Point Estimator of the Difference Between Two Population Means

$$\bar{x}_1 - \bar{x}_2$$

Standard Error of $\bar{x}_1 - \bar{x}_2$

$$\sigma_{\bar{x}_1 - \bar{x}_2} = \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$$

Interval Estimate of the Difference Between Two Population Means: σ_1 and σ_2 Known

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

Test Statistic for Hypothesis Tests About $\mu_1 - \mu_2$: σ_1 and σ_2 Known

$$z = \frac{(\bar{x}_1 - \bar{x}_2) - D_0}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$$

Interval Estimate of the Difference Between Two Population Means: σ_1 and σ_2 Unknown

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

Test Statistic for Hypothesis Tests About $\mu_1 - \mu_2$: σ_1 and σ_2 Unknown

$$t = \frac{(\bar{x}_1 - \bar{x}_2) - D_0}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

Test Statistic for Hypothesis Tests Involving Matched Samples

$$t = \frac{\bar{d} - \mu_d}{s_d / \sqrt{n}}$$

Point Estimator of the Difference Between Two Population Proportions

$$\bar{p}_1 - \bar{p}_2$$

Standard Error of $\bar{p}_1 - \bar{p}_2$

$$\sigma_{\bar{p}_1 - \bar{p}_2} = \sqrt{\frac{p_1(1 - p_1)}{n_1} + \frac{p_2(1 - p_2)}{n_2}}$$

Interval Estimate of the Difference Between Two Population Proportions

$$\bar{p}_1 - \bar{p}_2 \pm z_{\alpha/2} \sqrt{\frac{\bar{p}_1(1-\bar{p}_1)}{n_1} + \frac{\bar{p}_2(1-\bar{p}_2)}{n_2}}$$

Standard Error of $\bar{p}_1 - \bar{p}_2$ when $p_1 = p_2 = p$

$$\sigma_{\bar{p}_1 - \bar{p}_2} = \sqrt{p(1-p)\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}$$

Pooled Estimator of p when $p_1 = p_2 = p$

$$\bar{p} = \frac{n_1 \bar{p}_1 + n_2 \bar{p}_2}{n_1 + n_2}$$

Test Statistic for Hypothesis Tests About $p_1 - p_2$

$$z = \frac{(\bar{p}_1 - \bar{p}_2)}{\sqrt{\bar{p}(1 - \bar{p})\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}}$$

Interval Estimate of a Population Variance

$$\frac{(n-1)s^2}{\chi_{a/2}^2} \le \sigma^2 \le \frac{(n-1)s^2}{\chi_{(1-a/2)}^2}$$

Test Statistic for Hypothesis Tests About a Population Variance

$$\chi^2 = \frac{(n-1)s^2}{\sigma_0^2}$$

Test Statistic for Hypothesis Tests About Population Variances with $\sigma_1^2 = \sigma_2^2$

$$F = \frac{s_1^2}{s_2^2}$$

Test Statistic for Goodness of Fit

$$\chi^{2} = \sum_{i=1}^{k} \frac{(f_{i} - e_{i})^{2}}{e_{i}}$$

Expected Frequencies for Contingency Tables Under the Assumption of Independence

$$e_{ij} = \frac{(\text{Row } i \text{ Total})(\text{Column } j \text{ Total})}{\text{Sample Size}}$$

