

Principles of Statistical Inference

Curs d'Estadística Bàsica per a la Recerca Biomèdica

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The objective of statistical inference

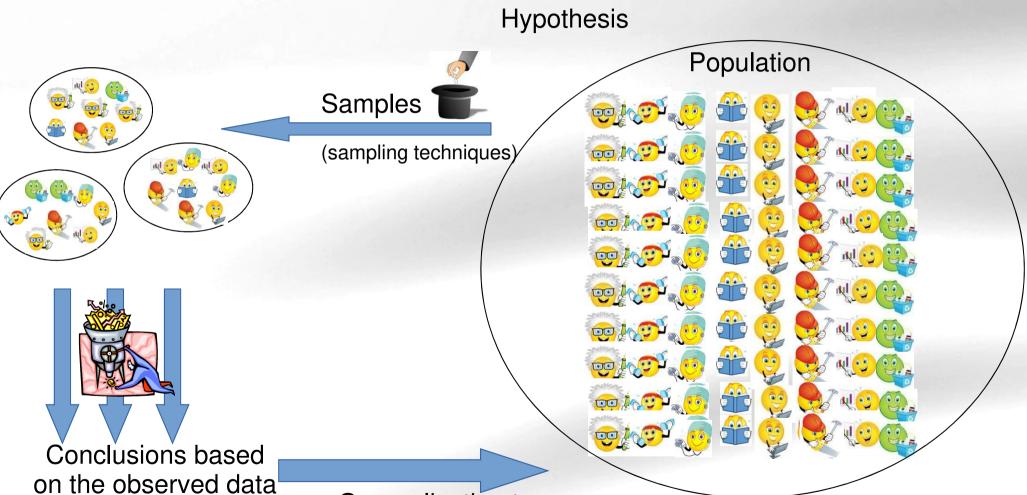
Taking the observed (measured) values of one (or more) of samples...

... Determine ("infer") the properties of the entire population.





The objective of statistical inference



(Statistical inference) (Parameters estimation) (Hypothesis testing) Generalization to the population



Estimation



- The aim of estimation is to infer properties (parameters) of the distribution of population data from sample data
- Some key concepts
 - Point estimate: Give a numerical value to the parameter of interest.
 - Estimator: Mathematical function to obtain the estimate
 - Interval Estimation: Give two values between which is the value of the population parameter with a preset confidence level (or probability)
 - Random error: Difference between estimation and real value if the sample is random



Point estimation (1)



- Data from qualitative variables
 - Parameter: Probability to observe a certain category
 - Estimate: Sample proportion: % of that category in the sample
 - Example: In the Osteoporosis dataset, what is the probability of observing a woman without ostheoporosis



Point estimation (II)



- Data from quantitative variables
 - Population parameters: μ , σ , etc.
 - Population parameters:
 - Estimate the mean, μ , with the sample mean, \overline{X}
 - Estimate, σ with the sample standard deviation, \hat{s}





Exercise

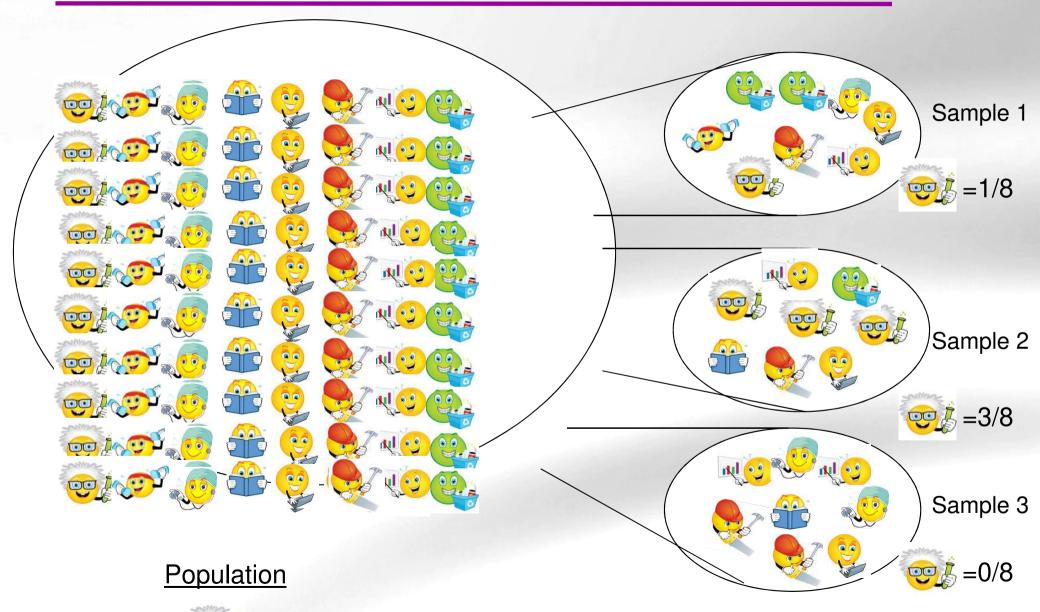
- In the osteoporosis dataset (osteo100) estimate the mean bone density (BUA)
 - for all the population indistinctly
 - depending on the CLASSIFIC variable





