

One- and Two-Sample Estimation Problems

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9.4 Single Sample: Estimating the Mean.

- Confidence Interval on  $\mu$ ,  $\sigma^2$  Known.

If  $\bar{x}$  is the mean of a random sample of size  $n$  from a population with known variance  $\sigma^2$ , a  $100(1-\alpha)\%$  confidence interval for  $\mu$  is given by

$$\bar{x} - z_{\alpha/2} \frac{\sigma}{\sqrt{n}} < \mu < \bar{x} + z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$$

where  $z_{\alpha/2}$  is the  $z$ -value leaving an area of  $\alpha/2$  to the right

Theorem 9.1 If  $\bar{x}$  is used as an estimate of  $\mu$ , we can be  $100(1-\alpha)\%$  confident that the error will not exceed  $z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$ .

Theorem 9.2 If  $\bar{x}$  is used as an estimate of  $\mu$ , we can be  $100(1-\alpha)\%$  confident that the error will not exceed a specified amount  $e$  when the sample size is

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

## - One-Sided Confidence Bounds.

- One-sided Confidence Bounds on  $\mu$ ,  $\sigma^2$  Known.

If  $\bar{X}$  is the mean of a random variable of size  $n$  from a population with variance  $\sigma^2$ , the one-sided  $100(1-\alpha)\%$  confidence bounds for  $\mu$  are given by:

$$\text{upper one-sided bound: } \bar{x} + z_{\alpha} \sigma / \sqrt{n}$$

$$\text{lower one-sided bound: } \bar{x} - z_{\alpha} \sigma / \sqrt{n}.$$

## - The Case of $\sigma$ Unknown.

- Confidence Interval on  $\mu$ ,  $\sigma^2$  Unknown.

If  $\bar{x}$  and  $s$  are the mean and standard deviation of a random sample from a normal population with unknown variance  $\sigma^2$ , a  $100(1-\alpha)\%$  confidence interval for  $\mu$  is

$$\bar{x} - t_{\alpha/2} \frac{s}{\sqrt{n}} < \mu < \bar{x} + t_{\alpha/2} \frac{s}{\sqrt{n}},$$

where  $t_{\alpha/2}$  is the  $t$ -value with  $v = n - 1$  degrees of freedom, leaving an area of  $\alpha/2$  to the right.

## - Concept of a Large-Sample Confidence Interval

- Often statisticians recommend that even when normality cannot be assumed,  $\sigma$  is unknown, and  $n \geq 30$ ,  $s$  can be replaced  $\sigma$  and the confidence interval

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

may be used. This is often referred to as a large-sample confidence interval.

## 9.5 Standard Error of a Point Estimate.

- Confidence Limits on  $\mu$ ,  $\sigma^2$  Unknown.

$$\bar{x} \pm t_{\alpha/2} \frac{s}{\sqrt{n}} = \bar{x} \pm t_{\alpha/2} \text{s.e.}(\bar{x}).$$

## 9.6 Prediction Intervals.

- Prediction Interval of a Future Observation,  $\sigma^2$  Known.  
For a normal distribution of measurements with unknown mean  $\mu$  and known variance  $\sigma^2$ , a  $100(1-\alpha)\%$  prediction interval of a future observation  $x_0$  is

$$\bar{x} - z_{\alpha/2} \sigma \sqrt{1 + \frac{1}{n}} < x_0 < \bar{x} + z_{\alpha/2} \sigma \sqrt{1 + \frac{1}{n}}$$

Where  $z_{\alpha/2}$  is the  $z$ -value leaving an area of  $\alpha/2$  to the right.

- Prediction Interval of a Future Observation,  $\sigma^2$  Unknown.

For a normal distribution of measurements with unknown mean  $\mu$  and unknown variance  $\sigma^2$ , a  $100(1-\alpha)\%$  prediction interval of a future observation  $x_0$  is

$$\bar{x} - t_{\alpha/2} s \sqrt{1 + \frac{1}{n}} < x_0 < \bar{x} + t_{\alpha/2} s \sqrt{1 + \frac{1}{n}}$$

where  $t_{\alpha/2}$  is the  $t$ -value with  $v = n - 1$  degrees of freedom, leaving an area of  $\alpha/2$  to the right

- = Tolerance Limit. Unknown  $\mu$ , Unknown  $\sigma$

For a normal distribution of measurements with unknown mean  $\mu$  and unknown standard deviation  $\sigma$ , tolerance limits are given by  $\bar{x} \pm ks$ , where  $k$  is determined such that one can assert with  $100(1-\gamma)\%$  confidence that the given limits contain

at least the proportion  $1-\alpha$  of the measurements.