Economics 103 – Statistics for Economists

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Lecture 16

Confidence Intervals – Part I

What We've Done So Far

- ▶ Random Sampling: $X_1, ..., X_n \sim \text{iid}$
- lacktriangle Use estimator $\widehat{ heta}$ to learn about population parameter $heta_0$
- Estimator $\widehat{\theta}$ is a random variable:
 - lacktriangle Distribution of $\widehat{ heta}$ is called sampling distribution
 - Bias of an estimator
 - Variance of an estimator
 - Mean-squared Error (MSE) of an estimator
 - Consistency of an Estimator

Inference

Confidence Intervals

What values of θ_0 are consistent with the data we observed?

Hypothesis Testing

I think that $\theta_0=0$. Do the data we observed suggest that I should change my mind?

Am I Taller Than The Average American Male?



Source: Centers for Disease Control (pg. 16)

My height is 73 inches. Based on a sample of US males aged 20 and over, the Centers for Disease Control (CDC) reported a mean height of about 69 inches in a recent report.

Clearly I'm taller than the average American male!

Do you agree or disagree?

- (a) Agree
- (b) Disagree
- (c) Not Sure

Just because the sample mean is 69 inches it doesn't follow that the population mean is 69 inches!

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What Else Should We Consider?

► How big was the sample?

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 - If the sample was very small there's a higher chance that it won't be representative of the population as a whole

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 - ▶ Why? The variance of the sample mean is *decreasing with* sample size so bigger samples are less noisy.

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 - Why? The variance of the sample mean is decreasing with sample size so bigger samples are less noisy.
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 - If everyone is very similar in height, any sample we take will be representative of the population.

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 - ▶ Why? The variance of the sample mean is *decreasing with* sample size so bigger samples are less noisy.
- How much variability is there in height in the population?
 - If everyone is very similar in height, any sample we take will be representative of the population.
 - ▶ Remember: the variance of the sample mean is *increasing* with the population standard deviation.

Am I Taller Than The Average American Male?

Source: Centers for Disease Control (pg. 16)

Table: Height in inches for Males aged 20 and over (approximate)

Sample Mean	69 inches
Sample Std. Dev.	6 inches
Sample Size	5647
My Height	73 inches

We'll return to this example later.

For Now – Single Population, Normally Distributed

$$X_1, X_2, \ldots, X_n \sim \text{iid } N(\mu, \sigma^2)$$

Later we'll look at more than one population and talk about what happens if Normality doesn't hold.



Suppose $X_1, X_2, \dots, X_n \sim \text{iid } N(\mu, \sigma^2)$. What is the sampling distribution of $\sqrt{n}(\bar{X}_n - \mu)/\sigma$?

- (a) $N(\mu, \sigma^2)$
- (b) N(0,1)
- (c) $N(0,\sigma)$
- (d) $N(\mu, 1)$
- (e) Not enough information to determine.

Z-score!

Suppose $X_1, X_2, \dots, X_n \sim \text{iid } N(\mu, \sigma^2)$. From above,

$$E[\bar{X}_n] = \mu$$
 $Var(\bar{X}_n) = \sigma^2/n$
 $\Rightarrow SD(\bar{X}_n) = \sigma/\sqrt{n}$

Thus,

$$\sqrt{n}(\bar{X}_n - \mu)/\sigma = \frac{\bar{X}_n - \mu}{\sigma/\sqrt{n}} = \frac{\bar{X}_n - E[\bar{X}_n]}{SD(\bar{X}_n)} \sim N(0, 1)$$

Remember that we call the standard deviation of a sampling distribution the standard error, written SE, so

$$\frac{X_n - \mu}{2\pi \sqrt{x}} \sim N(0, 1)$$



Suppose $X_1, X_2, ..., X_n \sim \text{iid } N(\mu, \sigma^2)$. What is the approximate value of the following?

$$P\left(-2 \le \frac{\bar{X}_n - \mu}{SE(\bar{X}_n)} \le 2\right)$$



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$$P\left(-2 \le \frac{\bar{X}_n - \mu}{SE(\bar{X}_n)} \le 2\right) \approx 0.95$$

$$P\left(-2 \le \frac{\bar{X}_n - \mu}{SE(\bar{X}_n)} \le 2\right) = 0.95$$

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$$P\left(\bar{X}_n - 2 \cdot SE \le \mu \le \bar{X}_n + 2 \cdot SE\right) = 0.95$$