Biological variability. Sampling

=1/8







Sampling distribution

Population is 5 Children with age

$$x_1=6$$
, $x_2=8$, $x_3=10$, $x_4=12$, $x_5=14$

- Mean μ =10
- Variance $\sigma^2 = 8$
- Extract all possible samples with replacement and compute the mean in each sample

In this problem we can compute the population parameters because we know all the population values!!!



25 Samples n=2



	Second Data						
		6	8	10	12	14	
	6	6,6	6,8	6,10	6,12	6,14	
		(6)	(7)	(8)	(9)	(10)	
Fist Data	8	8,6	8,8	8,10	8,12	8,14	
		(7)	(8)	(9)	(10)	(11)	
	10	10,6	10,8	10,10	10,12	10,14	
		(8)	(9)	(10)	(11)	(12)	
	12	12,6	12,8	12,10	12,12	12,14	
		(9)	(10)	(11)	(12)	(13)	
	14	14,6	14,8	14,10	14,12	14,14	
		(10)	(11)	(12)	(13)	(14)	



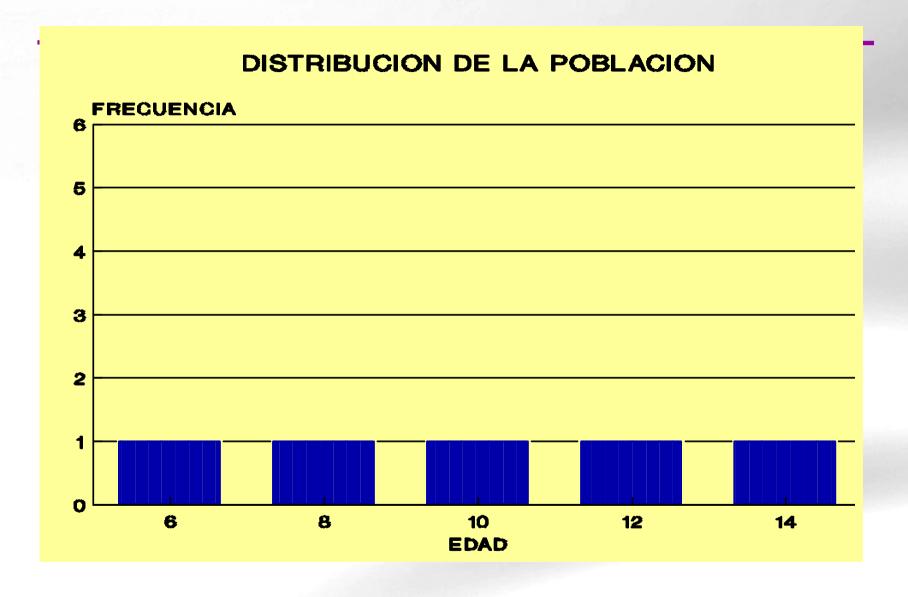
Frequency table



media	frecuencia	frec relativa			
6	1	1/25			
7	2	2/25			
8	3	3/25			
9	4	4/25			
10	5	5/25			
11	4	4/25			
12	3	3/25			
13	2	2/25			
14	1	1/25			





