

Formula Sheet

$$\bar{x} = \frac{\sum x}{n} = \frac{\sum xf}{n}$$

$$SS(x) = \sum (x - \bar{x})^2 = \sum x^2 - n(\bar{x})^2 = \sum x^2 - \frac{(\sum x)^2}{n} = \sum x^2 f - \frac{(\sum xf)^2}{n}$$

$$s^2 = \frac{SS(x)}{n-1} \quad s = \sqrt{s^2} \quad z = \frac{x - \mu}{s}$$

$$\text{Chebyshev's Theorem: } 1 - \frac{1}{y^2}$$

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

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

$$b_1 = \frac{SS(xy)}{SS(x)} = \frac{\sum (x - \bar{x})(y - \bar{y})}{\sum (x - \bar{x})^2} \quad b_0 = \frac{\sum y - (b_1 \cdot \sum x)}{n} = \bar{y} - (b_1 \cdot \bar{x})$$

$$r = \frac{SS(xy)}{\sqrt{SS(x)SS(y)}}$$

$${}_nP_k = \frac{n!}{(n-k)!} \quad {}_nC_k = \frac{{}_nP_k}{k!} = \frac{n!}{(n-k)!k!}$$

$$P(A \cup B \cup C) = P(A) + P(B) + P(C) - P(A \cap B) - P(B \cap C) - P(A \cap C) + P(A \cap B \cap C)$$

$$\text{Law of Total Probability: } P(B) = P(B|A_1)P(A_1) + P(B|A_2)P(A_2) + \dots + P(B|A_k)P(A_k)$$

$$\text{Bayes' Theorem: } P(A_j|B) = \frac{P(B|A_j)P(A_j)}{P(B)} = \frac{P(B|A_j)P(A_j)}{\sum_{i=1}^k P(B|A_i)P(A_i)}$$

$$\mu = \sum xP(x) \qquad E(x) = n \cdot p$$

$$\sigma^2 = \sum (x-\mu)^2 P(x) = \sum x^2 P(x) - \mu^2 \qquad \sigma^2 = n \cdot p \cdot q \qquad \sigma = \sqrt{\sigma^2}$$

$$P(x) = {}_n C_x p^x q^{n-x} \qquad P(x) = \frac{\mu^x e^{-\mu}}{x!} \qquad P(x) = \frac{{}^M C_x \cdot {}^{(N-M)} C_{(n-x)}}{{}^N C_n}$$

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

$$\sigma_{\bar{x}}=\frac{\sigma_x}{\sqrt{n}} \qquad E=z(\alpha/2)\cdot\sigma_{\bar{x}}$$

$$p'=\frac{x}{n} \qquad \mu_{p'}=p \qquad \sigma_{p'}=\sqrt{\frac{p(1-p)}{n}} \qquad E=z(\alpha/2)\cdot\sqrt{\frac{p'(1-p')}{n}}$$

$$n=(z(\alpha/2))^2\cdot\frac{p(1-p)}{E^2} \qquad z=\frac{p'-p_0}{\sqrt{\frac{p_0(1-p_0)}{n}}}$$

$$df=n-1 \qquad E=t(\alpha/2)\cdot\frac{s}{\sqrt{n}} \qquad t=\frac{\bar{x}-\mu}{s/\sqrt{n}}$$

$$\overline{d}=\frac{\sum d}{n}=\overline{x_1}-\overline{x_2} \qquad \mu_{\overline{d}}=\mu_d \qquad s_d=\sqrt{\frac{\sum d^2-\frac{(\sum d)^2}{n}}{n-1}} \qquad E=t(\alpha/2)\cdot\frac{s_d}{\sqrt{n}} \qquad t=\frac{\overline{d}-\mu_d}{s_d/\sqrt{n}}$$

$$\mu_{\overline{x_1}-\overline{x_2}}=\mu_{x_1}-\mu_{x_2} \qquad s_{\overline{x_1}-\overline{x_2}}=\sqrt{\frac{s_1^2}{n_1}+\frac{s_2^2}{n_2}} \qquad E=t(\alpha/2)\cdot s_{\overline{x_1}-\overline{x_2}} \qquad t=\frac{(\overline{x_1}-\overline{x_2})-(\mu_1-\mu_2)}{s_{\overline{x_1}-\overline{x_2}}}$$

$$df=\frac{\left(\frac{s_1^2}{n_1}+\frac{s_2^2}{n_2}\right)^2}{\frac{s_1^4}{n_1^2(n_1-1)}+\frac{s_2^4}{n_2^2(n_2-1)}}$$

$$E=z(\alpha/2)\sqrt{\frac{p_1'(1-p_1')}{n_1}+\frac{p_2'(1-p_2')}{n_2}} \qquad P_p'=\frac{x_1+x_2}{n_1+n_2} \qquad z=\frac{p_1'-p_2'}{\sqrt{P_p'(1-P_p')\left(\frac{1}{n_1}+\frac{1}{n_2}\right)}}$$

$$\chi^2=\sum\frac{(f_i-e_i)^2}{e_i} \qquad df=k-1$$