# Chapter 8 - Lecture 2(1) Large-Sample Confidence Intervals for a Population Mean

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- **1** Review:  $N(\mu, \sigma^2)$  with known  $\sigma^2$
- **2** General population with known  $\sigma^2$
- **3** General population with unknown  $\sigma^2$
- 4 One-Sided Confidence Intervals (Confidence Bounds)

# Review

• A Random sample  $X_1, \ldots, X_n \sim N(\mu, \sigma^2)$ ,  $\sigma^2$  is known.

Two key assumptions:

- 1 Normality
- **2**  $\sigma^2$  is known.
- We have seen this last lecture: the  $(1-\alpha)\%$  Confidence Interval for  $\mu$  is

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

Think what we can do if dropping normality assumption, but still assume  $\sigma^2$  is known.

In this case, suppose we want to give an approximate 95% CI of the population mean  $\mu$  based on a large sample (n > 30).

# Solution:

CLT 
$$\sqrt{n}(\bar{X}-\mu)/\sigma \approx N(0,1)$$
;

$$P(-1.96\sigma/\sqrt{n} < \bar{X} - \mu < 1.96\sigma/\sqrt{n}) \approx 0.95$$

Approximate 95% CI for  $\mu$  is  $(\bar{x} - 1.96\sigma/\sqrt{n}, \bar{x} + 1.96\sigma/\sqrt{n})$ .

Example 1. The sweetness of the apples in a farm has mean  $\mu$  and variance  $0.2^2$ . To estimate  $\mu$ , 100 apples are selected randomly and the sweetness of each of them is measured. Suppose the average sweetness is 0.5.

- **1** Please give a point estimate of  $\mu$ ;
- **2** Please give an approximate 95% confidence interval of  $\mu$ .

### Solution:

- **1** A sensible point estimate of  $\mu$  is  $\bar{x} = 0.5$ .
- **2** An approximate 95% confidence interval of  $\mu$  is

$$(\bar{x} - 1.96\sigma/\sqrt{n}, \bar{x} + 1.96\sigma/\sqrt{n})$$
=  $(0.5 - 1.96 \times 0.2/\sqrt{100}, 0.5 + 1.96 \times 0.2/\sqrt{100})$   
=  $(0.4608, 0.5392)$ 

Be aware of the difference:

- **1** only works when n > 30;
- **2** The  $100(1-\alpha)\%$  CI is approximate, so that the actual confidence is not exactly  $100(1-\alpha)$ .

Example 2. With the new drug, the time to recovery from a disease has the population mean  $\mu$  and variance  $3.5^2$ . Suppose 49 patients received the new drug and the sample average of recovery time is 5. Based on these data please give an approximate 90% CI for  $\mu$ .

**Solution**: For the two sided CI,  $z_{\frac{\alpha}{2}} = 1.645$ .

An approximate 95% confidence interval of  $\mu$  is

$$(\bar{x} - 1.645\sigma/\sqrt{n}, \bar{x} + 1.645\sigma/\sqrt{n})$$
=  $(5 - 1.645 \times 3.5/\sqrt{49}, 5 + 1.645 \times 3.5/\sqrt{49})$   
=  $(4.1775, 5.8225)$ 

### So far:

• When the random data is normally distributed with known variance  $\sigma^2$ , the exact  $100(1-\alpha)\%$  CI for the mean  $\mu$  is:

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

• Without normality assumption, we use CLT to get an approximate CI for  $\mu$  when the sample size is large and  $\sigma^2$  is known. The approximate  $100(1-\alpha)\%$  CI for the mean  $\mu$  is:

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

How about if we also drop the assumptions of known  $\sigma^2$  ?

We know when n is large, we have CLT and consistency of  $S^2$ .

$$\sqrt{n}(\bar{X}-\mu)/\sigma \approx N(0,1)$$

$$S^2 \approx \sigma^2$$

$$ightarrow \sqrt{n}(ar{X}-\mu)/S pprox N(0,1)$$

.

The result is: without normality assumption, when n is large and  $\sigma^2$  is unknown, we still get approximate Z CI for the population mean  $\mu$ .

An approximate  $100(1-\alpha)\%$  confidence interval for  $\mu$  is:

$$(\bar{x}-z_{\alpha/2}s/\sqrt{n},\bar{x}+z_{\alpha/2}s/\sqrt{n})$$

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# Exercises 8.12

A random sample of 110 lighting flashes in a region resulted in a sample average radar echo duration of 0.81 second and a sample standard deviation 0.34 second. Calculate a 99% (two-sided) CI for the true average echo duration  $\mu$ .

# Solution:

For the two sided CI,  $z_{\frac{\alpha}{2}} = z_{\frac{1-0.99}{2}} = 2.575829$ .

An approximate 99% confidence interval of  $\mu$  is

$$(\bar{x} - 2.575829s/\sqrt{n}, \bar{x} + 2.575829s/\sqrt{n})$$
  
=  $(0.81 - 2.575829 \times 0.34/\sqrt{110}, 0.81 + 2.575829 \times 0.34/\sqrt{110})$   
=  $(0.726, 0.894)$ 



The confidence intervals discussed thus far give both a lower confidence bound and an upper confidence bound.

- Bound on the error of estimation associated with  $(1-\alpha)$  CI is the half-width  $z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$  or  $z_{\alpha/2} \frac{s}{\sqrt{n}}$
- $\alpha/2$  in the formula is due to two sided.

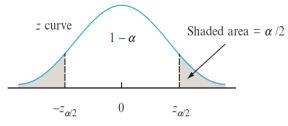


Figure 8.4  $P(-z_{a/2} \le Z \le z_{a/2}) = 1-\alpha$ 

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

In two-sided CI, depends on the conditions, we have

- upper bound  $\bar{x} + z_{\alpha/2} s / \sqrt{n}$ ;  $\bar{x} + z_{\alpha/2} \sigma / \sqrt{n}$
- lower bound  $\bar{x}-z_{\alpha/2}s/\sqrt{n};~\bar{x}-z_{\alpha/2}\sigma/\sqrt{n}$

However, for one-sided CI, we only need one of these two types of bound. Thus we assign all the  $\alpha$  to one side. For example, an approximate 95% one-sided CI  $\mu$  is

$$P\left(\frac{\bar{X}-\mu}{S/\sqrt{n}} < 1.645\right) \approx 0.95$$

# **Proposition**

A large-sample upper confidence bound for  $\mu$  is

$$\mu < \bar{x} + z_{\alpha} \frac{s}{\sqrt{n}}$$
 or written as  $(-\infty, \bar{x} + z_{\alpha} \frac{s}{\sqrt{n}})$ 

A large-sample lower confidence bound for  $\mu$  is

$$\mu > \bar{x} - z_{\alpha} \frac{s}{\sqrt{n}}$$
 or written as  $(\bar{x} - z_{\alpha} \frac{s}{\sqrt{n}}, \infty)$ 

# Example 8.10

A random sample of 50 patients who had been seen at an outpatient clinic was selected, and the waiting time to see a physician was determined for each one, resulting in a sample mean time of 40.3 min and a sample standard deviation of 28 min. What is the upper confidence bound for true average waiting time with a confidence level roughly 95% ?

### Solution:

- n = 50
- $\bar{x} = 40.3$
- *s* = 28
- $z_{0.05} = 1.645$
- upper confidence bound is  $\bar{x} + z_{\alpha} \frac{s}{\sqrt{n}} = 40.3 + 1.645 \times 28/\sqrt{50} = 46.8$