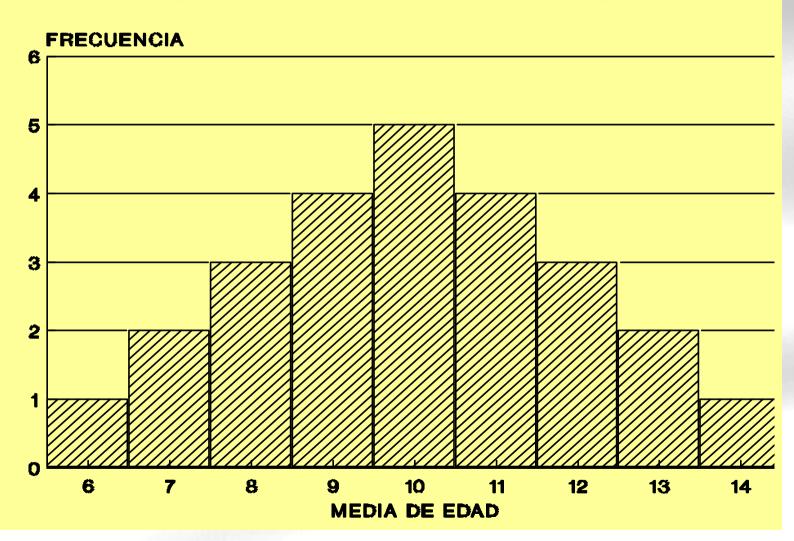






Histograma







Summary



Mean of 25 sample means

$$\mu_{\text{med}} = (6+7+...+14)/25=10$$

Variance of 25 sample means

$$\sigma^2_{\text{med}} = \{(6-10)^2 + (7-10)^2 + ... + (14-10)^2\}/25 = 4$$

The mean of sample means is population mean

$$\sigma_{\text{med}}^2 = \sigma^2/2 = 8/2 = 4$$

 Variance of 25 sample means equals population variance divided by sample size



Standard error



- Standard deviation of the distribution of sample means
- Usually it is defined as population standard deviation divided by squared root of sample size

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



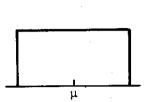
DISTRIBUTION IN THE POPULATION

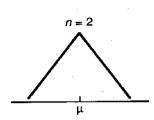
SAMPLING DISTRIBUTION OF THE MEAN, \bar{X}

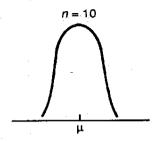
IEAN, \bar{X} n = 30

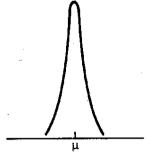
UNITAT D'ESTADÍSTICA I BIOINFORMÀTICA

Row A. Uniform or rectangular

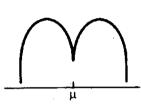


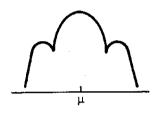


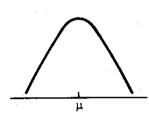


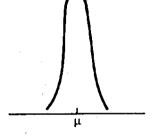


Row B. Bimodal

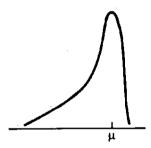


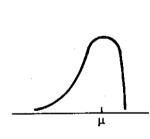


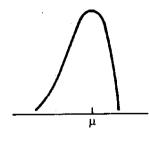


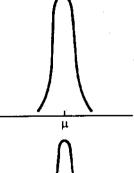


Row C. Skewed

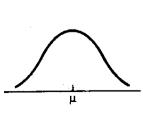


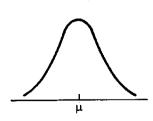


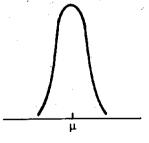




Row D. Similar to normal







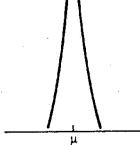


Figure 6-3. Illustration of ramifications of central limit theorem.



Unbiased estimators



- An estimator is unbiased if the mean of the sample estimates is the parameter we are looking for.
 - Sample mean and proportion are unbiased estimators of population mean and probability (percentage)
 - Sample variance is a biased estimator of population variance, but not if we divided by n-1
 - That is why computers compute sample variance dividing by (n-1) instead of dividing by n.



Confidence interval of Mean

- Population blood pressure in hipertensives is normally distributed with mean μ and standard deviation 12
- We extract a sample of n=186 and we observe a sample mean m=118,8)
- We can compute a *confidence interval* for the mean:

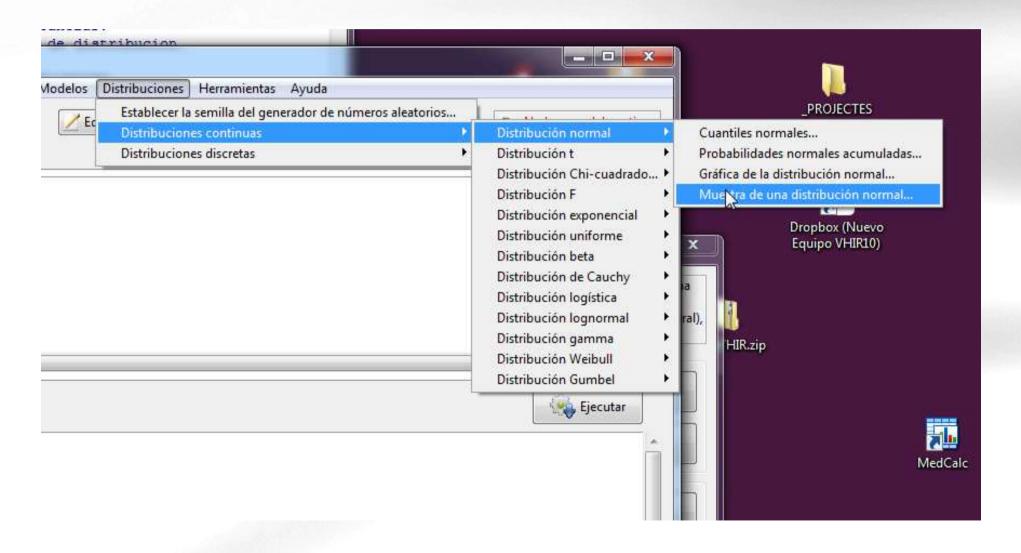
$$\overline{x} \pm z_{\frac{\alpha}{2}} \times \frac{s}{\sqrt{n}} = 118 \pm 1,96 \times 12/\sqrt{186}$$

