

USEFUL FORMULAE

$$E(X) = \sum_{i=1}^k p_i x_i$$

$$Var(X) = E[(X - \mu_X)^2] = \sum_{i=1}^k (x_i - \mu_X)^2 p_i$$

$$Pr(X = x) = \sum_{i=1}^k Pr(X = x, Y = y_i)$$

$$Pr(Y = y | X = x) = \frac{Pr(X = x, Y = y)}{Pr(X = x)}$$

$$E(Y | X = x) = \sum_{i=1}^k y_i Pr(Y = y_i | X = x)$$

$$E(Y) = \sum_{i=1}^m E(Y | X = x_i) Pr(X = x_i)$$

$$Var(Y | X = x) = \sum_{i=1}^k [y_i - E(Y | X = x)]^2 Pr(Y = y_i | X = x)$$

$$E(a + bX + cY) = a + bE(X) + cE(Y)$$

$$Cov(X, Y) = \sum_{i=1}^k \sum_{j=1}^m (x_j - \mu_X)(y_i - \mu_Y) Pr(X = x_j, Y = y_i)$$

$$Var(a + bY) = b^2 Var(Y)$$

$$Corr(X, Y) = \rho_{XY} = \frac{Cov(X, Y)}{\sqrt{Var(X)Var(Y)}}$$

$$Var(aX + bY) = a^2 Var(X) + b^2 Var(Y) + 2ab Cov(X, Y)$$

$$E(Y^2) = Var(Y) + E(Y)^2$$

$$Cov(a + bX + cV, Y) = bCov(X, Y) + cCov(V, Y)$$

$$E(XY) = Cov(X, Y) + E(X)E(Y)$$

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

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

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

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

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

$$r_{XY} = s_{XY} / s_X s_Y$$

$$\hat{Y}_i = \hat{\beta}_0 + \hat{\beta}_1 X_{1i} + \hat{\beta}_2 X_{2i} + \dots + \hat{\beta}_k X_{ki}$$

For the simple linear regression model $Y_i = \beta_0 + \beta_1 X_i + \varepsilon_i$ we have: $\hat{\beta}_1 = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sum_{i=1}^n (X_i - \bar{X})^2}$ $\hat{\beta}_0 = \bar{Y} - \hat{\beta}_1 \bar{X}$

For the even simpler regression model $Y_i = \beta X_i + \varepsilon_i$ we have: $\hat{\beta} = \frac{\sum_{i=1}^n (X_i Y_i)}{\sum_{i=1}^n (X_i)^2}$