

## Cover Sheet for Submission of Maths Examinations Summer 2020

We would advise preparing your coversheets with your CID, Module Name and Code and Date, before the exams are due to take place.

**CID: 01738166**

**Module Name: Probability and Statistics**

**Module Code: MATH40005**

**Date: 14/05/2020**

### Questions Answered (in the file):

Please tick next to the question or questions you have answered in this file.

<b>Q1</b>	
<b>Q2</b>	
<b>Q3</b>	
<b>Q4</b>	
<b>Q5</b>	✓
<b>Q6</b>	

(Note: this is a coversheet for all students - not all students will have exams with 6 questions. Please tick the boxes which are appropriate for your exam and/or the file you are submitting).

### (Optional) Page Numbers for each question;

<b>Page Number</b>	<b>Question Answered</b>

If handwritten, please complete in CAPITAL Letters, in Blue or Black Ink, ensuring the cover sheet is legible.

(a) For any given  $a$

$$\sum_{i=1}^n (x_i - a)^2 = \sum_{i=1}^n [(x_i - \bar{x}) + (\bar{x} - a)]^2 = \sum_{i=1}^n [(x_i - \bar{x})^2 + 2(x_i - \bar{x})(\bar{x} - a) + (\bar{x} - a)^2]$$

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

$$= \sum_{i=1}^n (x_i - \bar{x})^2 + 2(\bar{x} - a) \cdot 0 + n(\bar{x} - a)^2$$

$$= \sum_{i=1}^n (x_i - \bar{x})^2 + n(\bar{x} - a)^2, \text{ where we have used}$$

$$\sum_{i=1}^n (x_i - \bar{x}) = \sum_{i=1}^n x_i - \sum_{i=1}^n \bar{x} = n\bar{x} - n\bar{x} = 0$$

Since  $n(\bar{x} - a)^2 \geq 0$ , we get  $\sum_{i=1}^n (x_i - a)^2 \geq \sum_{i=1}^n (x_i - \bar{x})^2$ ,  
with equality only when  $\bar{x} = a$ .

(b) Denote  $RSS(b_0, b_1) = \sum_{i=1}^n [(y_i - b_1 x_i) - b_0]^2$  and write  $z_i = y_i - b_1 x_i$

$$\Rightarrow RSS(b_0, b_1) = \sum_{i=1}^n (z_i - b_0)^2.$$

From (a) we get that  $RSS(b_0, b_1) \geq \sum_{i=1}^n [z_i - \bar{z}]^2$

$$\Rightarrow \hat{b}_0 = \frac{1}{n} \sum_{i=1}^n (y_i - b_1 x_i) = \bar{y} - b_1 \bar{x}. \quad RSS(\hat{b}_0, b_1) \leq RSS(b_0, b_1).$$

Now find  $\hat{b}_1$ :

$$\begin{aligned} RSS(\hat{b}_0, b_1) &= \sum_{i=1}^n [(y_i - b_1 x_i) - (\bar{y} - b_1 \bar{x})]^2 = \sum_{i=1}^n [(y_i - \bar{y}) - b_1(x_i - \bar{x})]^2 \\ &= \sum_{i=1}^n [(y_i - \bar{y})^2 - 2b_1(x_i - \bar{x})(y_i - \bar{y}) + b_1^2(x_i - \bar{x})^2] = S_{yy} - 2b_1 S_{xy} + b_1^2 S_{xx} \end{aligned}$$

$$\Rightarrow \text{RSS}(\hat{b}_0, \hat{b}_1) = \sum_{i=1}^n S_{xx} \left( b_1 - \frac{S_{xy}}{S_{xx}} \right)^2 + \left( S_{yy} - \frac{(S_{xy})^2}{S_{xx}} \right)^2$$

$$\Rightarrow \text{RSS}(\hat{b}_0, \hat{b}_1) \leq \text{RSS}(\hat{b}_0, b_1) \quad \text{where} \quad \hat{b}_1 = \frac{S_{xy}}{S_{xx}}$$

$$\Rightarrow \hat{b}_1 = \frac{S_{xy}}{S_{xx}}, \quad \hat{b}_0 = \bar{y} - \hat{b}_1 \bar{x} = \bar{y} - \left( \frac{S_{xy}}{S_{xx}} \right) \bar{x}$$

(c)

$$(i) H_0: \mu_x = \mu_y$$

(ii) Assumptions:  $x_1, x_2, \dots, x_n$  are observations of independent random variables

Assume  $x_1, \dots, x_n$  the observations are drawn independent r.v. following a normal distribution with unknown mean  $\mu_x$  and unknown variance  $\sigma_x^2$ . Same for  $(y_1, \dots, y_m)$  with  $\mu_y, \sigma_y^2$ . Also assume the unknown variances are equal:  $\sigma_y^2 = \sigma_x^2 = \sigma^2$ .

$$L(\theta | x) = f(\vec{x} | \theta) = \prod_{i=1}^n f(x_i | \theta) \quad (\text{since } x_i \text{ are independent})$$

$$\Rightarrow L(\theta | x) = \prod_{i=1}^n \left( \frac{1}{\theta} \right) = \begin{cases} \frac{1}{\theta^n} & \text{for } 0 \leq x_i \leq \theta \\ 0 & \text{otherwise} \end{cases}$$

$$f_x(x) = \begin{cases} \frac{1}{\theta - a} = \frac{1}{\theta} & , x \in [a, \theta] \\ 0 & , \text{otherwise} \end{cases}$$

$$\frac{dL(\theta | x)}{d\theta} = \frac{d e^{-n \ln \theta}}{d\theta} = -n \theta^{-n-1} = -\frac{n}{\theta^{n+1}}$$

which is ~~always~~ always negative (as  $n > 0, \theta > 0$ )

Now find maximum.

$$\frac{dL(\theta | x)}{d\theta} < 0, \forall \theta \Rightarrow L(\theta | x) \text{ is decreasing,}$$

~~max value~~ ~~at~~ ~~start~~ ~~of~~ ~~interval~~  ~~$[0, \theta]$~~   ~~$\Rightarrow$~~   ~~$\theta \rightarrow 0$~~

$L(\theta|X)$  is decreasing  $\Rightarrow$  the MLE will be the smallest value of  $\theta$  such that  $\theta \geq x_i$ . This value is  $\theta = \max(x_1, \dots, x_n)$

$$\Rightarrow \underline{\hat{\theta} = \max(x_1, \dots, x_n)}$$