

Formula Sheet

Measures of Center/Location:

- Sample Mean:

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$$

- Median: the center value that divides the numerically ordered data collection in two halves.
- Percentiles: The p^{th} percentile in a collection of ordered data is a value that divides the data set into two parts. The lower segment contains at least $p\%$ and the upper segment contains at least $(100 - p)\%$ of the data.
- Quartiles: Quartiles are a special case of the percentiles where the first quartile, Q_1 , has $p = 25$ and the third quartile, Q_3 , has $p = 75$.

Measures of Spread:

- Range: The range of the data is the difference between the maximum and minimum value in the data set.
- Interquartile Range: The interquartile range of the data is the difference between the third and first quartile.

$$\text{IQR} = Q_3 - Q_1$$

- Sample Variance:

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

- Sample Standard Deviation:

$$s = \sqrt{s^2} = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2}$$

Probability:

- Any Two Events:

- $P(E_1 \text{ or } E_2) = P(E_1) + P(E_2) - P(E_1 \text{ and } E_2)$
- $P(E_1|E_2) = \frac{P(E_1 \text{ and } E_2)}{P(E_2)}, \quad P(E_2) > 0$
- $P(E_1 \text{ and } E_2) = P(E_1)P(E_2|E_1) = P(E_2)P(E_1|E_2)$

- Mutually Exclusive Events:

- $P(E_1 \text{ or } E_2) = P(E_1) + P(E_2)$

- Independent Events:

- $P(E_1|E_2) = P(E_1), \quad P(E_2)$
- $P(E_2|E_1) = P(E_2), \quad P(E_1)$
- $P(E_1 \text{ and } E_2) = P(E_1)P(E_2)$

Normal Distribution

- Standardized Value:

$$z = \frac{x - \mu}{\sigma}$$

Sampling Distributions

- Sample Means:

$$\bar{x} \sim N\left(\mu, \frac{\sigma}{\sqrt{n}}\right)$$

$$z = \frac{\bar{x} - \mu_{\bar{x}}}{\sigma_{\bar{x}}} = \frac{\bar{x} - \mu}{\left(\frac{\sigma}{\sqrt{n}}\right)}$$

- Sample Proportions:

$$\hat{p} = \frac{x}{n}$$

$$\hat{p} \sim N\left(\pi, \sqrt{\frac{\pi(1-\pi)}{n}}\right)$$

$$z = \frac{\hat{p} - \mu_{\hat{p}}}{\sigma_{\hat{p}}} = \frac{\hat{p} - \pi}{\left(\sqrt{\frac{\pi(1-\pi)}{n}}\right)}$$

Confidence Intervals

- General Form:

- Point Estimate \pm Margin of Error
- Margin of Error = Critical Value \times Standard Error

- Sample Means:

- Confidence Interval:

$$\bar{x} \pm (t^*) \frac{s}{\sqrt{n}}$$

- Sample Size:

$$n = \frac{z^2 \hat{\sigma}^2}{e^2}$$

- Sample Proportions:

- Confidence Interval:

$$\hat{p} \pm (z^*) \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$$

- Sample Size:

$$n = \frac{z^2 \hat{\pi}(1-\hat{\pi})}{e^2}$$

Hypothesis Testing

- General Form of Test Statistic:

$$\text{Test Statistic} = \frac{\text{Statistic} - \text{Null Value}}{\text{Standard Error}}$$

- Sample Means:

– Test Statistic:

$$t = \frac{\bar{x} - \mu_0}{\left(\frac{s}{\sqrt{n}}\right)}$$

- Sample Proportions:

– Test Statistic:

$$z = \frac{\hat{p} - \pi_0}{\sqrt{\frac{\pi_0(1-\pi_0)}{n}}}$$

Correlation

- Sample Correlation Coefficient:

$$r = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\left[\sum_{i=1}^n (x_i - \bar{x})^2\right] \left[\sum_{i=1}^n (y_i - \bar{y})^2\right]}} = \frac{1}{n-1} \sum_{i=1}^n \left(\frac{x_i - \bar{x}}{s_x}\right) \left(\frac{y_i - \bar{y}}{s_y}\right)$$

Simple Linear Regression

- Population Simple Linear Regression Model:

$$y_i = \beta_0 + \beta_1 x_i + \varepsilon_i,$$

- Sample Simple Linear Regression Model:

$$\hat{y}_i = b_0 + b_1 x_i,$$

– Sample Slope Coefficient Calculation:

$$b_1 = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2} = r \cdot \frac{\sqrt{\sum_{i=1}^n (y_i - \bar{y})^2}}{\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2}} = r \cdot \frac{s_y}{s_x}$$

– Sample Intercept Coefficient Calculation:

$$b_0 = \bar{y} - b_1 \bar{x}$$

- Sum of Squares Error (Residuals):

$$SSE = \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

- Sum of Squares Regression:

$$SSR = \sum_{i=1}^n (\hat{y}_i - \bar{y})^2$$

- Total Sum of Squares:

$$TSS = SSE + SSR = \sum_{i=1}^n (y_i - \bar{y})^2$$

- Coefficient of Determination:

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

- Inference for Regression:

$$\text{Test Statistic} = \frac{\text{Statistic} - \text{Null Value}}{\text{Standard Error}}$$

$$t = \frac{b_1 - 0}{s_{b_1}}, \quad d.f. = n - 2$$

– Estimate for σ_{b_1} :

$$s_{b_1} = \frac{s_\varepsilon}{\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2}}$$

– Estimate for σ_ε :

$$s_\varepsilon = \sqrt{\frac{SSE}{n - k - 1}}$$

- Confidence Interval for Slope:

$$b_1 \pm t^* \cdot s_{b_1}$$

Multiple Linear Regression

- Population Multiple Linear Regression Model:

$$y_i = \beta_0 + \beta_1 x_{1,i} + \beta_2 x_{2,i} + \dots + \beta_k x_{k,i} + \varepsilon_i$$

- Sample Multiple Linear Regression Model:

$$\hat{y}_i = b_0 + b_1 x_{1,i} + b_2 x_{2,i} + \dots + b_k x_{k,i}$$

- Adjusted R^2 Value:

$$R_A^2 = 1 - (1 - R^2) \left(\frac{n - 1}{n - k - 1} \right)$$

- Mallows C_p :

$$C_p = p + \frac{(MSE_p - MSE_{full})(n - p)}{MSE_{full}}$$

- Akaike's Information Criteria (AIC):

$$AIC = n \log \left(\frac{SSE}{n} \right) + 2p$$

- Schwartz Bayesian Criteria (BIC/SBC/SC):

$$BIC = n \log \left(\frac{SSE}{n} \right) + p \log(n)$$

- Sample Polynomial Linear Regression Model:

$$\hat{y}_i = b_0 + b_1 x_{1,i} + b_2 x_{1,i}^2 + \dots + b_k x_{1,i}^k$$

Inference for Multiple Regression

- Sum of Squares Error (Residuals):

$$SSE = \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

- Sum of Squares Regression:

$$SSR = \sum_{i=1}^n (\hat{y}_i - \bar{y})^2$$

- Total Sum of Squares:

$$\text{TSS} = \text{SSE} + \text{SSR} = \sum_{i=1}^n (y_i - \bar{y})^2$$

- Mean Square Regression:

$$\text{MSR} = \frac{\sum_{i=1}^n (\hat{y}_i - \bar{y})^2}{k} = \frac{\text{SSR}}{k}$$

- Mean Square Error:

$$\text{MSE} = \frac{\sum_{i=1}^n (y_i - \hat{y}_i)^2}{n - k - 1} = \frac{\text{SSE}}{n - k - 1}$$

- F -Test Statistic:

$$F = \frac{\left(\frac{\text{SSR}}{k}\right)}{\left(\frac{\text{SSE}}{n-k-1}\right)} = \frac{\text{MSR}}{\text{MSE}}$$

- Test Statistic:

$$\text{Test Statistic} = \frac{\text{Statistic} - \text{Null Value}}{\text{Standard Error}}$$

$$t = \frac{b_j - 0}{s_{b_j}}, \quad d.f. = n - k - 1$$

- Standard Error:

$$s_{b_j} = \frac{s_\varepsilon}{(1 - R_j^2)\sqrt{\sum_{i=1}^n (x_{j,i} - \bar{x}_j)^2}}, \quad s_\varepsilon = \sqrt{\frac{\text{SSE}}{n - k - 1}} = \sqrt{\text{MSE}}$$

Multicollinearity

- Variance Inflation Factor:

$$VIF = \frac{1}{1 - R_j^2}$$

Two Population Means

- Equal Variances:

- Test Statistic:

$$t = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_{1,0} - \mu_{2,0})}{s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

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

$$d.f. = n_1 + n_2 - 2$$

- Confidence Interval:

$$(\bar{x}_1 - \bar{x}_2) \pm t^* \cdot s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$$

- Unequal Variances:

- Test Statistic:

$$t = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_{1,0} - \mu_{2,0})}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

$$d.f. = \frac{(s_1^2/n_1 + s_2^2/n_2)^2}{\left(\frac{(s_1^2/n_1)^2}{n_1 - 1} + \frac{(s_2^2/n_2)^2}{n_2 - 1}\right)}$$

- Confidence Interval:

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

Two Population Variances

- Test Statistic:

$$F = \frac{s_i^2}{s_j^2}$$

numerator $d.f. = n_i - 1$, denominator $d.f. = n_j - 1$

Paired Differences

- Sample Differences:

$$\bar{d} = \frac{1}{n_d} \sum_{i=1}^{n_d} d_i, \quad d_i = x_{1,i} - x_{2,i}$$

- Test Statistic:

$$t = \frac{\bar{d} - \mu_{d,0}}{\left(\frac{s_d}{\sqrt{n_d}}\right)}$$

$$s_d = \sqrt{\frac{1}{n_d - 1} \sum_{i=1}^{n_d} (d_i - \bar{d})^2}$$

$$d.f. = n_d - 1$$

- Confidence Interval:

$$\bar{d} \pm t^* \cdot \frac{s_d}{\sqrt{n_d}}$$

Two Population Proportions

- Test Statistic:

$$z = \frac{(\hat{p}_1 - \hat{p}_2) - (\pi_{1,0} - \pi_{2,0})}{\sqrt{\bar{p}(1 - \bar{p}) \left(\frac{1}{n_1} + \frac{1}{n_2}\right)}}$$

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

- Confidence Interval:

$$(\hat{p}_1 - \hat{p}_2) \pm z^* \cdot \sqrt{\frac{\hat{p}_1(1 - \hat{p}_1)}{n_1} + \frac{\hat{p}_2(1 - \hat{p}_2)}{n_2}}$$

One-Way ANOVA

- Sum of Squares Between:

$$SSB = \sum_{i=1}^k n_i (\bar{x}_i - \bar{\bar{x}})^2$$

- Sum of Squares Within:

$$SSW = \sum_{i=1}^k \sum_{j=1}^{n_i} (x_{ij} - \bar{x}_i)^2$$

- Total Sum of Squares:

$$TSS = SSB + SSW = \sum_{i=1}^k \sum_{j=1}^{n_i} (x_{ij} - \bar{\bar{x}})^2$$

- Mean Sum of Squares Between:

$$MSB = \frac{1}{k-1} \sum_{i=1}^k n_i (\bar{x}_i - \bar{\bar{x}})^2 = \frac{SSB}{k-1}$$

- Mean Sum of Squares Within:

$$MSW = \frac{1}{N-k} \sum_{i=1}^k \sum_{j=1}^{n_i} (x_{ij} - \bar{x}_i)^2 = \frac{SSW}{N-k}$$

- Hypothesis Statement:

$$H_0 : \mu_1 = \mu_2 = \dots = \mu_k$$

$$H_A : \text{At least two means are not equal}$$

- Test Statistic:

$$F = \frac{MSB}{MSW}, \text{ numerator } d.f. = k-1, \text{ denominator } d.f. = N-k$$

- Tukey-Kramer Critical Range:

$$\text{Critical Range (Margin of Error)} = q_\alpha \cdot \sqrt{\frac{MSW}{2} \left(\frac{1}{n_i} + \frac{1}{n_j} \right)}$$

ANOVA with Randomized Block

- Sum of Squares Blocks:

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

- Total Sum of Squares:

$$TSS = SSBL + SSB + SSW$$

- Mean Sum of Squares Blocks:

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

- Mean Sum of Squares Between:

$$MSB = \frac{SSB}{k-1}$$

- Mean Sum of Squares Within:

$$MSW = \frac{SSW}{(k-1)(b-1)}$$

- Hypothesis Statement:

$$H_0 : \mu_1 = \mu_2 = \dots = \mu_k$$

$$H_A : \text{At least two means are not equal}$$

- Test Statistic:

$$F = \frac{MSB}{MSW}, \text{ numerator } d.f. = k-1, \text{ denominator } d.f. = (k-1)(b-1)$$

- Hypothesis Statement:

$$H_0 : \mu_{b_1} = \mu_{b_2} = \dots = \mu_{b_b}$$

$$H_A : \text{At least two block means are not equal}$$

- Test Statistic:

$$F = \frac{MSBL}{MSW}, \text{ numerator } d.f. = b-1, \text{ denominator } d.f. = (k-1)(b-1)$$

- Fisher's Least Squares Difference:

$$LSD = t^* \cdot \sqrt{MSW} \cdot \sqrt{\frac{2}{b}}$$

Categorical Data Analysis

- Pearson χ^2 Test of Association:

$$Q_P = \sum \left[\frac{(O_{i,j} - E_{i,j})^2}{E_{i,j}} \right]$$

- Likelihood Ratio χ^2 Test of Association:

$$Q_{LR} = 2 \cdot \sum_i \sum_j O_{i,j} \log \left(\frac{O_{i,j}}{E_{i,j}} \right)$$

- Mantel-Haenszel χ^2 Test of Association:

$$Q_{MH} = (n - 1)r^2$$

- Cramer's V Statistic:

$$V = \sqrt{\frac{Q_P/n}{\min(R - 1, C - 1)}}$$

- Odds:

$$odds(A) = \frac{P(A)}{1 - P(A)}$$

Logistic Regression

- Logit Transformation:

$$logit(p_i) = \log \left(\frac{p_i}{1 - p_i} \right)$$

- Logistic Regression Model:

$$logit(p_i) = \beta_0 + \beta_1 x_1 + \dots + \beta_k x_k$$