

# Lecture 24

## Confidence Intervals

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# Confidence Intervals

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# Confidence Intervals

- MLE gives us a “point estimate” of the unknown parameter.
- But  $\hat{\theta}$  probably won't *exactly* equal  $\theta$  due to sampling error.  
 $\rightarrow P(\theta = \hat{\theta}) = 0$
- Create a confidence interval to give range of reasonable values for the unknown parameter  $\theta$ .

## Example 1: Polling

Today's poll shows 58% of people favor the new bill. The margin of error is  $\pm 3\%$ .

The confidence interval for the proportion of people that favor the bill is  $[0.55, 0.61]$ .

# Confidence Interval

## Definition

A random interval  $[a, b]$  is a  $(1 - \alpha)100\%$  *confidence interval* for the parameter  $\theta$  if it contains  $\theta$  with probability  $(1 - \alpha)$

$$P(a \leq \theta \leq b) = 1 - \alpha$$

- $(1 - \alpha)$  is called the confidence level
- When you estimate an unknown parameter  $\theta$ , it should be accompanied by a confidence interval
- **Interpretation:** We are  $[(1 - \alpha)\%]$  confident that the [insert population parameter + context] is between [insert interval + units].

# Constructing Confidence Intervals

In this class, we will construct normal distribution based intervals.

Suppose we have an estimator  $\hat{\theta}$  for unknown parameter  $\theta$ .

1.  $\hat{\theta}$  is unbiased:  $E(\hat{\theta}) = \theta$
2.  $\hat{\theta}$  follows a normal distribution.

We can standardize  $\hat{\theta}$  to get

$$Z = \frac{\hat{\theta} - \theta}{SE(\hat{\theta})} \sim N(0, 1)$$

where  $SE(\hat{\theta}) = \sqrt{\text{Var}(\hat{\theta})}$  = standard deviation of  $\hat{\theta}$

## Constructing Confidence Intervals

Let  $z_{\alpha/2}$  be the  $1 - \frac{\alpha}{2}$  quantile of the standard normal distribution.

$$P\left(-z_{\alpha/2} \leq \frac{\hat{\theta} - \theta}{SE(\hat{\theta})} \leq z_{\alpha/2}\right) = 1 - \alpha$$

## Constructing Confidence Intervals

Isolating  $\theta$  in the middle, we get

$$P\left(\hat{\theta} - z_{\alpha/2}SE(\hat{\theta}) \leq \theta \leq \hat{\theta} + z_{\alpha/2}SE(\hat{\theta})\right) = 1 - \alpha$$

Thus, a  $(1 - \alpha)100\%$  confidence interval for  $\theta$  is

$$\hat{\theta} \pm z_{\alpha/2}SE(\hat{\theta})$$

Common choices for  $\alpha$  are 0.01, 0.05, and 0.1

$(1 - \alpha)100\%$	80	90	95	98	99
$z_{\alpha/2}$	1.282	1.645	1.96	2.326	2.576

## Constructing Confidence Intervals Cont.

We will make confidence intervals for four cases:

1.  $\mu$  (population mean)
2.  $p$  (population proportion)
3.  $\mu_1 - \mu_2$  (difference in population means)
4.  $p_1 - p_2$  (difference in population proportions)

Confidence intervals for all 4 of the above cases can be constructed using normal distribution based inference.

Follow the same general procedure to construct these intervals.



# Confidence Interval for Mean

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# Confidence Interval for $\mu$

## Confidence interval for the population mean

$X_1, \dots, X_n \stackrel{iid}{\sim} f_X(x)$  with  $E(X_i) = \mu$  and  $\text{Var}(X_i) = \sigma$

First, we estimate  $\mu$  using the *statistic*  $\bar{X}$ . From CLT, we know

- $\bar{X} \sim N(\mu, \frac{\sigma^2}{n})$
- $SE(\bar{X}) = \sqrt{\text{Var}(\bar{X})} = \sqrt{\frac{\sigma^2}{n}} = \frac{\sigma}{\sqrt{n}}$

A  $(1 - \alpha)100\%$  confidence interval for  $\mu$  is

$$\bar{X} \pm z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$$

In most cases, the population standard deviation  $\sigma$  will be unknown. Replace  $\sigma$  with the sample standard deviation  $s$ .

$$\bar{X} \pm z_{\alpha/2} \frac{s}{\sqrt{n}}$$

## Confidence Interval for $\mu$ Cont.

If we want a 95% confidence interval, then

$$1 - \alpha = 0.95$$

$$\rightarrow \alpha = 0.05$$

$$\rightarrow \alpha/2 = 0.025$$

$z_{\alpha/2} = z_{0.025}$  is the 0.975<sup>th</sup> quantile of the  $N(0, 1)$  distribution.

