

# PROBABILISTIC METHODS IN ENGINEERING VE401

## Assignment VII

Due: April 26, 2018

Team number: 20

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#### Exercise 7.1

The estimator of p is

$$\hat{p} = \frac{E}{n} = \frac{\overline{X}}{n} = \frac{0 \cdot 39 + 1 \cdot 23 + 2 \cdot 12 + 3 \cdot 1}{75 \cdot 24} = \frac{1}{36}.$$

The expected frequency can be calculated by  $E_i = nP[X = i] = n\binom{n}{i}\hat{p}^i(1-\hat{p})^{n-i}$ .

Values	0	1	2	3
Expected Frequency	38.145	26.156	8.594	2.105

We notice that  $E_i \geq 5$  less than 80% categories. Thus we merge the last two categories.

Values	0	1	2
Expected Frequency	38.145	26.156	10.699

Then, we calculate the Pearson's Statistic

$$X^{2} = \sum_{i=1}^{N} \frac{(O_{i} - E_{i})^{2}}{E_{i}} = \frac{(39 - 38.145)^{2}}{38.145} + \frac{(23 - 26.156)^{2}}{26.156} + \frac{(13 - 10.699)^{2}}{10.699} = 0.876.$$

The degree of freedom is 3-1-1=1. Hence,  $X^2=0.876<\chi^2_{0.05,1}=3.84$ . Hence, we conclude that follows binomial distribution at 5% level of significance.

#### Exercise 7.2

$$H_0: p_{ij} = p_{i.}p_{.j}$$

Since 
$$E_{ij} = \frac{n_i \cdot n_{\cdot j}}{n}$$
, then

$$E_{11} = 18.43, \quad E_{12} = 44.66, \quad E_{13} = 17.01, \quad E_{14} = 14.88$$
  
 $E_{21} = 7.57, \quad E_{22} = 18.34, \quad E_{23} = 6.99, \quad E_{24} = 6.11$ 

$$X_{(2-1)(4-1)}^2 = \sum_{i=1}^r \sum_{j=1}^c \frac{(O_{ij} - E_{ij})^2}{E_{ij}} = 10.71.$$

Since the statistic follows 3 degrees of freedom chi-squared distribution, we compare the value with  $\chi^2_{0.05,3} = 7.8147$ . We find that  $X_3^2 > \chi^2_{0.05,3}$ . Thus, we reject the hypothesis at 5% level of significance. We claim that the type of failure depends on the mounting position.

#### Exercise 7.3

	1	2	3	4	5	
Male						
Female	21	27	50	35	17	150
	71	74	153	111	41	450

We assume that the increase in the salary is independent od the workers' gender, saying that

$$H_0: p_{ij} = p_{i\cdot}p_{\cdot j}$$

Since 
$$E_{ij} = \frac{n_i \cdot n_{\cdot j}}{n}$$
, then

$$E_{11} = 47.33$$
,  $E_{12} = 49.33$ ,  $E_{13} = 102$ ,  $E_{14} = 74$ ,  $E_{15} = 27.33$ 

$$E_{21} = 23.67$$
,  $E_{22} = 24.67$ ,  $E_{23} = 51$ ,  $E_{24} = 37$ ,  $E_{25} = 13.67$ 

$$X_{(2-1)(5-1)}^2 = \sum_{i=1}^r \sum_{j=1}^c \frac{(O_{ij} - E_{ij})^2}{E_{ij}} = 2.19.$$

Since the statistic follows 4 degrees of freedom chi-squared distribution, we compare the value with  $\chi^2_{0.05,4} = 9.4877$ . We find that  $X_4^2 < \chi^2_{0.05,3}$ . Thus, we support the hypothesis at 5% level of significance. We claim that increase in salary is independent of the gender.

#### Exercise 7.4

Using the Helmert transformation  $\mathbf{Y} \to \mathbf{D}$ , we rewrite  $SSE_{pe}$ 

$$SSE_{pe} = \frac{1}{\sigma^2} \sum_{i=1}^{k} \sum_{j=1}^{n_i} (Y_{ij} - \overline{Y_{ij}})^2$$

$$= \frac{1}{\sigma^2} \sum_{i=1}^{k} \left[ \sum_{j=1}^{n_i} Y_{ij}^2 - n_i \overline{Y_{ij}}^2 \right]$$

$$= \frac{1}{\sigma^2} \sum_{i=1}^{k} \left[ \sum_{j=1}^{n_i} D_{ij}^2 - D_{i1}^2 \right]$$

