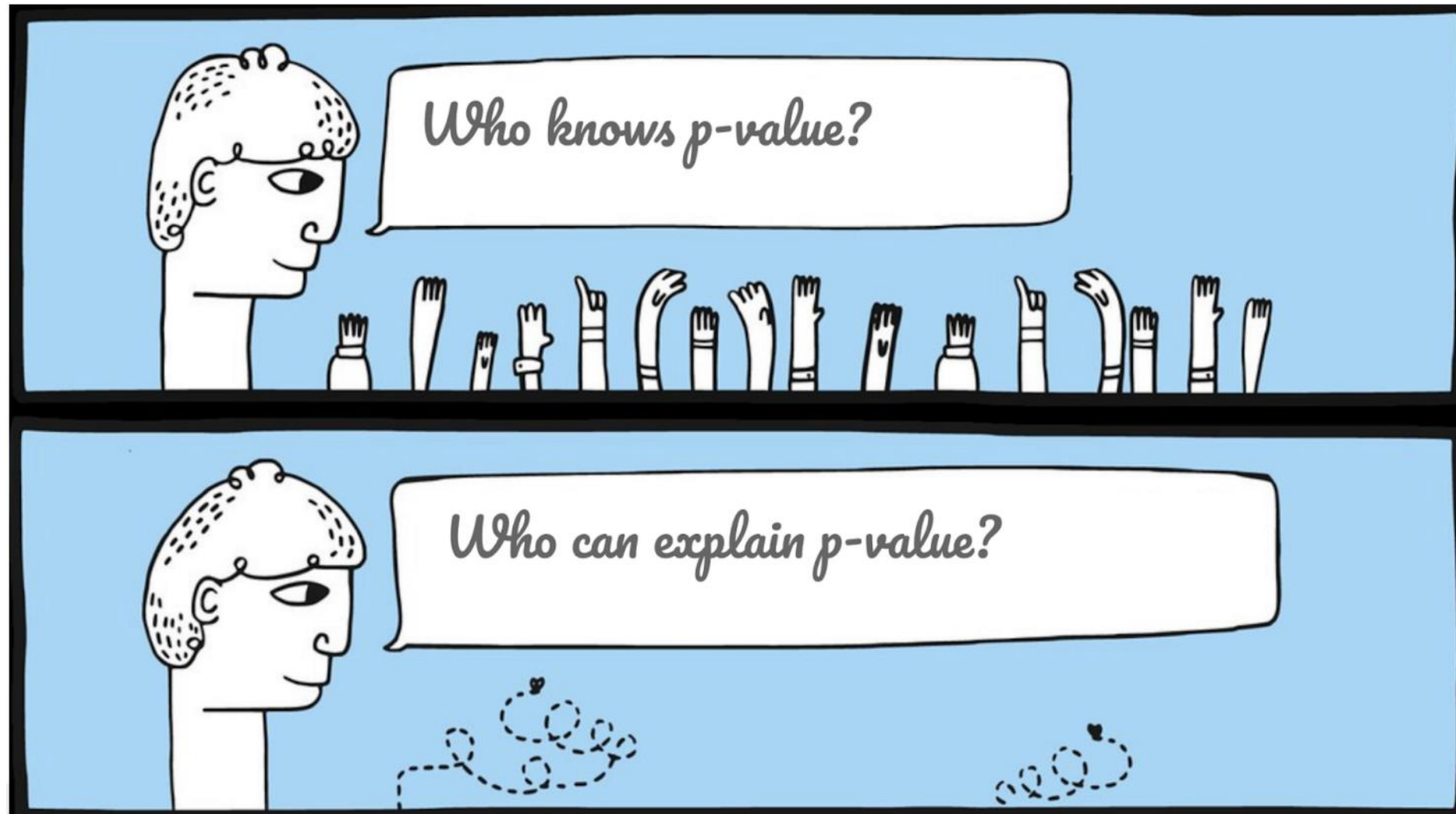
Sample size (n)

Fórmula		R
One parameter Estimation	Mean	$n = \left(\frac{2 \cdot z_{\alpha/2} \cdot \sigma}{A} \right)^2$ <i>sample.size.mean</i>
	Probabilities	$n = \left(\frac{z_{\alpha/2}}{A} \right)^2$ <i>sample.size.prop</i>
Means comparison	Independent	$n = \frac{2 \cdot \sigma^2 \cdot (z_{\alpha/2} + z_{\beta})^2}{\Delta^2}$ <i>TwoSampleMean.Equality</i>
	Paired	$N = \frac{2 \cdot \sigma_l^2 \cdot (z_{\alpha/2} + z_{\beta})^2}{\Delta^2}$ -
	Análisis del cambio	$n = \frac{2 \cdot \sigma_c^2 \cdot (z_{\alpha/2} + z_{\beta})^2}{\Delta^2}$ -
	Equivalent	$n = \frac{2\sigma^2(z_{\alpha} + z_{\beta/2})^2}{\Delta^2}$ <i>TwoSampleMean.Equivalence</i>
	Non-inferiority	$n = \frac{2\sigma^2(z_{\alpha} + z_{\beta})^2}{\Delta^2}$ <i>TwoSampleMean.NIS</i>
	Precision	$n = \frac{8 \cdot \sigma^2 \cdot z_{\alpha/2}^2}{A^2}$ -
Probabilities comparison	Independent	$n = \left(\frac{z_{\alpha/2} \sqrt{2p(1-p)}}{p_A - p_B} + \frac{z_{\beta} \sqrt{p_A(1-p_A) + p_B(1-p_B)}}{p_A - p_B} \right)^2$ <i>TwoSampleProportion.Equality</i>
Times comparison	Instant recruitment	$N = \frac{2E}{2 - \pi_A - \pi_B}$ <i>ssizeCT.default</i> <i>(para HRR constantes)</i>
	Recruitment for a period	$N = \frac{2(z_{1-\alpha/2} + z_{1-\beta})^2 (\Phi(\lambda_A) + \Phi(\lambda_B))}{(\lambda_A - \lambda_B)^2}$ <i>TwoSampleSurvival.Equality</i>

Nomenclature

- σ^2 : total variance
- σ_l^2 : intra-subject variance
- σ_c^2 : variance of the changing variable
- Δ : difference we want to detect
- ρ : correlation between both observations
- A : amplitude of the interval
- E : number of events we need to observe
- λ : event appearance rate
- π : estimated proportion of cases for which the event will NOT be present during the study.

Surviving the post *p-values* era



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