$\rightarrow$  Using the  $z$ -table, we get  $z_{0.025} = 1.96$

The 95% confidence interval for  $\mu$  is

$$\bar{X} \pm 1.96 \frac{\sigma}{\sqrt{n}} \quad \text{when } \sigma \text{ is known}$$

$$\bar{X} \pm 1.96 \frac{s}{\sqrt{n}} \quad \text{when } \sigma \text{ is unknown}$$

## Example

Example 2: A random sample of 50 batteries were taken for a particular brand. For the sample, the mean lifetime is 72.5 hours and variance is 19.3 hours<sup>2</sup>. Find a 95% confidence interval for the true mean lifetime of batteries from that particular brand.



# Selecting Sample Size for Means

## How to decide sample size?

- Can choose the sample size  $n$  to obtain a desired level of confidence & width for our confidence interval.
- Margin or error ( $\Delta$ ) is half the width of the confidence interval  
margin of error =  $\Delta = z_{\alpha/2}SE(\bar{X}) = z_{\alpha/2}\frac{\sigma}{\sqrt{n}}$
- The bigger the sample size, the smaller the standard error of the estimator, and smaller the size of our interval

To attain a particular margin of error  $\Delta$ , we need a sample size

$$n \geq \left( \frac{z_{\alpha/2}\sigma}{\Delta} \right)^2$$

# Confidence Interval for Proportion

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## Confidence interval for the population proportion

- In this scenario, we want to estimate the proportion of population belonging to a particular category.
- Any individual in the population either belongs to the category of interest (“1”), or they don’t (“0”).
- Thus, we can think of each random variable  $X$  as a Bernoulli distribution with unknown parameter  $p$
- We ultimately want to estimate and find a confidence interval for  $p$ .



## Confidence Interval for $p$ Cont.

$$X_1, \dots, X_n \stackrel{iid}{\sim} \text{Bern}(p)$$

First, estimate  $p$  using the *statistic*  $\hat{p} = \frac{\sum X_i}{n}$  = sample proportion.

- $E(\hat{p}) = E\left(\frac{\sum_{i=1}^n X_i}{n}\right) = \frac{1}{n}E(\sum_{i=1}^n X_i) = \frac{1}{n}np = p$  (unbiased)
- $\text{Var}(\hat{p}) = \text{Var}\left(\frac{\sum X_i}{n}\right) = \frac{1}{n^2} \text{Var}(\sum X_i) = \frac{np(1-p)}{n^2} = \frac{p(1-p)}{n}$   
 $\rightarrow SE(\hat{p}) = \sqrt{\text{Var}(\hat{p})} = \sqrt{\frac{p(1-p)}{n}}$

Since  $\hat{p}$  is the mean of the Bernoulli  $X$ 's, CLT for means applies

$$\hat{p} \sim N\left(p, \frac{p(1-p)}{n}\right)$$

Thus a  $(1 - \alpha)100\%$  confidence interval for  $p$  is

$$\hat{p} \pm z_{\alpha/2} \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}}$$

## Example

Example 3: In a random sample of 1000 U.S. adults, 38.8% stated they believed in the existence of ghosts. Find a 90% confidence interval for the population proportion of all U.S. adults who believe in the existence of ghosts.

# Selecting Sample Size for Proportions

## How to decide sample size?

- Just as before, we can select the sample size based on how large we want our margin or error to be

$$\text{margin or error} = \Delta = z_{\alpha/2} SE(\hat{p}) = z_{\alpha/2} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$$

- **Issue:** We haven't taken the sample, so we don't know  $\hat{p}$
- **Solution:** Use  $\hat{p} = 0.5$  for most conservative sample size

$$\Delta = z_{\alpha/2} \sqrt{\frac{0.5 \cdot 0.5}{n}} = z_{\alpha/2} \sqrt{\frac{0.5^2}{n}}$$

To attain a particular margin of error  $\Delta$ , we need a sample size

$$n \geq \left( \frac{z_{\alpha/2} \cdot 0.5}{\Delta} \right)^2$$

## Sample Size Calculation Cont.

Example 4: Political polls typically use 95% confidence and report margin of errors of 3%:  $\hat{p} \pm 0.03$ .

What sample size do we need to for such a poll?