

Chapter 8 Confidence Interval:

\bar{X} , S

An interval estimate a population parameter is Confidence Interval.

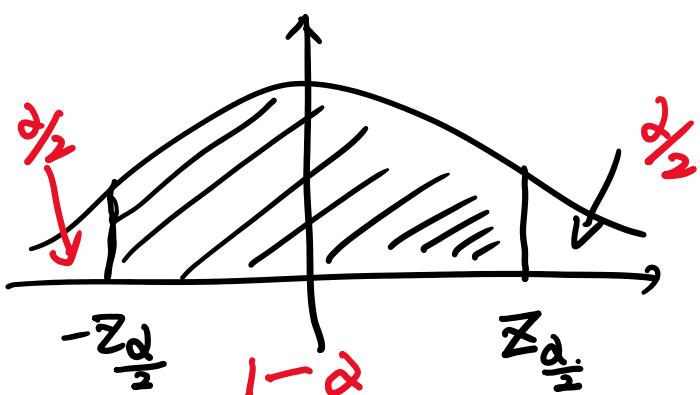
Suppose X_1, \dots, X_n is a random sample from a normal distribution with unknown mean μ and known variance σ^2 .

Central Limit Theorem.

\bar{X} is normally distributed with mean μ and variance $\frac{\sigma^2}{n}$.

$$Z_{\frac{\alpha}{2}} = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim N(0, 1)$$

standard normal
mean 0, variance 1.



$$0 \leq \alpha \leq 1$$

$$P(-Z_{\alpha/2} \leq Z \leq Z_{\alpha/2}) = 1 - \alpha.$$

$$P\left(-\frac{Z_{\alpha/2}}{2} \leq Z \leq \frac{Z_{\alpha/2}}{2}\right) = 1 - \alpha.$$

Example. $\alpha = 5\%$.

$$Z_{\alpha/2} = Z_{0.025}$$

$$Z_{\alpha/2} = 1.96 \quad -Z_{\alpha/2} = -1.96$$

$$P(-1.96 \leq Z \leq 1.96) = 95\%$$

$$P\left[-\frac{Z_{\alpha/2}}{2} \leq \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \leq \frac{Z_{\alpha/2}}{2}\right] = 1 - \alpha.$$

↓ sample mean
 ↓ population parameter
 ↑ n : sample size.
 σ known.

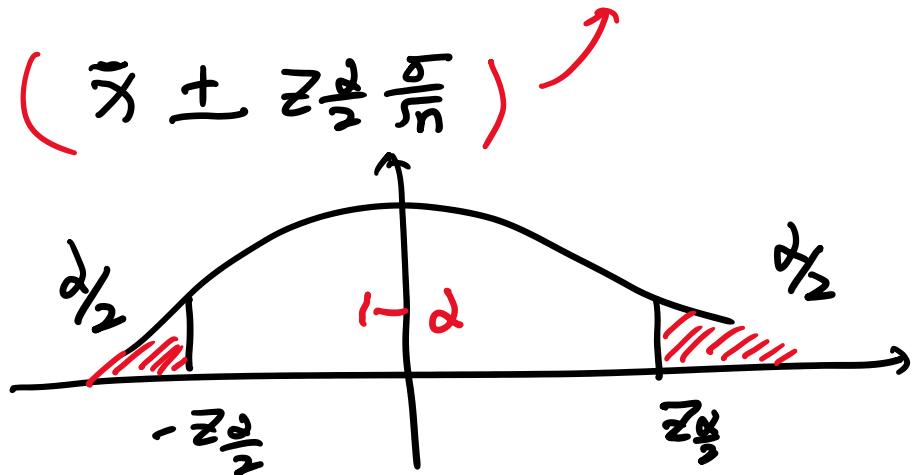
$$P\left[-\frac{Z_{\alpha/2} \cdot \sigma}{\sqrt{n}} \leq \bar{x} - \mu \leq \frac{Z_{\alpha/2} \cdot \sigma}{\sqrt{n}}\right] = 1 - \alpha.$$

$$P\left[\bar{x} - \frac{Z_{\alpha/2} \cdot \sigma}{\sqrt{n}} \leq \mu \leq \bar{x} + \frac{Z_{\alpha/2} \cdot \sigma}{\sqrt{n}}\right] = 1 - \alpha.$$

If \bar{x} is the sample mean of a random sample of size n from a normal population with known variance.

