



# Chapter 8 - Lecture 1

## Basic Properties of Confidence Intervals

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## Overview of Chapter 8

It mainly discusses confidence intervals for a population mean for the following situations:

- ❶  $N(\mu, \sigma)$  with  $\sigma$  known (8.1)
- ❷ General sample with  $\sigma$  known/unknown under large sample size (8.2)
- ❸  $N(\mu, \sigma)$  with  $\sigma$  unknown under small sample size (8.3)

Let  $n$  denote the sample size.

	$\sigma$ known	$\sigma$ unknown
Normal	8.1 $n$ does not matter	8.3 under $n$ is small/large
General	8.2 under $n$ is large	8.2 under $n$ is large



## 8.1 Case 1: $N(\mu, \sigma)$ with $\sigma$ known

- From chapter 7, we know  $\bar{X}$  is an estimator for  $\mu$ .
- To gain the precision, we can report the stand error.
- What's the confidence level? – Use **confidence intervals (CI)** for the parameters.
- CI gives an estimated range of values which is likely to include an unknown population parameter



## Definitions

A confidence interval is always calculated by first selecting a *confidence level*, which is a measure of the degree of reliability of the interval.

- The probability we allow ourselves to be wrong when we are estimating a parameter with a confidence interval, is called **significance/critical level** and is denoted with  $\alpha$ .
- So if  $\alpha = 0.05$  then  $1 - 0.05 = 0.95$ , so we call our confidence interval a **95% confidence interval**



## General strategy to construct CI for $\theta$

- Parameter  $\theta$
- Random sample  $X_1, \dots, X_n$
- Need a random variable  $Y$  such that
  - 1 Its functional form depends on  $X_1, \dots, X_n$  and  $\theta$ . Hence we can denote  $Y = h(X_1, \dots, X_n, \theta)$ .
  - 2 Its distribution does not depend on  $\theta$ .



## General strategy to construct CI for $\theta$

If you can find such  $Y$ , then based on the distribution of  $Y$ , you should be able to write down

$$P(a < h(X_1, \dots, X_n, \theta) < b) = 1 - \alpha$$

Both  $a$  and  $b$  do not depend on  $\theta$ . By transformation, you can derive :

$$P(l(X_1, \dots, X_n) < \theta < u(X_1, \dots, X_n)) = 1 - \alpha$$

Here,  $l(X_1, \dots, X_n) < \theta < u(X_1, \dots, X_n)$  is a random interval. We can plug in the observations to get the  $1 - \alpha$  confidence interval

$$l(x_1, \dots, x_n) < \theta < u(x_1, \dots, x_n)$$



## Now let's look at how to apply this general strategy

Assume that you have a random sample  $X_1, \dots, X_n \sim N(\mu, \sigma^2)$

- $\mu$  unknown .
- $\sigma$  known.

*Goal:* To find confidence intervals for the mean  $\mu$ .



## Construct 95% confidence interval

- From Chapter 6, we know that  $\bar{X} \sim N(\mu, (\sigma/\sqrt{n})^2)$

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

- We also know that the area under the standard normal curve between  $-1.96$  and  $1.96$  is  $0.95$

$$\Rightarrow P(-1.96 \leq \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \leq 1.96) = 0.95$$

- Manipulate to get the equivalent form  $l < \mu < u$ .

$$P\left(\bar{X} - 1.96 \frac{\sigma}{\sqrt{n}} < \mu < \bar{X} + 1.96 \frac{\sigma}{\sqrt{n}}\right) = 0.95$$





## 95% confidence interval

- Until now, we have a random interval  $\left(\bar{X} - 1.96 \frac{\sigma}{\sqrt{n}}, \bar{X} + 1.96 \frac{\sigma}{\sqrt{n}}\right)$
- After observing  $X_1 = x_1, \dots, X_n = x_n$ , we compute the observed sample mean  $\bar{x}$  and then substitute into the above random interval in place of  $\bar{X}$ , the resulting fixed interval is called **95% confidence interval for  $\mu$** .

### Definition

For  $X_1, \dots, X_n$  random sample from  $N(\mu, \sigma)$ , with  $\sigma$  known, the 95% confidence interval for  $\mu$  can be expressed as

$\left(\bar{x} - 1.96 \frac{\sigma}{\sqrt{n}}, \bar{x} + 1.96 \frac{\sigma}{\sqrt{n}}\right)$  is a 95% CI for  $\mu$   
or as  $\bar{x} - 1.96 \frac{\sigma}{\sqrt{n}} < \mu < \bar{x} + 1.96 \frac{\sigma}{\sqrt{n}}$  with 95% confidence.