- This provides an interval such that we are highly confident that the true population may be between the upper and lower value of the interval.
 - In practice this means that if we repeated the process of sampling and building the interval we would expect that 95% of the times it would contain the true population value





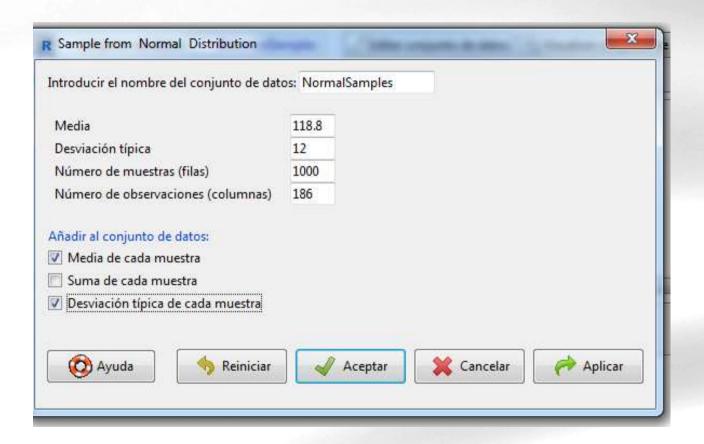
Let's simulate 1000 samples of size 186 with mean 118.8 and standard deviation 12







Let's simulate 1000 samples of size 186 with mean 118.8 and standard deviation 12

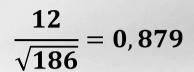


Calculate the mean and standard deviation of the mean of the 1000 samples Calculate the mean and standard deviation of the sd of the 1000 samples



