

$$\text{Sample Mean } \bar{X} = \frac{\sum x}{n}$$

## Confidence Interval Estimation

### I. One Sample Problem:

$$\sigma \text{ known: } \bar{x} \pm z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$$

$$\text{sample size: } n = \left( \frac{z_{\alpha/2} \sigma}{E} \right)^2$$

$\sigma$  unknown,

$$\text{small sample: } \bar{x} \pm t_{\alpha/2, n-1} \frac{s}{\sqrt{n}}$$

$$\text{large sample: } \bar{x} \pm z_{\alpha/2} \frac{s}{\sqrt{n}}$$

### Large Sample Confidence Interval Estimation of $p$ , a population proportion

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

$$\text{Sample Size: } n = \frac{z_{\alpha/2}^2 p(1-p)}{E^2}$$

$$\text{Maximum Sample Size: } \frac{z_{\alpha/2}^2}{4E^2}$$

### II. Two Sample Problem:

If  $\sigma_1$  and  $\sigma_2$  are known:

$$(\bar{x}_1 - \bar{x}_2) \pm z_{\alpha/2} \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$$

If  $\sigma_1$  and  $\sigma_2$  unknown, Large samples

$$(\bar{x}_1 - \bar{x}_2) \pm z_{\alpha/2} \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

$\sigma_1$  and  $\sigma_2$  unknown,  $\sigma_1^2 = \sigma_2^2$  small samples

$$(\bar{x}_1 - \bar{x}_2) \pm t_{\alpha/2, f} s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}},$$

$$s_p^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}, f = n_1 + n_2 - 2.$$

$$\text{Sample Variance } s^2 = \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n-1}$$

### Grouped Data:

$$x_i^* = \text{mid point of interval } i$$

$$f_i = \text{Frequency of interval } i$$

$$\text{Sample Mean } \frac{\sum x_i^* f_i}{\sum f_i}$$

$$\text{Sample Variance } \frac{\sum x_i^{*2} f_i - (\sum x_i^* f_i)^2 / n}{n-1}$$

### Random Variables:

Binomial

$$P(X = x) = \binom{n}{x} p^x (1-p)^{n-x}, \quad x = 0, 1, \dots, n$$

Hypergeometric

$$P(X = x) = \frac{\binom{a}{x} \binom{N-a}{n-x}}{\binom{N}{n}}, \quad x = 0, 1, \dots, \min(n, a)$$

$$\text{Geometric } P(X = x) = (1-p)^{x-1} p, \quad x = 1, 2, \dots$$

$$\text{Poisson } P(X = x) = \frac{e^{-\lambda} \lambda^x}{x!}, \quad x = 0, 1, \dots$$

$$\text{Uniform } f(x) = \begin{cases} \frac{1}{\beta-\alpha}, & \alpha < x < \beta \\ 0 & \text{elsewhere} \end{cases}$$

$$\text{Exponential } f(x) = \begin{cases} \frac{1}{\beta} e^{-x/\beta}, & x > 0 \\ 0 & \text{elsewhere} \end{cases}$$

## Hypothesis Testing

### I. One Sample Problem:

$$\sigma^2 \text{ known, Test Statistic } z = \frac{\bar{x} - \mu_0}{\sigma / \sqrt{n}}$$

$\sigma^2$  unknown, small sample,

$$\text{Test Statistic } t = \frac{\bar{x} - \mu_0}{s / \sqrt{n}}$$

$\sigma^2$  unknown, large sample

$$\text{Test Statistic } z = \frac{\bar{x} - \mu_0}{s / \sqrt{n}}$$

### A population proportion, large sample Test

$$z = \frac{p - p_0}{\sqrt{p_0(1-p_0)/n}}$$

### II. Two Sample Problem

$\sigma_1^2$  and  $\sigma_2^2$  known

$$\text{Test Statistic } z = \frac{\bar{x}_1 - \bar{x}_2 - d_0}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$$

$\sigma_1^2$  and  $\sigma_2^2$  unknown, large samples

$$\text{Test Statistic } z = \frac{\bar{x}_1 - \bar{x}_2 - d_0}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

$\sigma_1^2$  and  $\sigma_2^2$  unknown,  $\sigma_1^2 = \sigma_2^2$  small samples

$$\text{Test Statistic } z = \frac{\bar{x}_1 - \bar{x}_2 - d_0}{s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

## Simple linear Regression

### Estimated regression model:

$$\hat{y} = a + bx, \text{ where: } b = \frac{S_{xy}}{S_{xx}}, \quad a = \bar{y} - b \bar{x}$$

$$S_{xx} = \sum (x_i - \bar{x})^2 = \sum x_i^2 - n(\bar{x})^2$$

$$S_{yy} = \sum (y_i - \bar{y})^2 = \sum y_i^2 - n(\bar{y})^2$$

$$S_{xy} = \sum (x_i - \bar{x})(y_i - \bar{y}) = \sum x_i y_i - n \bar{x} \bar{y}$$

$$SSE = \sum (y_i - \hat{y}_i)^2 = S_{yy} - b S_{xy}$$

$$SST = \sum (y_i - \bar{y})^2$$

$$s_e^2 \equiv MSE = \frac{SSE}{n-2}$$

### Inference about the regression coefficients

$$\text{C.I for } \beta: \quad b \pm t_{\alpha/2, n-2} \frac{s_e}{\sqrt{S_{xx}}}$$

$$\text{C.I. for } \alpha: \quad a \pm t_{\alpha/2, n-2} \frac{s_e \sqrt{\sum x^2}}{\sqrt{n} S_{xx}}$$

$$\text{Testing } \beta: \quad t = \frac{b - \beta_0}{s_e / \sqrt{S_{xx}}}$$

$$\text{Testing } \alpha: \quad t = \frac{a - \alpha_0}{s_e \sqrt{\frac{\sum x^2}{n S_{xx}}}}$$

### C. I for the mean response at $x_0$ : $\mu_{Y|x_0}$

$$\hat{y}(x_0) \pm t_{\alpha/2, n-2} s_e \sqrt{\frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}}}$$

### P.I for a future response at $x_0$ is:

$$\hat{y}(x_0) \pm t_{\alpha/2, n-2} s_e \sqrt{1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}}}$$

### Coefficient of determination

$$R^2 = 1 - \frac{SSE}{SST}$$

### Correlation coefficient

$$r = b \sqrt{\frac{S_{xx}}{S_{yy}}} = \frac{S_{xy}}{\sqrt{S_{xx} S_{yy}}}, \quad r^2 = R^2$$

### Testing the correlation coefficient $\rho = 0$

$$\text{Test Statistic } Z = \sqrt{n-3} \ln \left( \frac{1+r}{1-r} \right)$$