

# Sample Size calculations

Curs d'Estadística Bàsica per a la Recerca Biomèdica

UEB - VHIR

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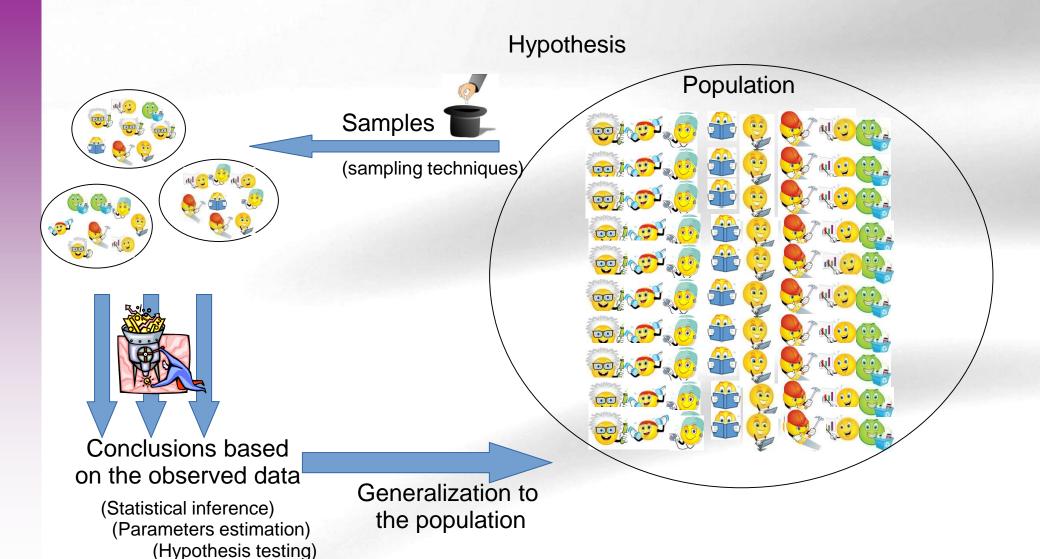








### The objective of statistical inference





# Sample Size in Statistical Studies

- Statistical inference is used to generalize,
  - It helps obtain conclusions from samples,
  - and apply them to populations,
  - with a certain degree of (known) precision.
- This can be made only if
  - Some assumptions hold (e.g. Normality)
  - The sample size is big enough as to warrant the desired precision.







- Before discussing sample size calculations there are several things to keep in mind
  - Type of calculations depend on study goal.
    - Estimation
    - Testing
  - Preliminary concepts to be used
    - Standard error of an estimator
    - Confidence interval
    - Type I and type II errors. Power of a test





### Standard error of the mean

- A measure of how variable is the sample mean when computed in distinct samples.
  - Standard deviation of the distribution of sample means
- Usually it is defined as population standard deviation divided by squared root of sample size
  - It is estimated substituting population by sample deviation

standard error = 
$$\frac{\sigma}{\sqrt{n}} \cong \frac{s}{\sqrt{n}}$$





# Standard error of a proportion

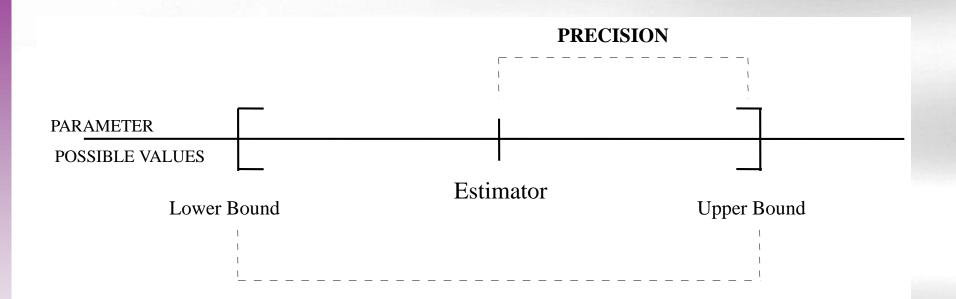
- The standard error of a proportion is computed similarly to the SEM.
  - Instead of the standard deviation it uses the population proportion in the formula.
  - Because p is usual unknown it is substituted by its estimator.
  - It is common to put q=1-p

standard error = 
$$\sqrt{\frac{p \cdot (1-p)}{n}} \cong \sqrt{\frac{\hat{p} \cdot (1-\hat{p})}{n}}$$





# Confidence interval



Values in which we are confident that real population parameter is inside With a prefixed confidence level (Usually 95%)



# Formulas for confidence intervals

- Data normally distributed
  - Population variance known (unrealistic assumption)
  - Population variance unkown, estimated by sample variance

