#### MATU9D2: PRACTICAL STATISTICS: FORMULA SHEET

## 1. One Sample Summary Statistics

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

$$s^2 = \frac{\sum x_i^2 - \frac{\left(\sum x_i\right)^2}{n}}{n-1}$$

## 2. Two Sample Summary Statistics

$$\overline{x}_{1} = \frac{\sum x_{1i}}{n_{1}} \qquad s_{1}^{2} = \frac{\sum x_{1i}^{2} - \frac{\left(\sum x_{1i}\right)^{2}}{n_{1}}}{n_{1} - 1}$$

$$\overline{x}_{2} = \frac{\sum x_{2i}}{n_{2}} \qquad s_{2}^{2} = \frac{\sum x_{2i}^{2} - \frac{\left(\sum x_{2i}\right)^{2}}{n_{2}}}{n_{2} - 1}$$

$$s_{p}^{2} = \frac{\left(n_{1} - 1\right)s_{1}^{2} + \left(n_{2} - 1\right)s_{2}^{2}}{n_{1} + n_{2} - 2}$$

## 3. Correlation and Regression

$$S_{XY} = \sum x_i y_i - \frac{\sum x_i \sum y_i}{n}$$

$$S_{XX} = \sum x_i^2 - \frac{\left(\sum x_i\right)^2}{n}$$

$$S_{YY} = \sum y_i^2 - \frac{\left(\sum y_i\right)^2}{n}$$

$$r = \frac{S_{XY}}{\sqrt{S_{XX} S_{YY}}}$$

$$R^2 = \frac{S_{XY}^2}{S_{XX} S_{YY}}$$

The least squares line is estimated using

$$\hat{\beta} = \frac{S_{XY}}{S_{YY}}, \qquad \hat{\alpha} = \bar{y} - \hat{\beta}\bar{x}$$

## 4. Relationships

The Chi-squared test has test statistic

$$X^{2} = \sum_{all \ cells} \frac{\left(O_{ij} - E_{ij}\right)^{2}}{E_{ij}}$$

Using a significance level of 0.05, this is compared with  $\chi^2(df; 0.05)$ , where df= (r-1)(c-1).

The test of the Null hypothesis  $H_0$ :  $\rho=0$  against the alternative  $H_1$ :  $\rho\neq 0$  has test Statistic

$$t = \frac{r\sqrt{n-2}}{\sqrt{1-r^2}}$$

Using significance level of 0.05, this is compared with  $\pm$  t(n-2; 0.025).

## 5. Probability

A random variable X following a Binomial probability model with n trials and probability of success  $\theta$  has probability distribution

$$\Pr[X = x] = \frac{n!}{x!(n-x)!} \theta^x (1-\theta)^{n-x}, \quad x = 0, \dots, n.$$

and mean and variance given by

$$E(X) = n\theta$$
,  $Var(X) = n\theta(1 - \theta)$ .

If X follows a Normal probability model with mean  $\mu$  and variance  $\sigma^2$ , then  $Z = \frac{(X - \mu)}{\sigma}$  follows a N(0,1) distribution.

## 6. One Sample Confidence Intervals

An interval estimate for a population *proportion*, having approximate confidence 0.95, is

$$\hat{\theta} \pm 1.96 \times \sqrt{\frac{\hat{\theta}(1-\hat{\theta})}{n}}$$

An interval estimate for a population mean, having confidence 0.95, is

$$\overline{x} \pm t(n-1;0.025) \times \frac{s}{\sqrt{n}}$$

An interval estimate for a population variance, having confidence 0.95, is

$$\left(\frac{(n-1)s^2}{\chi^2(n-1;0.025)},\frac{(n-1)s^2}{\chi^2(n-1;0.975)}\right).$$

## 7. Two Sample Tests

The test statistics for the test of the Null Hypothesis  $H_0$ :  $\mu_1 = \mu_2$  against the alternative  $H_1$ :  $\mu_1 \neq \mu_2$  is

$$t = \frac{\overline{x}_1 - \overline{x}_2}{\sqrt{s_p^2 \left(\frac{1}{n_1} + \frac{1}{n_2}\right)}}$$

For a test having significance level 0.05, we compare this with  $\pm$  t(  $\,n_1+n_2$  - 2 ; 0.025 ).

The test statistic for the test of the Null Hypothesis  $H_0: \sigma_1^2 = \sigma_2^2$  against the alternative  $H_1: \sigma_1^2 \neq \sigma_2^2$  is

$$\frac{s_1^2}{s_2^2} \left( \frac{larger}{smaller} \right)$$

For a test having significance level 0.05, we compare this with

$$F(n_1-1,n_2-1;0.025)$$

## 8. Inference in Simple Linear Regression

The residual variance  $\sigma^2$  is estimated by

$$\hat{\sigma}^2 = \frac{S_{YY} - \frac{S_{XY}^2}{S_{XX}}}{n - 2}$$

$$\hat{\alpha} = \bar{y} - \hat{\beta}\bar{x}$$

$$\hat{\beta} = \frac{S_{XY}}{S_{YY}}$$

A test of the Null Hypothesis  $H_0: \beta = 0$  against the alternative  $H_1: \beta \neq 0$  has the test statistic

$$T = \frac{\hat{\beta}}{\sqrt{\frac{\hat{\sigma}^2}{S_{XX}}}}$$

Using a significance level of 0.05, this is compared with  $\pm t(n-2; 0.025)$ .

Parameter	Estimate	Estimated Standard Error
α	$\hat{lpha}$	$\sqrt{\hat{\sigma}^2 \left( \frac{1}{n} + \frac{\overline{x}^2}{S_{XX}} \right)}$
β	$\hat{eta}$	$\sqrt{rac{\hat{\sigma}^2}{S_{XX}}}$
$\alpha + \beta x$	$\hat{\alpha} + \hat{\beta}x$	$\sqrt{\hat{\sigma}^2 \left( \frac{1}{n} + \frac{\left( x - \overline{x} \right)^2}{S_{XX}} \right)}$

In each case interval estimates, having confidence 0.95, are given by

Estimate 
$$\pm t(n-2;0.025) \times Estimated Standard Error$$

The formula for a 95% prediction interval for an individual observation is

$$\hat{\alpha} + \hat{\beta}x \pm t(n-2;0.025)\sqrt{\hat{\sigma}^2\left(1 + \frac{1}{n} + \frac{(x-\bar{x})^2}{S_{XX}}\right)}$$

# 9. **TO COME**