Confidence Intervals

Confidence Interval (CI)

A confidence interval is a range (A, B) constructed from the sample data that has a specified probability of containing a population parameter:

$$P(A \le \theta_0 \le B) = 1 - \alpha$$

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Confidence Level

The specified probability, typically denoted $1 - \alpha$, is called the confidence level. For example, if $\alpha = 0.05$ then the confidence level is 0.95 or 95%.

Confidence Interval for Mean of Normal Population

Population Variance Known

Confidence Interval for Mean of Normal Population

The interval $\left| \overline{X}_n \pm 2\sigma/\sqrt{n} \right|$ has approximately 95% probability of containing the population mean μ , provided that:

$$X_1, X_2, \ldots, X_n \sim \text{iid } N(\mu, \sigma^2)$$

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But What Does This Mean?

Which quantities are random?



Suppose $X_1, X_2, ..., X_n \sim \text{iid } N(\mu, \sigma^2)$. Which quantities are random variables?

- (a) μ only
- (b) σ and μ
- (c) σ only
- (d) σ, μ and \bar{X}_n
- (e) \bar{X}_n only

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Confidence Interval is a Random Variable!

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- 1. X_1, \ldots, X_n are RVs $\Rightarrow \bar{X}_n$ is a RV (repeated sampling)
- 2. μ , σ and n are constants
- 3. Confidence Interval $\bar{X_n} \pm 2\sigma/\sqrt{n}$ is also a RV!

Meaning of Confidence Interval

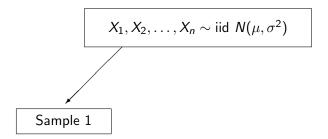
Meaning of Confidence Interval

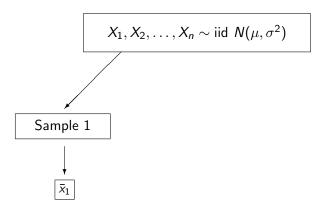
If we sampled many times we'd get many different sample means, each leading to a different confidence interval. Approximately 95% of these intervals will contain μ .

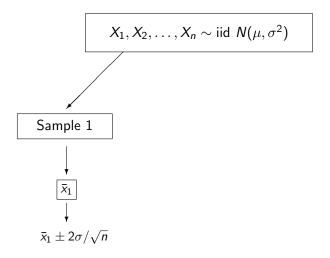
Rough Intuition

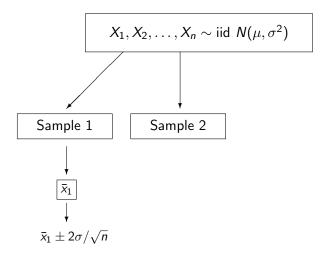
What values of μ are consistent with the data?

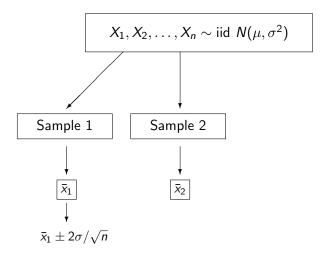
$$X_1, X_2, \ldots, X_n \sim \text{iid } N(\mu, \sigma^2)$$