Test Statistic for Independence

$$\chi^{2} = \sum_{i} \sum_{j} \frac{(f_{ij} - e_{ij})^{2}}{e_{ij}}$$



Overall Sample Mean

$$\bar{\bar{x}} = \frac{\sum_{j=1}^k \sum_{i=1}^{n_j} x_{ij}}{n_T}$$

$$n_T = n_1 + n_2 + \dots + n_k$$

Mean Square Due to Treatments

$$MSTR = \frac{SSTR}{k - 1}$$

Sum of Squares Due to Treatments

$$SSTR = \sum_{j=1}^{k} n_j (\bar{x}_j - \bar{\bar{x}})^2$$

Mean Square Due to Error

$$MSE = \frac{SSE}{n_T - k}$$

Sum of Squares Due to Error

$$SSE = \sum_{j=1}^{k} (n_j - 1)s_j^2$$

Test Statistic for the Equality of k Population Means

$$F = \frac{MSTR}{MSE}$$

Total Sum of Squares

SST =
$$\sum_{j=1}^{k} \sum_{i=1}^{n_j} (x_{ij} - \bar{\bar{x}})^2$$

Partitioning of Sum of Squares

$$SST = SSTR + SSE$$

TWO WAY ANOVA

Total Sum of Squares

SST =
$$\sum_{i=1}^{b} \sum_{j=1}^{k} (x_{ij} - \bar{x})^2$$

Sum of Squares Due to Treatments

SSTR =
$$b \sum_{j=1}^{k} (\bar{x}_{.j} - \bar{\bar{x}})^2$$

Sum of Squares Due to Blocks

SSBL =
$$k \sum_{i=1}^{b} (\bar{x}_{i} - \bar{x})^{2}$$

Sum of Squares Due to Error

$$SSE = SST - SSTR - SSBL$$

Fisher's LSD

Test Statistic for Fisher's LSD Procedure

$$t = \frac{\bar{x}_i - \bar{x}_j}{\sqrt{\text{MSE}\Big(\frac{1}{n_i} + \frac{1}{n_j}\Big)}}$$

Fisher's LSD

$$LSD = t_{\alpha/2} \sqrt{MSE\left(\frac{1}{n_i} + \frac{1}{n_j}\right)}$$

Linear Regression:

Simple Linear Regression Model

$$y = \beta_0 + \beta_1 x + \epsilon$$

Simple Linear Regression Equation

$$E(y) = \beta_0 + \beta_1 x$$

Estimated Simple Linear Regression Equation

$$\hat{y} = b_0 + b_1 x$$

Least Squares Criterion

$$\min \Sigma (y_i - \hat{y}_i)^2$$

Slope and y-Intercept for the Estimated Regression Equation

$$b_{1} = \frac{\sum (x_{i} - \bar{x})(y_{i} - \bar{y})}{\sum (x_{i} - \bar{x})^{2}}$$
$$b_{0} = \bar{y} - b_{1}\bar{x}$$

Sum of Squares Due to Error

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

Total Sum of Squares

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

Sum of Squares Due to Regression

$$SSR = \Sigma (\hat{y}_i - \bar{y})^2$$

Relationship Among SST, SSR, and SSE

$$SST = SSR + SSE$$

Coefficient of Determination

$$r^2 = \frac{\text{SSR}}{\text{SST}}$$

Sample Correlation Coefficient

$$r_{xy} = (\text{sign of } b_1) \sqrt{\text{Coefficient of determination}}$$

= $(\text{sign of } b_1) \sqrt{r^2}$

Mean Square Error (Estimate of σ^2)

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

Standard Error of the Estimate

$$s = \sqrt{\text{MSE}} = \sqrt{\frac{\text{SSE}}{n-2}}$$

Standard Deviation of b_1

$$\sigma_{b_1} = \frac{\sigma}{\sqrt{\sum (x_i - \bar{x})^2}}$$

Estimated Standard Deviation of b_1

$$s_{b_1} = \frac{s}{\sqrt{\sum (x_i - \bar{x})^2}}$$

t Test Statistic

$$t = \frac{b_1}{s_{b_1}}$$

Mean Square Regression

$$MSR = \frac{SSR}{Number of independent variables}$$

F Test Statistic

$$F = \frac{MSR}{MSE}$$

Estimated Standard Deviation of \hat{y}_p

$$s_{\hat{y}_p} = s \sqrt{\frac{1}{n} + \frac{(x_p - \bar{x})^2}{\sum (x_i - \bar{x})^2}}$$

Confidence Interval for $E(y_p)$

$$\hat{y}_{p} \pm t_{\alpha/2} s_{\hat{y}_{p}}$$

Estimated Standard Deviation of an Individual Value

$$s_{\text{ind}} = s \sqrt{1 + \frac{1}{n} + \frac{(x_p - \bar{x})^2}{\sum (x_i - \bar{x})^2}}$$