Then, $100(1-\alpha)\%$ C.I. on μ is given by.

$$\bar{x} - \frac{Z_{\alpha/2} \cdot \sigma}{\sqrt{n}} \leq \mu \leq \bar{x} + \frac{Z_{\alpha/2} \cdot \sigma}{\sqrt{n}}$$



Example:

Assume we have heights of students.
 $n=10$, $\bar{x} = 164.46$, $\sigma = 1$

Find the 95% CI of the heights:
 $100(1-\alpha)\%$.

$$\bar{x} - z_{\alpha/2} \frac{\sigma}{\sqrt{n}} \leq \mu \leq \bar{x} + z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$$

$$164.46 - 1.96 \cdot \frac{1}{\sqrt{10}} \leq \mu \leq 164.46 + 1.96 \cdot \frac{1}{\sqrt{10}}$$

Interpreting a Confidence Interval.

Lower
R.V.

Upper
R.V.

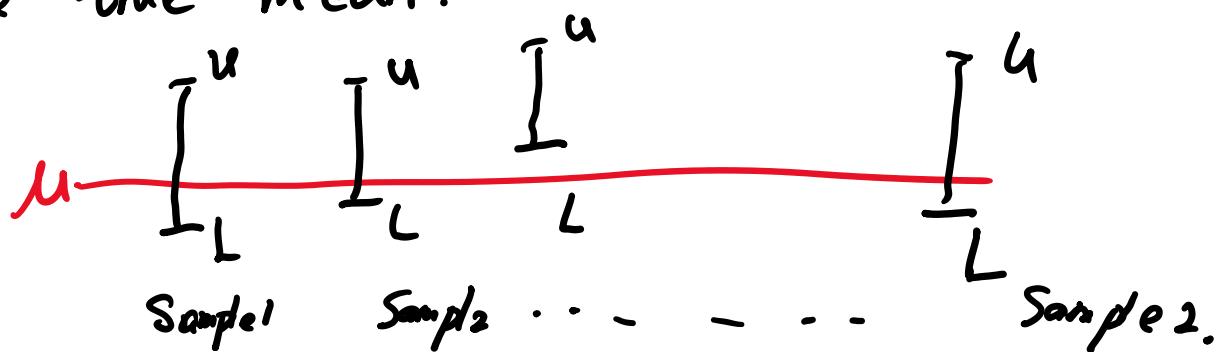
95% confidence interval.

repeat doing sampling n times.

repeat doing sampling n times.

$n \rightarrow$ large.

95% of these n C.I will contain the true mean.



We are 95% certain / confident that μ lies in the interval.

Marginal Error:

$$E = \frac{Z_{\alpha/2} \cdot \sigma}{\sqrt{n}} \quad \text{Marginal Error (Error).}$$

the larger the sample size, the less the error is.

$$\bar{x} - E \leq \mu \leq \bar{x} + E.$$

$$E = 0.01$$

Rearrange.

$$n = \left(\frac{Z_{\alpha/2} \sigma}{E} \right)^2$$

If \bar{x} is used as an estimate of μ , we can be $100(1-\alpha)\%$. confident that \bar{x} is an unbiased estimator.

we can be 100(1 - α)% confident
the error $|\bar{x} - \mu|$ will not exceed a specified
amount E when the sample size is

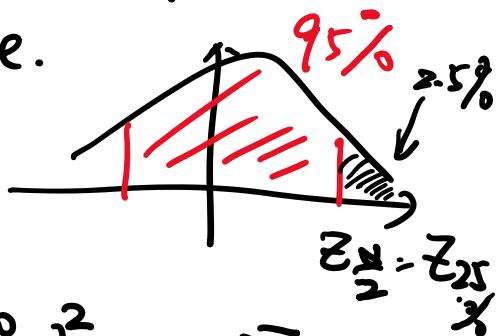
$$n = \left(\frac{Z_{\frac{\alpha}{2}} \sigma}{E} \right)^2 \quad \text{Given any 3, you could find the last one.}$$

Example:

knowing that the population variance for
of customers is 100.

Find 95% CI with error of 0.5.
What's the minimum sample size.

$$E \leq 0.5$$



$$n = \left(\frac{Z_{2.5\%} \sigma}{E} \right)^2 = \left(\frac{1.96 \cdot 10}{0.5} \right)^2 = 153.$$

Narrow Confidence Interval:

$$\bar{x} - \frac{Z_{\frac{\alpha}{2}} \sigma}{\sqrt{n}} \leq \mu \leq \bar{x} + \frac{Z_{\frac{\alpha}{2}} \sigma}{\sqrt{n}}$$

$100(1-\alpha)\%$

97.5%

① Lower Confidence Level (Larger α)

② Lower Standard deviation.

$$(-3, 1+3)$$

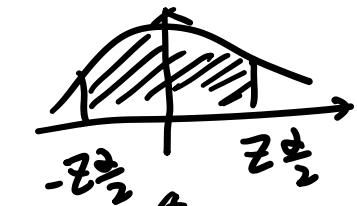
$$(-2, 1+2)$$

$$(-2, 4)$$

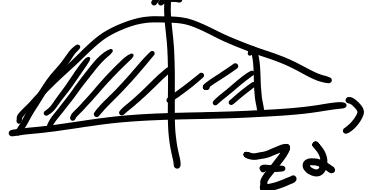
$$(-1, 3)$$

③ Larger sample size n .