$$= \frac{1}{\sigma^2} \sum_{i=1}^{k} \left[ \sum_{j=2}^{n_i} D_{ij}^2 \right]$$

$$= \frac{1}{\sigma^2} \left[ \sum_{i=1}^{n} D_i^2 - \sum_{i=1}^{k} D_{i1}^2 \right]$$

$$= \sum_{i=1}^{n} \frac{D_i^2}{\sigma^2} - \sum_{i=1}^{k} \frac{D_{i1}^2}{\sigma^2},$$

which is the sum of square of (n - k) normally distributed variables. Hence, it follows (n - k) degrees of freedom chi-squared distribution.

#### Exercise 7.5

Consider  $Cov(\overline{Y}, \hat{\beta}_1)$ , we obtain

$$\hat{\beta}_{1} = B_{1} = \frac{S_{xy}}{S_{xx}}$$

$$Cov(\overline{Y}, \hat{\beta}_{1}) = E[\overline{Y}\hat{\beta}_{1}] - E[\overline{Y}]E[B_{1}]$$

$$= E\left[\frac{\sum_{i=1}^{n}(x_{i} - \bar{x})Y_{i}\overline{Y}}{\sum_{i=1}^{n}(x_{i} - \bar{x})^{2}}\right] - (\beta_{0} + \beta_{1}\bar{x})\beta_{1}$$

$$= \frac{\sum_{i=1}^{n}(x_{i} - \bar{x})E[Y_{i}\overline{Y}]}{\sum_{i=1}^{n}(x_{i} - \bar{x})^{2}} - (\beta_{0} + \beta_{1}\bar{x})\beta_{1}$$

$$= \frac{\sum_{i=1}^{n}(x_{i} - \bar{x})(\beta_{0} + \beta_{1}x_{i})(\beta_{0} + \beta_{1}\bar{x})}{\sum_{i=1}^{n}(x_{i} - \bar{x})^{2}} - (\beta_{0} + \beta_{1}\bar{x})\beta_{1}$$

$$= 0$$

Consider  $Cov(\hat{\beta}_0, \hat{\beta}_1)$ , we obtain

$$\operatorname{Cov}(\hat{\beta}_{0}, \hat{\beta}_{1}) = E[B_{0}B_{1}] - E[B_{0}]E[B_{1}]$$

$$= E[(\overline{Y} - B_{1}\overline{x})B_{1}] - \beta_{0}\beta_{1}$$

$$= E[B_{1}\overline{Y} - B_{1}^{2}\overline{x}] - \beta_{0}\beta_{1}$$

$$= E[B_{1}\overline{Y}] - E[B_{1}^{2}\overline{x}] - \beta_{0}\beta_{1}$$

$$= (\beta_{0} + \beta_{1}\overline{x})\beta_{1} - \overline{x}E[B_{1}^{2}] - \beta_{0}\beta_{1}$$

$$= \overline{x}(\beta_{1}^{2} - E[B_{1}^{2}])$$

$$= -\overline{x}(E[(\frac{S_{xy}}{S_{xx}})^{2}] - E[\frac{S_{xy}}{S_{xx}}]^{2})$$

$$= -\overline{x}\operatorname{Var}(B_{1})$$

$$= -\overline{x}\frac{\sigma^{2}}{S_{xx}}$$

$$= -\frac{\overline{x}}{S_{xx}}\sigma^{2}.$$

#### Exercise 7.6

1)

$$B_1 = \frac{S_{xy}}{S_{xx}} = 184.6$$
  
 $B_0 = \overline{Y} - B_1 \overline{x} = -2150$ 

2)

$$S^2 = \hat{\sigma}^2 = \frac{S_{yy} - B_1 S_{xy}}{n - 2} = 1.15 \times 10^5.$$

3) For a 100(1-0.1)% confidence interval,

$$B_1 \pm t_{0.1/2,n-2} \frac{S}{\sqrt{S_{xx}}} = 184.6 \pm 1.68 \cdot \frac{339.2185}{\sqrt{828.2412}} = 184.6 \pm 19.8.$$

$$B_0 \pm t_{0.1/2,n-2} \frac{S\sqrt{\sum x_i^2}}{\sqrt{nS_{xx}}} = -2150 \pm 1.68 \cdot \frac{339.2185 \cdot 182.8296}{\sqrt{42 \cdot 828.2412}} = -2150 \pm 560.$$

4)