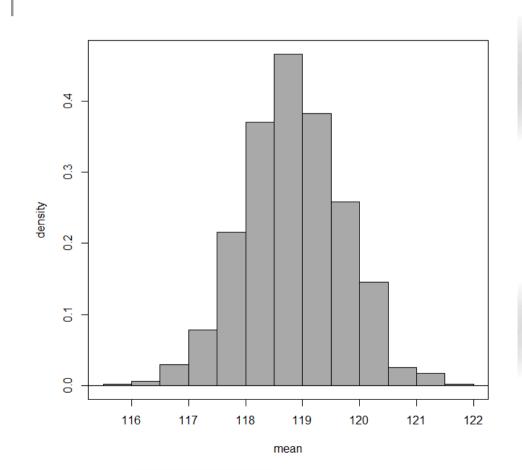


mean sd mean 118.84143 0.8746236 sd 11.97547 0.6613225



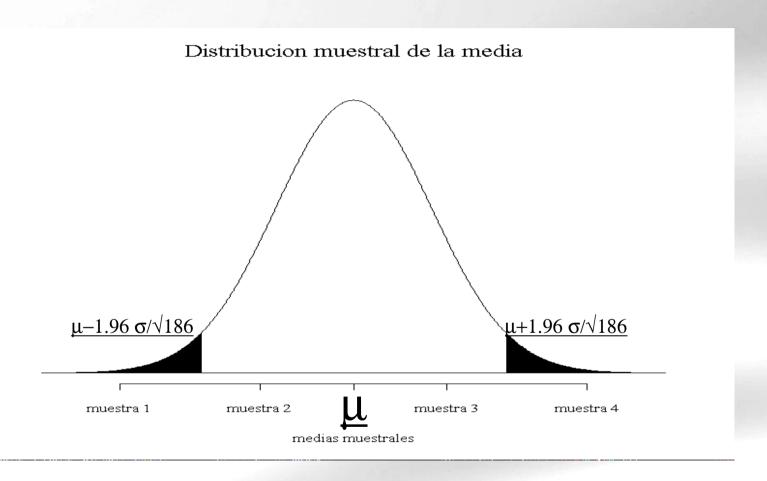
Standard ERROR

True mean =118,8



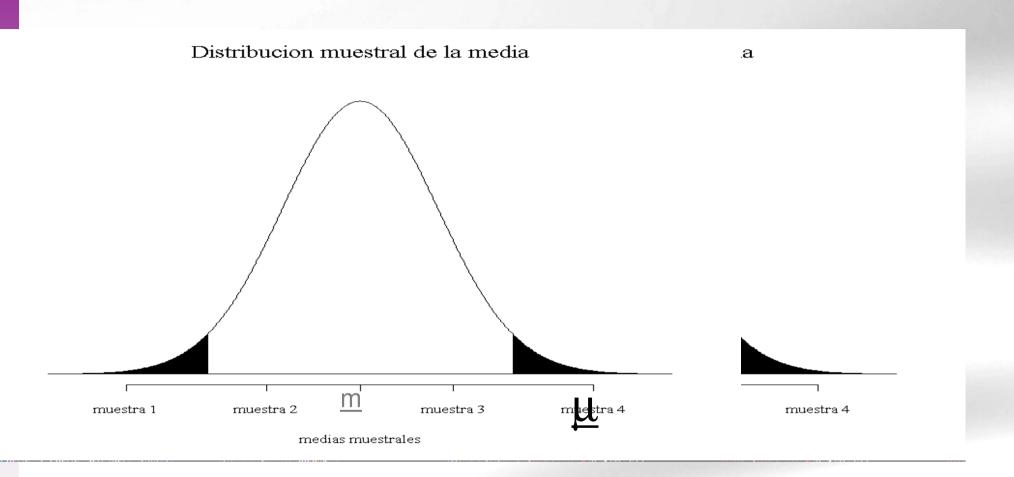








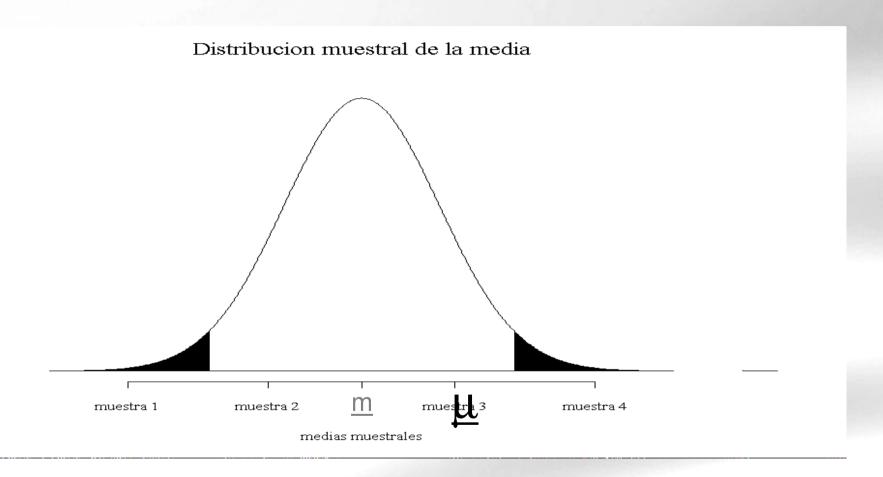




Population mean is outside de confidence interval



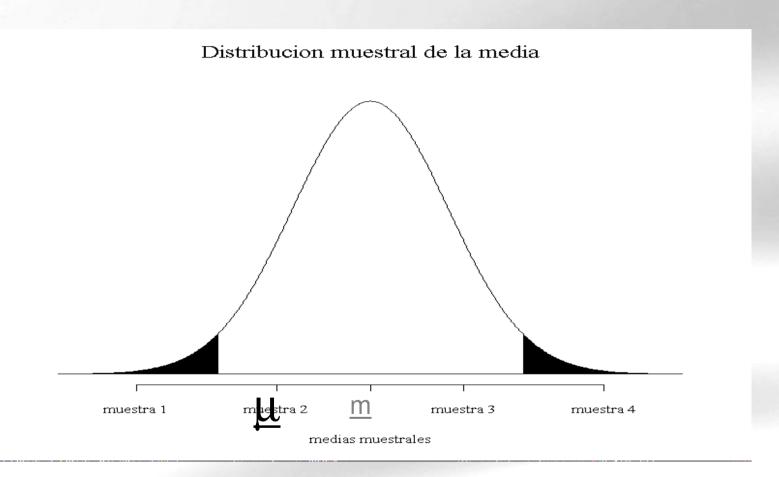




Population mean is inside confidence interval





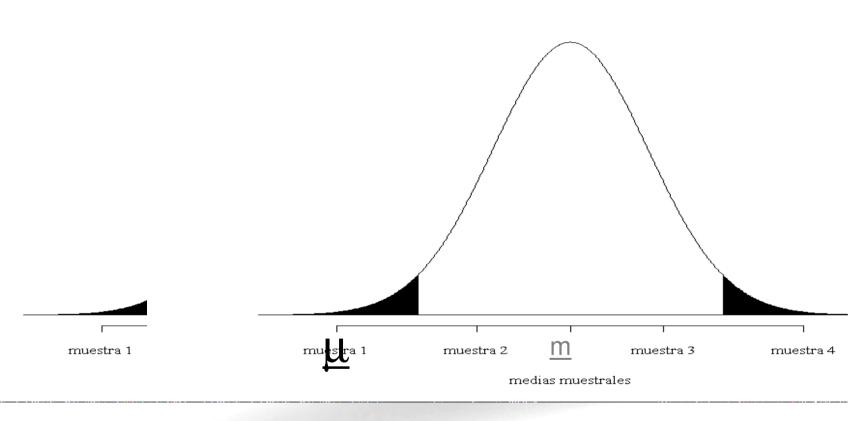


Population mean is inside confidence interval







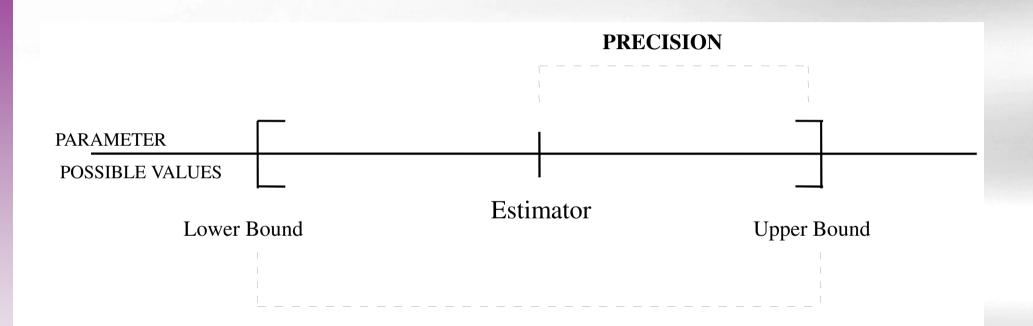


Population mean its outside confidence interval





Confidence interval

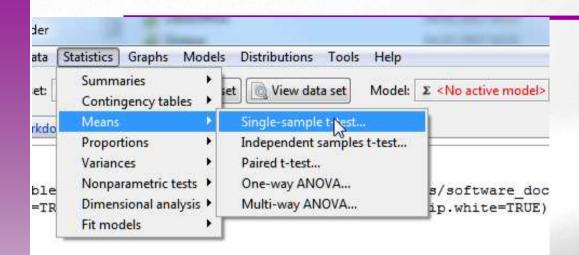


Values in which we are confident that real population parameter is inside With a prefixed confidence level (Usually 95%)



Confidence intervals in RCmdr





Variable (pick one) area									
oua			N						
	Ē		4						
edad_men									
_nac									
menarqui									
Alternative Hypothesis	52		5000			2011			
Population mean !=	mu0	Null h	ypothesi	s: mu	= /	0.0			
Population mean < r	mu0	Confid	lence Le	vel:	.95				
Population mean > r	mu0								
	and plants				- 2				.55
(C) Help	Res] [D	ок		w .	ancel	4	Apply





One Sample t-test

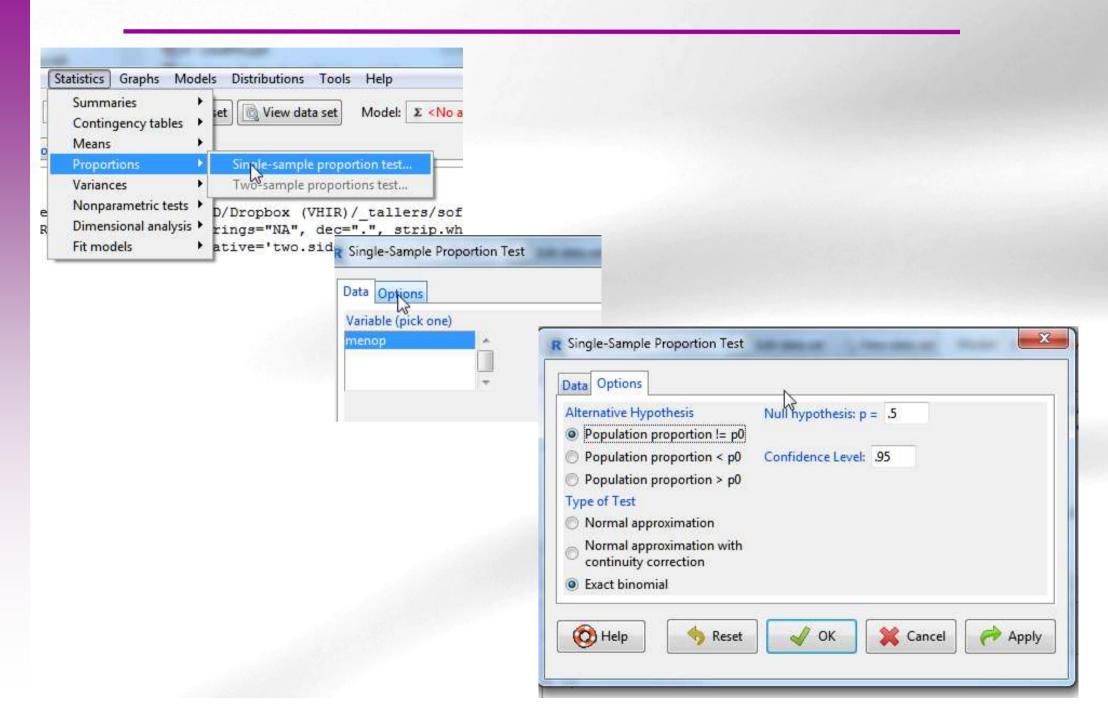
data: bua t = 137.89, df = 999, p-value < 2.2e-16 alternative hypothesis: true mean is not equal to 0 95 percent confidence interval: 72.2539 74.3401

sample estimates: mean of x 73.297





Confidence interval in RCmdr





Proportion Test Normal Aproximation



Frequency counts (test is for first level):

menop

NO SI

303 697

1-sample proportions test without continuity correction

data: rbind(.Table), null probability 0.5

X-squared = 155.24, df = 1, p-value < 2.2e-16

alternative hypothesis: true p is not equal to 0.5

95 percent confidence interval:

0.2753154 0.3321923

sample estimates:

p

0.303





Frequency counts (test is for first level): menop NO SI 303 697

Exact binomial test

data: rbind(.Table)
number of successes = 303, number of trials = 1000, p-value < 2.2e-16
alternative hypothesis: true probability of success is not equal to 0.5
95 percent confidence interval:
0.274632 0.332533

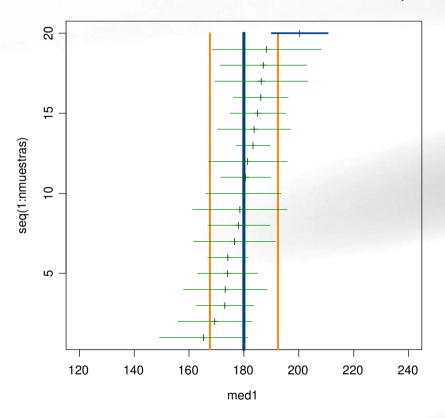
sample estimates: probability of success 0.303



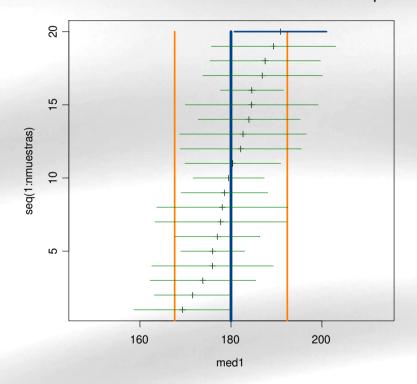


Sample size =10, Mean=180, sd=20

20 muestras de tamaño 10 media 180 desv.tip. 20



20 muestras de tamaño 10 media 180 desv.tip. 20

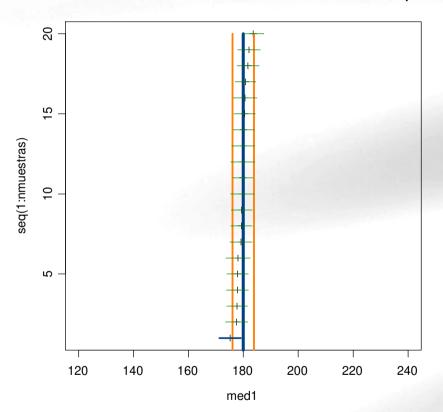




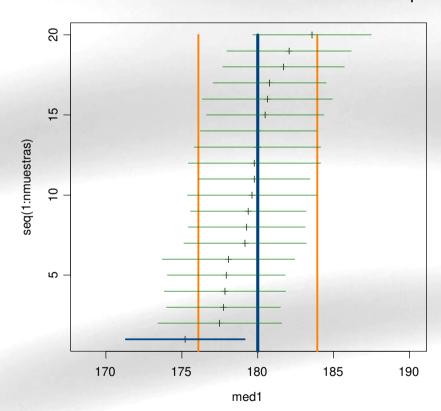


Sample size = 100, Mean=180, sd=20

20 muestras de tamaño 100 media 180 desv.tip. 20



20 muestras de tamaño 100 media 180 desv.tip. 20

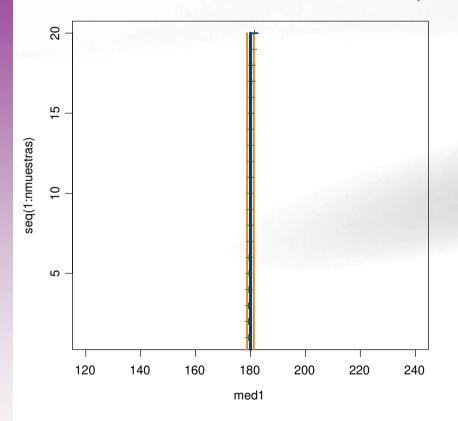




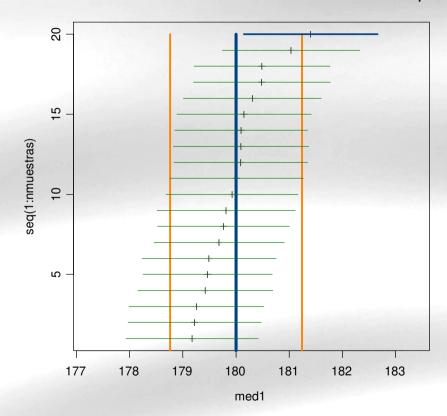


Sample size = 100, Mean=180, sd=20

20 muestras de tamaño 1000 media 180 desv.tip. 20



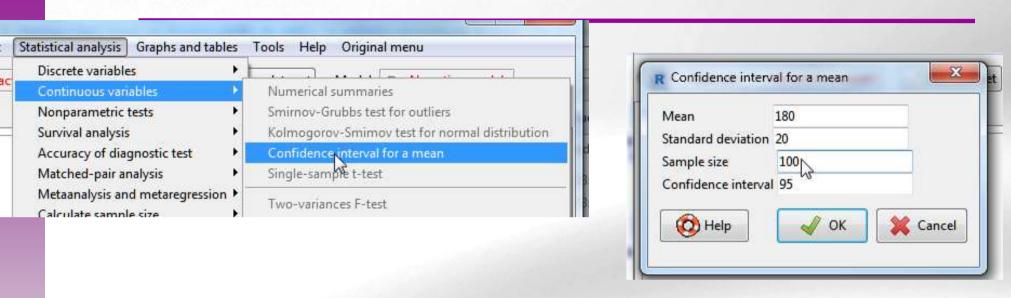
20 muestras de tamaño 1000 media 180 desv.tip. 20







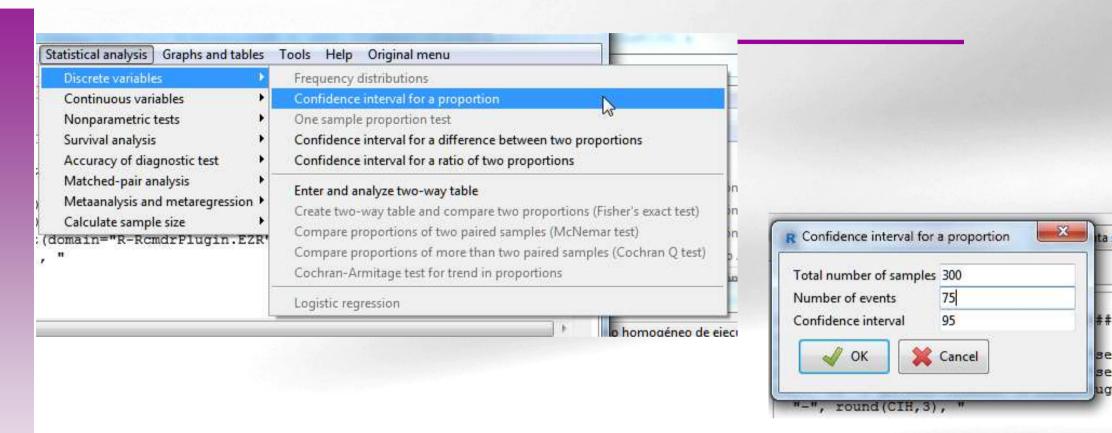
Confidence interval calculator (Plugin EzR)



95 %CI 176.032-183.968







[1] Probability: 0.25

[1] 95% confidence interval: 0.202 - 0.303