

# Set 22 - Confidence Intervals

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## Stat 260 Lecture Notes

### Set 22 - Confidence Intervals

→ calculate t number

Recall: A **point estimate** is a single valued statistic used to estimate a population parameter.  $\bar{x}$  is a point estimate for  $\mu$ .  
*sample mean* *population mean* *no sense of how accurate it is*

The downside of using a point estimate such as  $\bar{x}$  is that we don't know how accurate our estimate is. Another method of estimation is to give a range of possible values - an **interval estimate**.

→ anything in interval is good estimate

A **confidence interval (CI)** for  $\mu$  is an interval  $[L, U]$  which gives an estimate for the population mean  $\mu$  with some degree of certainty.

*Lower* *Upper*

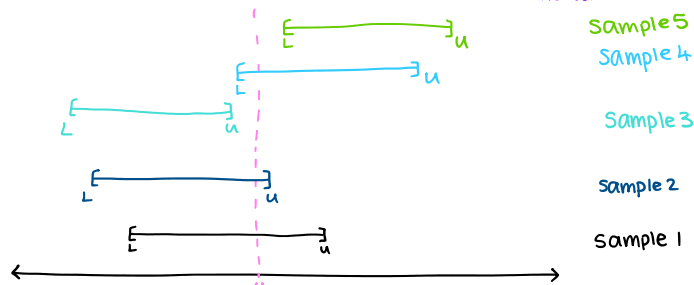
A 95% confidence interval for  $\mu$  has  $P(L \leq \mu \leq U) = 0.95$

A 99% confidence interval for  $\mu$  has  $P(L \leq \mu \leq U) = 0.99$ .

We find the numerical values for  $L$  and  $U$  by sampling. From each round of sampling we get an estimated range of values that  $\mu$  might be contained in.

**Interpretation:** For a 95% CI, in the long run 95% of the CIs we create by sampling will actually contain  $\mu$ .

*confidence interval*



95% of these intervals should contain  $\mu$  in the long run.

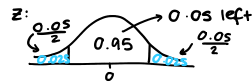
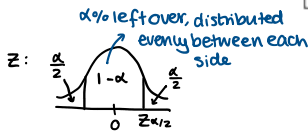
As long as  $\mu$  in interval = good job

\* Width of intervals not always the same

How do we find  $L$  and  $U$ ?

A  $(1 - \alpha)\%$  CI for  $\mu$  is

$$[L, U] = \left[ \bar{x} - z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}, \bar{x} + z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}} \right]$$



Sometimes we write this as

$$\bar{x} \pm z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}} \quad \text{or} \quad \bar{x} \pm z_{\alpha/2} \cdot \frac{s}{\sqrt{n}}$$

estimate  $\pm$  (critical value)  $\cdot$  (standard error)   
 (or estimated)   
  $\rightarrow$  to get rid of ambiguity

$\sigma$  = std. dev. for  $x$

$\frac{\sigma}{\sqrt{n}}$  = std. dev. for  $\bar{x}$  = standard error

$\frac{s}{\sqrt{n}}$  = estimated standard error

When can we use this formula?

Whenever  $\frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$  or  $\frac{\bar{X} - \mu}{s/\sqrt{n}}$  is normally distributed:

- (set 20) •  $X_1, X_2, \dots, X_n$  from a normal distribution and we know  $\sigma$
- (set 21) CLT •  $X_1, X_2, \dots, X_n$  from any distribution and  $n$  is big ( $n \geq 30$ ) and we know  $\sigma$
- (set 21) •  $X_1, X_2, \dots, X_n$  from any distribution and  $n$  is big ( $n \geq 30$ ) and we don't know  $\sigma$  (so we use the estimate  $s$  instead)

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e.s.e on formula sheet

estimate  $\pm$  (c.v.) (e.s.e)

$\rightarrow$  on formula sheet, this is a pattern.

$\bar{x}$  is normal

Will be written in words  $\rightarrow n=15$  "have a sample of 15"

Example 1:  $X_1, X_2, \dots, X_{15}$  from a normal distribution with  $\bar{x} = 147.33$  and  $s = 40$ . Find a 95% confidence interval for  $\mu$  and find a 99% confidence interval for  $\mu$ .

95% CI:  $\bar{x} \pm z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}$

$= 147.33 \pm (1.96 \cdot \frac{40}{\sqrt{15}})$  order of operations matter!!!

$= [127.09, 167.57]$

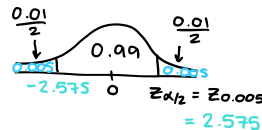
$\rightarrow$  need to calculate the lower and upper limits



99% CI:  $\bar{x} \pm z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}$

$= 147.33 \pm 2.575 \cdot \frac{40}{\sqrt{15}}$

$= [120.74, 173.92]$



Example 2: How does changing  $\sigma$ ,  $n$ , and the confidence level  $(1 - \alpha)\%$  affect the CI?

$\bar{x}$	$\sigma$	$n$	confidence level	$\alpha/2$	$z_{\alpha/2}$	$[L, U]$
147.33	40	15	95%	.025	1.96	[127.09, 167.57]
147.33	40	50	95%	.025	1.96	[136.25, 158.42]
147.33	40	15	99%	.005	2.575	[120.74, 173.92] $\leftarrow$ CI wider
147.33	80	15	95%	.025	1.96	[106.85, 187.82]

increasing the confidence level (%)  $\Rightarrow$  CI wider   
  $\rightarrow$  be more sure, pick wider interval,  $\mu$  more likely to be in there

$\rightarrow$  increasing  $n \Rightarrow$  narrower CI

increasing  $\sigma \Rightarrow$  wider CI

$\rightarrow$  is how much data is spread apart

Most problems will be finding confidence intervals

$\rightarrow$  work is finding  $z_{\alpha/2}$

$z_{\alpha/2}$  always positive

\*increasing  $n \Rightarrow$  narrower CI  
 increasing  $\sigma \Rightarrow$  wider CI  
 $\hookrightarrow \sigma$  is how much data is spread apart

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Example 3: How did we get the formula

$$[L, U] = \left[ \bar{x} - z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}, \bar{x} + z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}} \right] ?$$

Omit Idea:  $z = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}}$  rearrange and solve for  $\mu$ .

Look at the 95% confidence interval, where we have  $z_{\alpha/2} = z_{0.025} = 1.96$ .

We can say

$$P(L \leq \mu \leq U) = 0.95 \quad \text{this is okay!}$$

$$P\left(\bar{x} - 1.96 \cdot \frac{\sigma}{\sqrt{n}} \leq \mu \leq \bar{x} + 1.96 \cdot \frac{\sigma}{\sqrt{n}}\right) = 0.95 \quad \text{this is also okay!}$$

but once we fill in values for  $\bar{x}$ ,  $\sigma$ , and  $\sqrt{n}$  we can no longer use this probability.

So for example we can't say  $P(127.09 \leq \mu \leq 167.57) = 0.95$  with our confidence interval  $[127.09, 167.58]$  because either  $\mu$  is in this interval, or it isn't. (There are no variables here so there is no chance on where the value of  $\mu$  sits. The probability is 1 if the value is in the interval, or 0 if it isn't.)

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