 $H_0$ : the linear regression is appropriate

 $H_1$ : the linear regression is not appropriate

$$SSE_{pe} = \sum_{i=1}^{k} \sum_{j=1}^{n_i} (Y_{ij} - \overline{Y}_i)^2 = 109500.$$
  
 $SSE_{lf} = SSE - SSE_{pe} = 4.49 \times 10^6.$ 

Hence,

$$F_{37-2,42-37} = \frac{4.49 \times 10^6 / (37-2)}{109500 / (42-37)} = 5.8621$$

Since  $f_{0.05,35,5} = 4.4775 < F_{35,5}$ , then we claim that the linear regression model is not appropriate at 5% level of significance.

5)

$$R^2 = \frac{S_{yy} - SSE}{S_{yy}} = 0.8597.$$

The coefficient of determination is the proportion in  $S_{yy}$  that is determined by X. The larger  $R^2$  is, the more Y are determined by X.

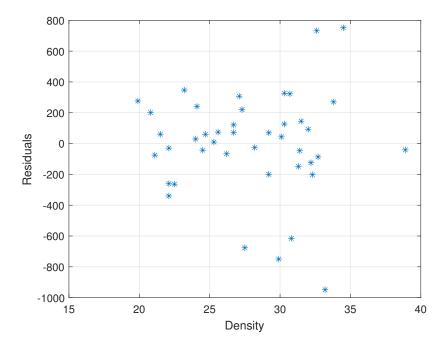


Figure 1: Residuals vs. Density

6) The assumption of constant variance is unreliable because the is irregular.

### Exercise 7.7

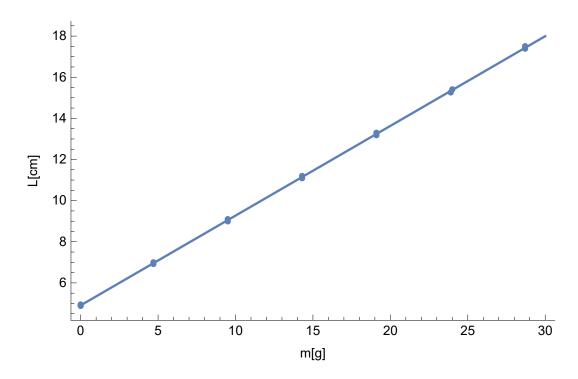


Figure 2: Linear regression model

1)

2)

$$R^2 = \frac{S_{yy} - SSE}{S_{yy}} = 0.9998.$$

Hence,

$$\frac{R\sqrt{n-2}}{\sqrt{1-R^2}} = 191.994 > t_{0.025,12} = 2.18.$$

It implies that our linear regression model is appropriate.

3)

 $H_0$ : the linear regression is appropriate

 $H_1$ : the linear regression is not appropriate

$$SSE_{pe} = \sum_{i=1}^{k} \sum_{j=1}^{n_i} (L_{ij} - \overline{L}_i)^2 = 0.0322.$$

$$SSE_{lf} = SSE - SSE_{pe} = 0.0398547 - 0.0322 = 0.00765.$$

Hence,

$$F_{8-2,14-8} = \frac{0.00765/(8-2)}{0.0322/(14-8)} = 0.2376.$$

Since  $f_{0.05,6,6} = 4.2839 > F_{6,6}$ , then we claim that the linear regression model is appropriate at 5% level of significance.

#### Exercise 7.8

LHS - RHS = 
$$\frac{S_{xy}}{S\sqrt{S_{xx}}} - \frac{\sqrt{n-2}S_{xy}/\sqrt{S_{xx}S_{yy}}}{\sqrt{1-S_{xy}^2/S_{xx}S_{yy}}}$$
  
=  $\frac{S_{xy}}{S\sqrt{S_{xx}}} - \frac{\sqrt{n-2}S_{xy}}{\sqrt{S_{xx}S_{yy}-S_{xy}^2}}$   
=  $\frac{S_{xy}}{\sqrt{(S_{xx}S_{yy}-S_{xy}^2)/(n-2)}} - \frac{\sqrt{n-2}S_{xy}}{\sqrt{S_{xx}S_{yy}-S_{xy}^2}}$   
=  $\frac{\sqrt{n-2}S_{xy}}{\sqrt{S_{xx}S_{yy}-S_{xy}^2}} - \frac{\sqrt{n-2}S_{xy}}{\sqrt{S_{xx}S_{yy}-S_{xy}^2}}$   
= 0.