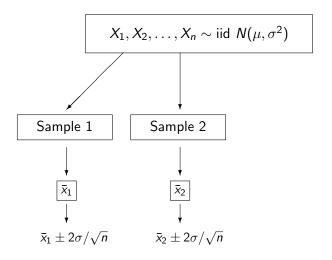


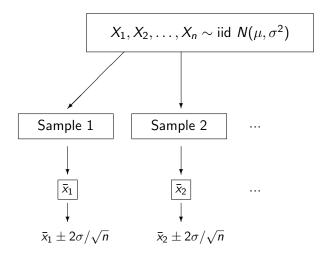


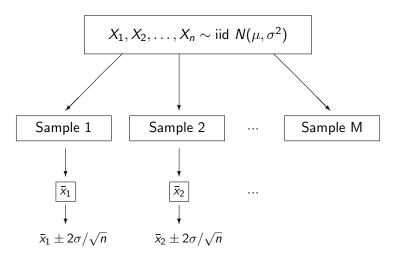


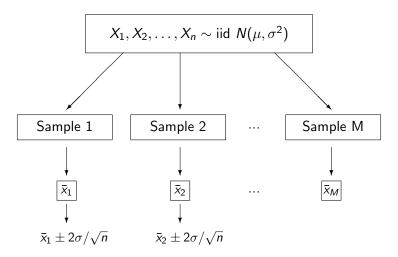


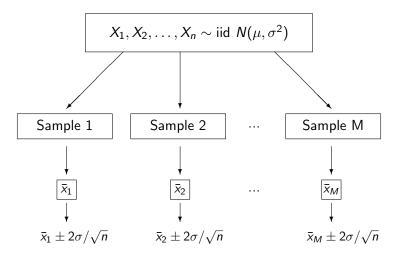


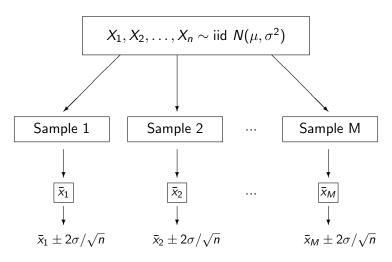




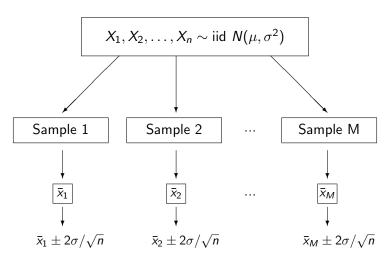








Repeat M times \rightarrow get M different intervals



Repeat M times \rightarrow get M different intervals Large M \Rightarrow Approx. 95% of these Intervals Contain μ

Simulation Example: $X_1, \ldots, X_5 \sim \text{iid } N(0, 1), M = 20$

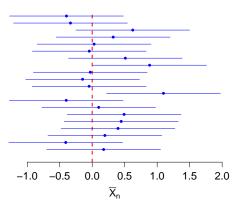


Figure: Twenty confidence intervals of the form $\bar{X}_n \pm 2\sigma/\sqrt{n}$ where $n=5,\ \sigma^2=1$ and the true population mean is 0.

Meaning of Confidence Interval for θ_0

$$P(A \le \theta_0 \le B) = 1 - \alpha$$

Each time we sample we'll get a different confidence interval, corresponding to different realizations of the random variables A and B. If we sample many times, approximately $100 \times (1-\alpha)\%$ of these intervals will contain the population parameter θ_0 .

True or False?



Suppose

$$X_1, X_2, \ldots, X_n \sim \text{iid } N(\mu, \sigma^2)$$

Then the population mean μ has approximately a 95% chance of falling in the interval $\bar{X}_n \pm 2\sigma/\sqrt{n}$.

- (a) True
- (b) False

True or False?



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$$X_1, X_2, \ldots, X_n \sim \text{iid } N(\mu, \sigma^2)$$

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- (a) True
- (b) False

FALSE! – μ is a constant!

Confidence Intervals: Some Terminology

Margin of Error

When a CI takes the form $\widehat{\theta} \pm \textit{ME}$, ME is the Margin of Error.

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Lower and Upper Confidence Limits

The lower endpoint of a CI is the lower confidence limit (LCL), while the upper endpoint is the upper confidence limit (UCL).

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Margin of Error

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The lower endpoint of a CI is the lower confidence limit (LCL), while the upper endpoint is the upper confidence limit (UCL).

Width of a Confidence Interval

The distance |UCL - LCL| is called the width of a CI. This means exactly what it says.

What is the Margin of Error



In the preceding example of a 95% confidence interval for the mean of a normal population when the population variance is known, which of these is the margin of error?