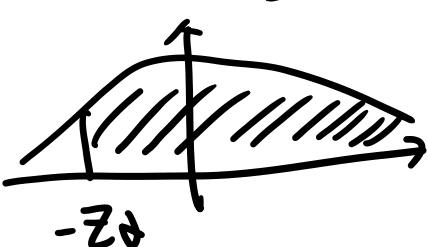
③ Larger sample size n .



two sided C.I.



$(-\infty, U)$



$(L, +\infty)$

one sided
C.I.

$$L \geq \bar{x} - \underline{z_{\alpha}} \frac{\sigma}{\sqrt{n}}$$

$(L, +\infty)$

$$U \leq \bar{x} + \underline{z_{\alpha}} \frac{\sigma}{\sqrt{n}}$$

$(-\infty, U)$.

General method to derive a C.I.

Pivot: A distribution does not depend on any unknown parameter.

For example:

$$Z = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \sim N(0, 1)$$

Large Sample Confidence Interval:

Let X_1, \dots, X_n a random sample from a population with unknown mean μ and unknown variance σ^2 . \bar{x}, S^2

variance σ^2 \bar{x}, s^2

When n is large enough. ($n \geq 30$)

$$Z = \frac{\bar{x} - \mu}{s/\sqrt{n}} \sim N(0,1).$$

Approximately Standard Normal.

Sample

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graph LR
    Sample -- "σ known" --> Z1["Z = (x̄ - μ) / σ/√n"]
    Sample -- "σ unknown" --> Z2["Z = (x̄ - μ) / s/√n"]
    Z2 -- "n ≥ 30" --> Z1
    Z2 -- "n < 30" --> T["T = (x̄ - μ) / s/√n"]
    T -- "? t distribution" --> Z1
  
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$Z = \frac{\bar{x} - \mu}{s/\sqrt{n}}$

$T = \frac{\bar{x} - \mu}{s/\sqrt{n}}$

T distribution:

Gosset. (Student t distribution).

Let x_1, \dots, x_n be a random sample from a normal distribution with unknown μ and unknown variance σ^2 .

Then

$$T = \frac{\bar{x} - \mu}{s/\sqrt{n}}$$

has a t distribution with $n-1$ degree of freedom.

freedom.

$$\underline{x_1 + \dots + x_n = 100}$$

$n-1$ degree of freedom.

P.D.F. for τ :

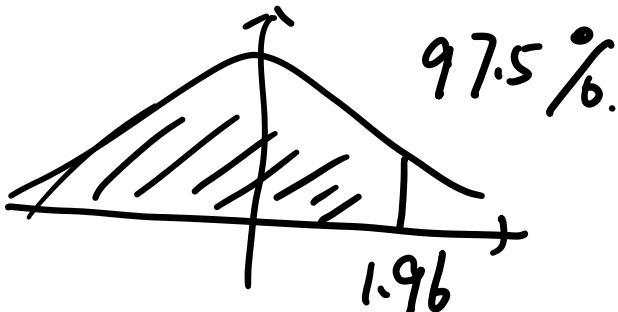
$$f(\tau) = \frac{\Gamma[(k+1)/2]}{\sqrt{\pi k} \Gamma(k/2)} \frac{1}{[\frac{\tau^2}{k} + 1]^{(k+1)/2}}, \quad -\infty < \tau < \infty.$$

where k is the number of degree of freedom.

$$\bar{E}(X) = 0$$

$$\text{Var}(X) = \frac{k}{k-2}.$$

T-table. Z-table



Example: $n=22$. $\bar{x}=13.71$, $S=3.55$.

