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$$\bar{x} = (\sum x_i)/n \quad Q_1 \text{ Position} = (n+3)/4 \quad Q_2 \text{ Position} = (n+1)/2 \quad Q_3 \text{ Position} = (3n+1)/4$$

$$s^2 = 1/(n-1) * [\sum x_i^2 - 1/n * (\sum x_i)^2] \quad \text{St.Dev.} = \sqrt{\text{Var}} \quad \text{Range} = \text{Max} - \text{Min} \quad \text{IQR} = Q_3 - Q_1$$

$$\text{CV} = (s / \bar{x}) * 100 \quad \text{Chebyshev's Rule: at least } [100 * (1 - 1/z^2)]\% \text{ where } z = \text{the number of st. dev.}$$

$$\text{Cov}(X, Y) = s_{xy} = 1/(n-1) * [\sum x_i y_i - 1/n * (\sum x_i) * (\sum y_i)] \quad r = s_{xy} / [s_x * s_y]$$

$$P(A) = 1 - P(A^c) \quad P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B) \quad P(B|A) = P(A \text{ and } B) / P(A)$$

$$P(A \text{ and } B) = P(B|A) * P(A) \quad \text{Two events are mutually exclusive if } P(A \text{ and } B) = 0$$

$$\text{Two events are independent if } P(B|A) = P(B)$$

$$E(X) = \sum \{x_i * P(X=x_i)\} \quad \text{Var}(X) = [\sum \{x_i^2 * P(X=x_i)\}] - [E(X)]^2$$

$$n! = n * (n-1) * (n-2) * \dots * 1 \quad {}_n C_x = n! / [x! * (n-x)!]$$

$$\text{Binomial Distribution: } {}_n C_x * p^x * (1-p)^{(n-x)} \quad E(X) = n * p \quad \text{Var}(X) = n * p * (1-p)$$

$$\text{Normal Distribution: } E(X) = \mu \quad \text{Var}(X) = \sigma^2$$

$$\text{Standard Normal Distribution: } E(Z) = 0 \quad \text{Var}(Z) = 1 \quad Z = (x - \mu) / \sigma$$

$$\text{Sampling Distribution of } \bar{x}: E(\bar{x}) = \mu \quad \text{Var}(\bar{x}) = \sigma^2 / n \quad Z = (\bar{x} - \mu) / (\sigma / \sqrt{n})$$

$$\text{Sampling Distribution of } \bar{p}: E(\bar{p}) = p \quad \text{Var}(\bar{p}) = p(1-p) / n \quad Z = (\bar{p} - p) / \sqrt{[p(1-p) / n]}$$

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Inference on μ , Known σ

$$Z_{\text{stat}} = (\bar{x} - \mu_0) / (\sigma / \sqrt{n})$$

$$\text{C.I.: } \bar{x} \pm Z_{\alpha/2} (\sigma / \sqrt{n})$$

Inference on μ , Unknown σ

$$t_{\text{stat}} = (\bar{x} - \mu_0) / (s / \sqrt{n})$$

$$\text{C.I.: } \bar{x} \pm t_{\alpha/2, n-1} (s / \sqrt{n})$$

Inference on p , Large Sample

$$Z_{\text{stat}} = (\bar{p} - p_0) / \sqrt{[p_0(1-p_0)/n]}$$

$$\text{C.I.: } \bar{p} \pm Z_{\alpha/2} \sqrt{[p(1-p)/n]}$$

Inference on $\mu_1 - \mu_2$, Known σ s, Indep. Samples

$$Z_{\text{stat}} = [(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)_0] / \sqrt{[(\sigma_1^2/n_1) + (\sigma_2^2/n_2)]}$$

$$\text{C.I.: } (\bar{x}_1 - \bar{x}_2) \pm Z_{\alpha/2} \sqrt{[(\sigma_1^2/n_1) + (\sigma_2^2/n_2)]}$$

Inference on $\mu_1 - \mu_2$, Unknown σ s, Indep. Samples

$$t_{\text{stat}} = [(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)_0] / \sqrt{[(s_1^2/n_1) + (s_2^2/n_2)]}$$

$$\text{C.I.: } (\bar{x}_1 - \bar{x}_2) \pm t_{\alpha/2, df} \sqrt{[(s_1^2/n_1) + (s_2^2/n_2)]}$$

$$\text{where } df = [s_1^2/n_1 + s_2^2/n_2]^2 / [(s_1^2/n_1)^2/(n_1-1) + (s_2^2/n_2)^2/(n_2-1)] \\ \text{round } df \text{ down}$$