- (a) σ/\sqrt{n}
- (b) \bar{X}_n
- (c) σ
- (d) $2\sigma/\sqrt{n}$
- (e) $1/\sqrt{n}$

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- (a) σ/\sqrt{n}
- (b) $2\sigma/\sqrt{n}$
- (c) $3\sigma/\sqrt{n}$
- (d) $4\sigma/\sqrt{n}$
- (e) $5\sigma/\sqrt{n}$

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Example: Calculate the Margin of Error



 $X_1,\ldots,X_{100}\sim {
m iid}~ {\it N}(\mu,1)$ but we don't know $\mu.$ Want to create a 95% confidence interval for $\mu.$

What is the margin of error?

Example: Calculate the Margin of Error



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What is the margin of error?

The confidence interval is $\bar{X}_n \pm 2\sigma/\sqrt{n}$ so

$$ME = 2\sigma/\sqrt{n} = 2 \cdot 1/\sqrt{100} = 2/10 = 0.2$$

Example: Calculate the Lower Confidence Limit



 $X_1,\ldots,X_{100} \sim N(\mu,1)$ but we don't know μ . Want to create a 95% confidence interval for μ .

We found that ME = 0.2. The sample mean $\bar{x} = 4.9$. What is the lower confidence limit?

Example: Calculate the Lower Confidence Limit



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$$LCL = \bar{x} - ME = 4.9 - 0.2 = 4.7$$

Example: Similarly for the Upper Confidence Limit. . .

 $X_1,\ldots,X_{100} \sim \textit{N}(\mu,1)$ but we don't know $\mu.$ Want to create a 95% confidence interval for $\mu.$

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Example: Similarly for the Upper Confidence Limit...

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 but we don't know $\mu.$ Want to create a 95% confidence interval for $\mu.$

We found that ME = 0.2. The sample mean $\bar{x} = 4.9$. What is the upper confidence limit?

$$UCL = \bar{x} + ME = 4.9 + 0.2 = 5.1$$

Example: 95% CI for Normal Mean, Popn. Var. Known

 $X_1,\ldots,X_{100}\sim N(\mu,1)$ but we don't know μ .

95% CI for
$$\mu = [4.7, 5.1]$$

What values of μ are plausible?

Example: 95% CI for Normal Mean, Popn. Var. Known

$$X_1,\ldots,X_{100}\sim N(\mu,1)$$
 but we don't know μ .

95% CI for
$$\mu = [4.7, 5.1]$$

What values of μ are plausible?

The data actually came from a N(5,1) Distribution.

$$P\left(-c \le \frac{\bar{X}_n - \mu}{\sigma/\sqrt{n}} \le c\right) = 1 - \alpha$$

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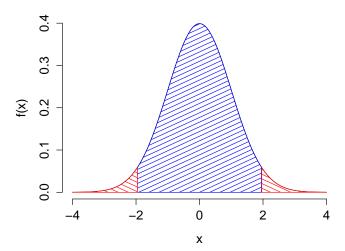
Take
$$c = qnorm(1 - \alpha/2)$$

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$$ar{X}_n \pm ext{qnorm}(1-lpha/2) imes \sigma/\sqrt{n}$$



Confidence Interval for a Normal Mean, σ Known

$$\left|ar{X}_{n} \pm \mathtt{qnorm}(1-lpha/2) imes \sigma/\sqrt{n}
ight|$$

What Affects the Margin of Error?

$$ar{X}_n \pm \mathtt{qnorm}(1-lpha/2) imes \sigma/\sqrt{n}$$

Sample Size *n*

ME decreases with n: bigger sample \implies tighter interval

Population Std. Dev. σ

ME increases with σ : more variable population \implies wider interval

Confidence Level $1-\alpha$

ME increases with $1 - \alpha$: higher conf. level \implies wider interval

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Confidence Level $1-\alpha$

ME increases with $1 - \alpha$: higher conf. level \implies wider interval

Conf. Level	90%	95%	99%
α	0.1	0.05	0.01
$\texttt{qnorm}(1-\alpha/2)$	1.64	1.96	2.56

But What if σ is Unknown?

- \blacktriangleright What we've done so far assumed that σ was known.
- In real applications this is typically not the case.

Why not try using the sample standard deviation s?

This works, but requires a small change. Instead of basing the interval on quantiles of a normal distribution, we need to use a t distribution. We'll look at this next time.