- Data: Counts of presence or absence of an event
  - Sample must be "big enough"
- $z_{\epsilon/2}$  are quantiles of standard Normal N(0,1) distribution

$$|\overline{X}_n - z_{\varepsilon/2} \frac{\sigma}{\sqrt{n}} \le \mu \le \overline{X}_n + z_{\varepsilon/2} \frac{\sigma}{\sqrt{n}}|$$

$$\overline{X}_n - t_{\varepsilon/2} \frac{s}{\sqrt{n}} \le \mu \le \overline{X}_n + t_{\varepsilon/2} \frac{s}{\sqrt{n}}$$

$$\hat{p} \pm z_{\varepsilon/2} \sqrt{\frac{\hat{p}\hat{q}}{n}}; \quad n \ge 30, n\hat{p} \ge 5, n\hat{q} \ge 5$$

1- ε	0,90	0,95	0,99
$Z_{arepsilon/2}$	1,64	1,96	2,58





### Example: Confidence interval for the mean

- Goal: Estimate ureic nitrogen concentracion in serum (SUN) in rats that have been eating a certain diet.
- A sample of size 10 has been taken.
- Confidence interval is computed from formula (2) above





# Confidence interval for a proportion

#### **Problem**

- A molecular diagnosis lab is doing tests to detect hereditary venous pathology (PVH).
- In a series of 150 affected patients 18 show in their genètic profile the AGx allele for the gene related with the disease.
- With a confidence of 99% which is the estimation for the percentage of AGx individuals between people affected by PVH?

#### **Solution**

Relative frequency in the sample:

$$\hat{p} = \frac{18}{150} = 0.12$$

Conditions that make the approximation reliable are verified

$$n \ge 30, n\hat{p} \ge 5, n\hat{q} \ge 5$$

From this one may compute:

pute: 
$$\hat{p} \pm z_{\varepsilon/2} \sqrt{\frac{\hat{p}\hat{q}}{n}} = 0.12 \pm 2.56 \sqrt{\frac{0.12 \times 0.88}{150}}$$

With a 99% confidence proportion is between 0.052 and 0.188

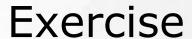




#### Computing confidence interval with R

```
> prop.test(x=18, n=150, conf.level = 0.99, correct = TRUE)
        1-sample proportions test with continuity correction
data: 18 out of 150, null probability 0.5
X-squared = 85.127, df = 1, p-value < 2.2e-16
alternative hypothesis: true p is not equal to 0.5
99 percent confidence interval:
 0.0648676 0.2088192
sample estimates:
0.12
```







- Simulate 3 random samples from a normal population of mean 15 and Standard deviation 2.
  - Sample sizes must be 9, 25 and 100 respectively
  - Compute a 95% confidence interval for the mean in each sample
- Use the following code

```
x9 <- rnorm (n=9, mean=15, sd=2)
x25 <- rnorm (n=25, mean=15, sd=2)
x100 <- rnorm (n=100, mean=15, sd=2)
t.test(x9)
t.test(x25)
t.test(x100)</pre>
```

What do you observe?





# Sample Size





# Sample Size Calculation

- Some questions must be answered before we can compute the sample size needed to estimate the mean or percentage.
  - Precision (interval range) of estimations ("how accurate I want the estimate to be"?)
  - Level of confidence of estimations ("how confident will I be on the estimation"?)





# Sample Size Calculation

- The question "what is the sample size" must be rephrased as:
  - What **sample size** is needed
  - to estimate the mean, so that
  - we have a high confidence (say 95%)
  - that the estimation error will be less than a given threshold?



# Sample size for mean



Remember: Confidence Interval for the mean

$$\begin{split} & \bar{X}_n \pm z_{\varepsilon/2} \frac{\sigma}{\sqrt{n}} = \bar{X}_n \pm precision \\ & precision = z_{\varepsilon/2} \times \frac{\sigma}{\sqrt{n}} \Longrightarrow \\ & n = \frac{z_{\varepsilon/2}^2 \sigma^2}{precision^2} \end{split}$$

#### Example:

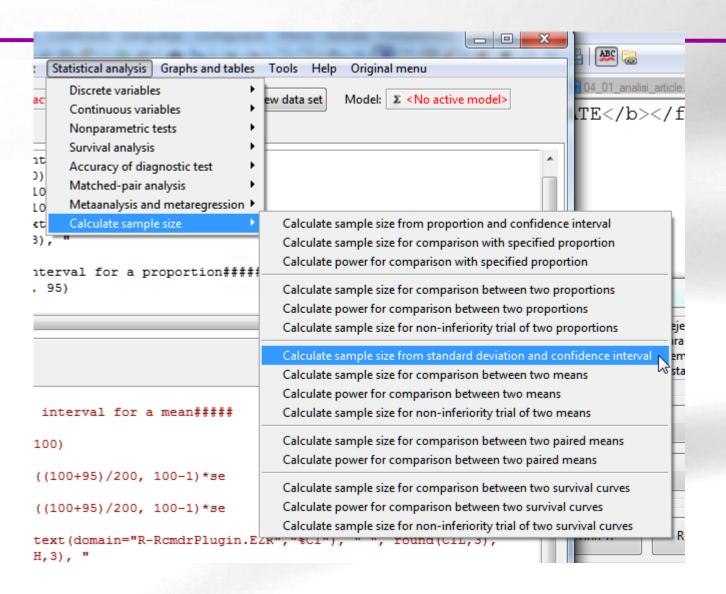
The sample size needed to estimate the mean with a confidence interval of width 10 (precision =10/2=5), a confidence level of 95%, if we know that the standard deviation is 20, will be:

$$n = 1.96^2 20^2 / 5^2 = 62$$



# Computing Sample Size with eZR







## Computing Sample Size with eZR



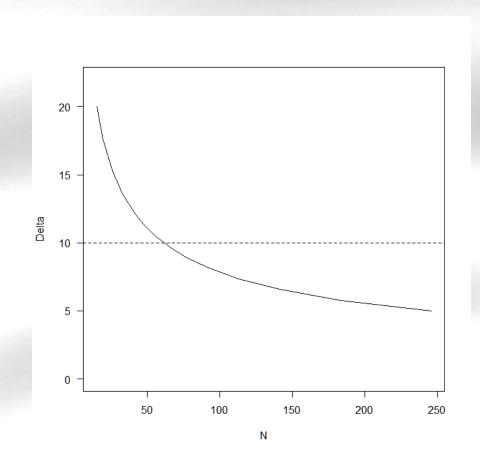
> SampleMeanCI(20, 10, 95)

Assumptions
-------------

Standard deviation 20 Confidence interval 10 Confidence level 0.95

**Estimated** 

Required sample size 62







### Sample size for proportion

$$\Pr ecision = z_{\varepsilon/2} \times ee = z_{\varepsilon/2} \times \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} \Rightarrow$$

$$n = \frac{z_{\varepsilon/2}^{2} \hat{p} (1 - \hat{p})}{precision^{2}}$$

If p is unknown one can take p=q=0.5

Assume precision is 5% (Interval =  $p\pm.05$ ) and confidence level is 95%

If it is known that p is around 12.5%

$$n = 1.96^2 .125 (1 - .125) / .05^2 = 168$$

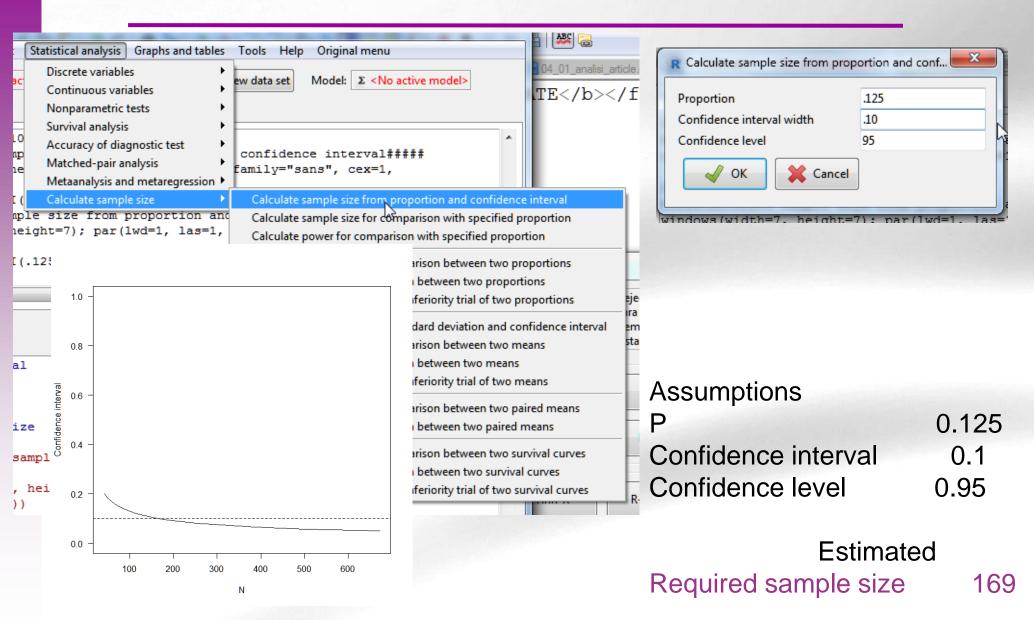
If p is unknown maximum sample size will be if p=.50

$$n=1.96^2.5(1-.5)/.05^2=384$$





### Computing Sample Size with eZR







# Sample size calculations for testing

- Similarly to sample size calculations for estimation, several points need to be considered so that the right question is:
- What is the sample size needed to detect at least a difference  $\Delta$  with the null hypothesis with a power  $\beta$  and a confidence  $(1-\alpha)$ 
  - The computations also need to know or estimate parameters such as standard deviation or the percentage.





# Examples

#### • The question:

 What sample size is needed to test the belief that systolic pressure in a hypertense population is 90 or bigger than that

#### Needs to be formulated as

- What sample size is needed to test the belief that systolic pressure in a hypertense population is 90 or bigger than that with a difference of at least 5 units, a power of 80% and a confidence of 95% assumint that the standard deviation is 11?





# Computing sample size with eZR

 eZR can only compute sample size for paired differences, so we cheat a little:

