$$\mu = \frac{\sum_{i=1}^{N} x_i}{N}$$

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

$$\sigma^2 = \frac{\sum_{i=1}^{N} (x_i - \mu)^2}{N}$$

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

$$s^2 = \frac{1}{n - 1} \left[\sum_{i=1}^{n} x_i^2 - \frac{(\sum_{i=1}^{n} x_i)^2}{n} \right]$$

$$\sigma = \sqrt{\sigma^2}$$

$$s = \sqrt{s^2}$$

$$1 - \frac{1}{k^2}$$

$$L_P = (n + 1) \frac{P}{100}$$

$$s_{xy} = \frac{\sum_{i=1}^{n} (x_i - \bar{x})(y_i - \bar{y})}{n - 1}$$

$$s_{xy} = \frac{1}{n - 1} \left[\sum_{i=1}^{n} x_i y_i - \frac{\sum_{i=1}^{n} x_i \sum_{i=1}^{n} y_i}{n} \right]$$

$$r = \frac{s_{xy}}{s_x s_y}$$

$$\rho = \frac{\sigma_{xy}}{\sigma_x \sigma_y}$$

$$P(A \mid B) = \frac{P(A \text{ and } B)}{P(B)}$$

$$P(A \mid B) = P(A)$$

$$P(B \mid A) = P(B)$$

$$P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$$

$$P(A \text{ and } B) = P(A \mid B) \times P(B)$$

$$P(A \text{ and } B) = P(A \mid B) \times P(B)$$

$$P(A \text{ and } B) = P(B | A) \times P(A)$$

$$P(A | B) = \frac{P(B | A) \times P(A)}{P(B)}$$

$$E(X) = \mu = \sum_{all x} xP(x)$$

$$V(X) = \sigma^2 = E[(X - \mu)^2] = \sum_{all x} (x - \mu)^2 P(x) = E(X^2) - E(X)^2$$

$$\sigma = \sqrt{\sigma^2}$$

$$E(c) = c$$

$$E(X + c) = E(X) + c$$

$$E(cX) = cE(X)$$

$$V(c) = 0$$

$$V(X + c) = V(X)$$

$$V(cX) = c^2 V(X)$$

$$P(x) = \sum_{y} P(x, y)$$

$$P(y) = \sum_{x} P(x, y)$$

$$COV(X, Y) = \sigma_{xy} = \sum_{x} \sum_{y} (x - \mu_x)(y - \mu_y)P(x, y) = E[(x - \mu_x)(y - \mu_y)]$$

$$COV(X, Y) = \sigma_{xy} = \sum_{x} \sum_{y} xyP(x, y) - \mu_x \mu_y = E(XY) - E(X)E(Y)$$

$$\rho = \frac{COV(X, Y)}{\sigma_x \sigma_y}$$

$$E(X + Y) = \sum_{x} \sum_{y} (x + y)P(x, y) = V(X) + V(Y) + 2COV(X, Y)$$

$$E(Y | X) = \mu_{y|x} = \sum_{y} yP(y|x)$$

$$P(x) = \frac{n!}{x!(n - x)!} p^x (1 - p)^{n - x}$$

$$\mu = np$$

$$\sigma^{2} = np(1-p)$$

$$\sigma = \sqrt{np(1-p)}$$

$$P(x) = \frac{e^{-\mu}\mu^{x}}{x!}$$

$$E(X) = V(X) = \mu$$

$$f(x) = \frac{1}{b-a}$$

$$E(X) = \frac{(a+b)}{2}$$

$$V(X) = \frac{(b-a)^{2}}{12}$$

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{1}{2}(\frac{x-\mu}{\sigma})^{2}}$$

$$E(X) = \mu$$

$$V(X) = \sigma^{2}$$

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

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

$$\bar{X} \sim N\left(\mu, \frac{\sigma^{2}}{n}\right)$$

$$\bar{\rho} = \frac{X}{n}$$

$$\hat{P} \sim N(p, \frac{p(1-p)}{n})$$

$$\sigma_{\hat{p}} = \sqrt{p(1-p)/n}$$

$$Z = \frac{\hat{P} - p}{\sqrt{p(1-p)/n}}$$

$$\mu_{\bar{x}-\bar{y}} = \mu_{x} - \mu_{y}$$

$$\sigma_{\bar{x}-\bar{y}} = \sqrt{\frac{\sigma_{x}^{2}}{n_{x}} + \frac{\sigma_{y}^{2}}{n_{y}}}$$

$$\bar{X} - \bar{Y} \sim N \left(\mu_{x} - \mu_{y}, \frac{\sigma_{x}^{2}}{n_{x}} + \frac{\sigma_{y}^{2}}{n_{y}}\right)$$

$$\bar{X} \pm z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$$

$$n = \left(\frac{z_{\alpha/2}\sigma}{B}\right)^{2}$$

$$Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$$

$$t = \frac{\bar{x} - \mu}{s/\sqrt{n}}$$

$$N \left[\bar{x} \pm t_{\alpha/2} \frac{s}{\sqrt{n}}\right]$$

$$\frac{(n-1)s^{2}}{\sigma^{2}} \sim \chi_{n-1}^{2}$$

$$\frac{(n-1)s^{2}}{\chi_{\alpha/2}^{2}}$$

$$\frac{(n-1)s^{2}}{\chi_{1-\alpha/2}^{2}}$$

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

$$t = \frac{(\bar{x}_{1} - \bar{x}_{2}) - (\mu_{1} - \mu_{2})}{\sqrt{s_{p}^{2}} \left(\frac{1}{n_{1}} + \frac{1}{n_{2}}\right)} \sim t_{n_{1}+n_{2}-2}$$

$$\bar{s}_{p}^{2} = \frac{(n_{1} - 1)s_{1}^{2} + (n_{2} - 1)s_{2}^{2}}{n_{1} + n_{2} - 2}$$

$$\bar{x}_{1} - \bar{x}_{2} \pm t_{\alpha/2} \sqrt{s_{p}^{2} \left(\frac{1}{n_{1}} + \frac{1}{n_{2}}\right)}$$

$$t = \frac{(\bar{x}_{1} - \bar{x}_{2}) - (\mu_{1} - \mu_{2})}{\sqrt{s_{1}^{2}} + \frac{s_{2}^{2}}{n_{2}}}$$

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

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

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

$$F_{1-A, v_1, v_2} = \frac{1}{F_{A, v_2, v_1}}$$

$$t = \frac{\bar{x}_D - \mu_D}{s_D / \sqrt{n_D}} \sim t_{n_D - 1}$$

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

$$\hat{p} = \frac{x_1 + x_2}{n_1 + n_2}$$

$$z = \frac{(\hat{p}_1 - \hat{p}_2) - (p_1 - p_2)}{\sqrt{\hat{p}(1 - \hat{p})\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}}$$

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

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

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

$$b_1 = \frac{s_{xy}}{s_x^2}$$

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

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

$$s_{\varepsilon}^2 = \frac{SSE}{n - 2}$$

$$s_{\varepsilon} = \sqrt{\frac{SSE}{n - 2}}$$

$$\begin{split} s_{b_1} &= \frac{s_{\varepsilon}}{\sqrt{(n-1)s_{x}^{2}}} \\ & t = \frac{b_{1} - \beta_{1}}{s_{b_{1}}} \sim t_{n-2} \\ R^{2} &= 1 - \frac{SSE}{SST} = 1 - \frac{\sum_{i=1}^{n} (y_{i} - \hat{y}_{i})^{2}}{\sum_{i=1}^{n} (y_{i} - \overline{y}_{i})^{2}} \end{split}$$