Find 95% two sided }
one sided upper } C.I.
lower }

(1) two sided:

$$n-1=21$$

two sided

$$\alpha = 5\% \\ = 0.05$$

$$n-1=21 \quad \text{LW since} \quad \alpha = \beta \\ = 0.05$$

$t_{21, \text{two side}} \alpha = 5\%$

$$= 2.08$$

$$\bar{x} - t_{21, n-1} \frac{s}{\sqrt{n}} \leq \mu \leq \bar{x} + t_{21, n-1} \frac{s}{\sqrt{n}}$$

$$13.71 - 2.08 \cdot \frac{3.55}{\sqrt{22}} \leq \mu \leq 13.71 + 2.08 \cdot \frac{3.55}{\sqrt{22}}$$

(2) One sided:

$t_{21, \text{one side}} \alpha = 5\%$

$$= 1.721$$

$$\text{Upper: } \mu \leq 13.71 + 1.721 \cdot \frac{3.55}{\sqrt{22}}$$

$$\text{Lower: } \mu \geq 13.71 - 1.721 \cdot \frac{3.55}{\sqrt{22}}$$

Two sided

$$P\left(\bar{x} - t_{21, n-1} \frac{s}{\sqrt{n}} \leq \mu \leq \bar{x} + t_{21, n-1} \frac{s}{\sqrt{n}}\right) = 1-\alpha$$

$$\left(\bar{x} - t_{21, n-1} \frac{s}{\sqrt{n}}, \bar{x} + t_{21, n-1} \frac{s}{\sqrt{n}}\right)$$

One sided

$$\left(\bar{x} - t_{21, n-1} \frac{s}{\sqrt{n}}, +\infty\right) \text{ Lower}$$

$$\left(-\infty, \bar{x} + t_{21, n-1} \frac{s}{\sqrt{n}}\right) \text{ Upper}$$

25. Sample Test 2.

$$n=15, \underbrace{95\%}_{\alpha=0.05} \Rightarrow \underline{(83.9219, 90.6781)}$$

What's 99% C.I.

$$\bar{x} = \frac{83.9219 + 90.6781}{2} = 87.3$$

95% } ME = 87.3 - 83.9219 = 3.3781.
ME = $\bar{x} - \text{Lower}$

$$n-1=14$$

$$ME = t_{\alpha/2, n-1} \cdot \frac{s}{\sqrt{n}}$$

$$3.3781 = 2.145 \cdot \frac{s}{\sqrt{n}}$$
$$\frac{s}{\sqrt{n}} = 1.57487$$

$$\frac{t_{2.5\%, 14}}{= 2.145}$$

99% } ME = $t_{0.5\%, 14} \cdot \frac{s}{\sqrt{n}}$ $t_{0.5\%, 14}$
 $= 2.977 \cdot 1.57487 = 2.977$
 $= 4.6883933$

$$99\% \text{ CI: } (87.3 - 4.6883933, 87.3 + 4.6883933)$$
$$= (82.61, 91.99)$$

Large Sample C.I. for population proportion.

$$\sigma = \sqrt{\frac{np(1-p)}{n}}$$

$$\Sigma = \frac{x - np}{\sqrt{np(1-p)}}$$

$$\Sigma = \frac{\hat{p} - p}{\sqrt{\frac{p(1-p)}{n}}} \quad \text{where } \hat{p} = \frac{x}{n}$$

$$P(\hat{p} - Z_{\frac{\alpha}{2}} \sqrt{\frac{p(1-p)}{n}} \leq p \leq \hat{p} + Z_{\frac{\alpha}{2}} \sqrt{\frac{p(1-p)}{n}}) = 1 - \alpha.$$

Example: $x = 10$ out of $n = 85$.

$$p = \frac{10}{85} = 0.12.$$

What's 95% CI for p (two-sided)

$$n = 85, \hat{p} = \frac{10}{85} = 0.12.$$

$$-Z_{\frac{\alpha}{2}} = -1.96$$

$$(0.12 - 1.96 \cdot \sqrt{\frac{0.12 \cdot 0.88}{85}}, 0.12 + 1.96 \cdot \sqrt{\frac{0.12 \cdot 0.88}{85}})$$

$$95\% \text{ C.I. } (0.0509, 0.2243)$$

Choice of sample size:

$$n = \frac{Z_{\frac{\alpha}{2}}^2 p(1-p)}{\epsilon^2}$$

$$\therefore n = \left(\frac{Z_{\alpha/2}}{E} \right)^2 p(1-p)$$

If p is not specified $p=0.5$.

$$n = \left(\frac{Z_{\alpha/2}}{E} \right)^2 \cdot 0.5^2$$

One-sided P C.I.: $(100(1-\alpha)\%)$

Upper: $P \leq \hat{P} + Z_{\alpha} \cdot \sqrt{\frac{\hat{P}(1-\hat{P})}{n}}$

Lower: $P \geq \hat{P} - Z_{\alpha} \cdot \sqrt{\frac{\hat{P}(1-\hat{P})}{n}}$

Test 2.