Example 8.2 Assume the preferred height is normally distributed as  $N(\mu, \sigma = 2)$ . A random sample of size  $n = 31$  is collected. The sample mean is  $\bar{x} = 80$ . What is the resulting 95% confidence interval ?

$$\begin{aligned}\bar{x} \pm 1.96 \frac{\sigma}{\sqrt{n}} &= 80 \pm 1.96 \frac{2}{\sqrt{31}} \\ &= 80 \pm .7 \\ &= (79.3, 80.7)\end{aligned}$$



# Interpretation

How do we interpret a 95% Confidence Interval?



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**For random interval:**

With probability 0.95, the random interval

$(\bar{X}_n - 1.96\sigma/\sqrt{n}, \bar{X}_n + 1.96\sigma/\sqrt{n})$  will cover the true value  $\mu$ .



## Interpretation

How do we interpret a 95% Confidence Interval?

Attention!!

### **For random interval:**

With probability 0.95, the random interval  $(\bar{X}_n - 1.96\sigma/\sqrt{n}, \bar{X}_n + 1.96\sigma/\sqrt{n})$  will cover the true value  $\mu$ .

### **For confidence interval:**

If the experiment is taken (the random sample is drawn) independently over and over again, about 95% of the intervals derived from this formula will cover the true value  $\mu$ .



## Interpretation

How do we interpret a 95% Confidence Interval?

Attention!!

### **For random interval:**

With probability 0.95, the random interval  $(\bar{X}_n - 1.96\sigma/\sqrt{n}, \bar{X}_n + 1.96\sigma/\sqrt{n})$  will cover the true value  $\mu$ .

### **For confidence interval:**

If the experiment is taken (the random sample is drawn) independently over and over again, about 95% of the intervals derived from this formula will cover the true value  $\mu$ .

*Misspecified interpretation:* with probability 0.95,  $\mu$  will take value in the confidence interval.



## Confidence interval of other confidence levels

A **100(1 -  $\alpha$ )% confidence interval** for the mean  $\mu$  of a normal population when the value of  $\sigma$  is known is given by

$$\left( \bar{x} - z_{\alpha/2} \frac{\sigma}{\sqrt{n}}, \bar{x} + z_{\alpha/2} \frac{\sigma}{\sqrt{n}} \right)$$

where  $P(-z_{\alpha/2} < Z < z_{\alpha/2}) = 1 - \alpha$





## Confidence interval of other confidence levels

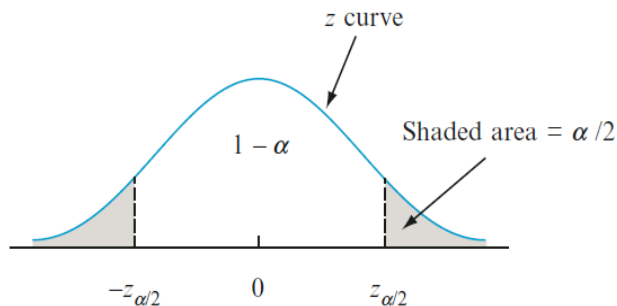


Figure 8.4  $P(-z_{\alpha/2} \leq Z \leq z_{\alpha/2}) = 1 - \alpha$

Figure: Modern Mathematical Statistics with Applications, 2rd, P387



## Confidence interval of other confidence levels

Example 8.2 (revised) Assume the preferred height is normally distributed as  $N(\mu, \sigma = 2)$ . A random sample of size  $n = 31$  is collected. The sample mean is  $\bar{x} = 80$ . What is the resulting 90% confidence interval ?

For 90% confidence interval, we need to find  $z_{0.05}$ . By checking table or use R, we can get  $z_{0.05} = 1.645$ .

$$\begin{aligned}\bar{x} \pm 1.645 \frac{\sigma}{\sqrt{n}} &= 80 \pm 1.645 \frac{2}{\sqrt{31}} \\ &= 80 \pm .59 \\ &= (79.41, 80.59)\end{aligned}$$



- The width of confidence interval

$$w = 2z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$$

- Larger confidence levels means more reliable. However, for a given sample size  $n$ , the width  $w$  is increasing with  $z_{\alpha/2}$ .
- Therefore, the gain in reliability entails a loss in precision.
- Solution: calculate the sample size.
- Given the width  $w_0$ , the sample size that ensures  $w_0$  is

$$n = \left( 2z_{\alpha/2} \frac{\sigma}{w_0} \right)^2$$



## Sample Size calculation

Example: We want to get a 95% confidence interval for the mean  $\mu$  of a random sample coming from  $N(\mu, \text{var} = 4)$ . Find the sample size that we need in order to get interval with width 4.

$$\begin{aligned}n &= \left(2z_{\alpha/2} \frac{\sigma}{w}\right)^2 \\&= \left(2z_{0.025} \frac{2}{4}\right)^2 \\&= z_{0.025}^2 \\&= 1.96^2 \\&= 3.84\end{aligned}$$