

Confidence Interval Estimation

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

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

Grouped Data:

x_i^* = mid point of interval i

f_i = Frequency of interval i

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

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

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

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

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

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

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

I. One Sample Problem:

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

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

σ unknown, small sample:

$\bar{x} \pm t_{\alpha/2, f} \frac{s}{\sqrt{n}}, \quad f = n - 1$

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

Large Sample Confidence Interval Estimation of p , a population proportion

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

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

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

II. Two Sample Problem:

If σ_1 and σ_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 σ_1 and σ_2 are unknown:

Large samples

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

Small Samples:

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

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

Hypothesis Testing

I. One Sample Problem:

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

σ^2 unknown, small sample,

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

σ^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

σ_1^2 and σ_2^2 known

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

σ_1^2 and σ_2^2 unknown, large samples

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

σ_1^2 and σ_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:

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

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

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

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

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

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

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

Inference about the regression coefficients

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

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

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

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

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

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

P.I for a future response at x_0 is:

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

Coefficient of determination

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

Correlation coefficient

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