Inference on $\mu_1 - \mu_2$, Unknown σ s, Indep. Samples, $\sigma_1^2 = \sigma_2^2$ Assumption

$$t_{\text{stat}} = [(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)_0] / \sqrt{[(s_{\text{pool}}^2/n_1) + (s_{\text{pool}}^2/n_2)]}$$

$$\text{C.I.: } (\bar{x}_1 - \bar{x}_2) \pm t_{\alpha/2, n_1+n_2-2} \sqrt{[(s_{\text{pool}}^2/n_1) + (s_{\text{pool}}^2/n_2)]}$$

$$\text{where } s_{\text{pool}}^2 = [(n_1-1)s_1^2 + (n_2-1)s_2^2] / (n_1+n_2-2)$$

Inference on $\mu_1 - \mu_2$, Unknown σ s, Dep. Samples ($\mu_1 - \mu_2 = \mu_d$)

$$t_{\text{stat}} = (\bar{d} - \mu_{d,0}) / (s_d / \sqrt{n}) \quad \text{where } \bar{d} = (\sum d_i) / n$$

$$\text{C.I.: } \bar{d} \pm t_{\alpha/2, n-1} (s_d / \sqrt{n}) \quad s_d^2 = \{1/(n-1)\} [\sum d_i^2 - (1/n)(\sum d_i)^2]$$

Inference on $p_1 - p_2$, Large Indep. Samples, $(p_1 - p_2)_0 = 0$

$$Z_{\text{stat}} = [(\bar{p}_1 - \bar{p}_2) - 0] / \sqrt{\{ \bar{p}_{\text{pool}}(1 - \bar{p}_{\text{pool}})/n_1 \} + \{ \bar{p}_{\text{pool}}(1 - \bar{p}_{\text{pool}})/n_2 \}}$$

$$\text{C.I.: } (\bar{p}_1 - \bar{p}_2) \pm Z_{\alpha/2} \sqrt{\{ \bar{p}_1(1 - \bar{p}_1)/n_1 \} + \{ \bar{p}_2(1 - \bar{p}_2)/n_2 \}}$$

$$\text{where } \bar{p}_{\text{pool}} = (x_1 + x_2) / (n_1 + n_2)$$

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$$SS_{xx} = \sum x_i^2 - 1/n(\sum x_i)^2$$

$$b_1 = SS_{xy}/SS_{xx}$$

$$df_{Reg} = 1$$

$$df_{Err} = n - 2$$

$$df_{Tot} = n - 1$$

$$r^2 = SSR/SST$$

$$s_{b1} = \sqrt{[MSE/SS_{xx}]}$$

$$s_{yhat} = \sqrt{[MSE\{1/n + (x_0 - \bar{x})^2/SS_{xx}\}]}$$

$$s_{ind} = \sqrt{[MSE\{1 + 1/n + (x_0 - \bar{x})^2/SS_{xx}\}]} = \sqrt{[MSE + (s_{yhat})^2]} \text{ for the same } x_0$$

$$\text{P.I. for an individual new y given } x=x_0: y_{hat} \pm t_{\alpha/2, n-2} * s_{ind}$$

$$SS_{yy} = \sum y_i^2 - 1/n(\sum y_i)^2$$

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

$$SSR = b_1 * SS_{xy} = \sum (y_{i,hat} - \bar{y})^2$$

$$SSE = SST - SSR = \sum (y_i - y_{i,hat})^2$$

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

$$corr = \star \sqrt{r^2}$$

$$\text{C.I. for } \beta_1: b_1 \pm t_{\alpha/2, n-2} * s_{b1}$$

$$\text{C.I. for the mean of y given } x=x_0: y_{hat} \pm t_{\alpha/2, n-2} * s_{yhat}$$

$$SS_{xy} = \sum x_i y_i - 1/n(\sum x_i)(\sum y_i)$$

$$y_{hat} = b_0 + b_1 x$$

$$MSR = SSR/df_{Reg}$$

$$MSE = SSE/df_{Err} = s_e^2$$

$$\text{Global } F_{stat} = MSR/MSE$$

$$t_{stat} = [b_1 - 0] / s_{b1}$$