

# Homework 1

DSA1101

Introduction to Data Science

September 7, 2018

**Problem 1 (10 points).** Suppose we have two data vectors  $x = c(x_1, x_2, \dots, x_n)$  and  $y = c(y_1, y_2, \dots, y_n)$ , both of length  $n$ . Remember in lecture that their means are given by  $\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$  and  $\bar{y} = \frac{1}{n} \sum_{i=1}^n y_i$  respectively.

(a) Show that  $\frac{1}{n} \sum_{i=1}^n (x_i - \bar{x}) = 0$ .

$$\begin{aligned} \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x}) &= \frac{1}{n} \sum_{i=1}^n x_i - \frac{1}{n} n \bar{x} \\ &= \bar{x} - \bar{x} = 0. \end{aligned}$$

(b) Show that  $\frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2 = \frac{1}{n} \sum_{i=1}^n x_i^2 - \bar{x}^2$ .

$$\begin{aligned} \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2 &= \frac{1}{n} \sum_{i=1}^n (x_i^2 - 2x_i\bar{x} + \bar{x}^2) \\ &= \frac{1}{n} \sum_{i=1}^n x_i^2 - 2\bar{x} \frac{1}{n} \sum_{i=1}^n x_i + \frac{1}{n} n \bar{x}^2 \\ &= \frac{1}{n} \sum_{i=1}^n x_i^2 - 2\bar{x}^2 + \bar{x}^2 = \frac{1}{n} \sum_{i=1}^n x_i^2 - \bar{x}^2. \end{aligned}$$

**Problem 2 (15 points).** Suppose we have two data vectors  $x = c(x_1, x_2, x_3) = c(1, 2.5, 4)$  and  $y = c(y_1, y_2, y_3) = (0, 3, 3)$ . We postulate the following simple linear relationship between  $y$  and  $x$ :

$$y \approx \beta_0 + \beta_1 x.$$

- (a) Complete the following table based on the 3 data points, leaving your answers in terms of  $\beta_0$  and  $\beta_1$ :

$i$	$x_i$	$y_i$	$\beta_0 + \beta_1 x_i$	residual: $e_i = y_i - (\beta_0 + \beta_1 x_i)$
1	1	0	$\beta_0 + \beta_1$	$-(\beta_0 + \beta_1)$
2	2.5	3	$\beta_0 + 2.5\beta_1$	$3 - (\beta_0 + 2.5\beta_1)$
3	4	3	$\beta_0 + 4\beta_1$	$3 - (\beta_0 + 4\beta_1)$

- (b) Write down an expression for the Residual Sum of Squares,  $RSS = e_1^2 + e_2^2 + e_3^2$ , leaving your answer in terms of  $\beta_0$  and  $\beta_1$ :

$$RSS = [\beta_0 + \beta_1]^2 + [3 - (\beta_0 + 2.5\beta_1)]^2 + [3 - (\beta_0 + 4\beta_1)]^2$$

- (c) Based on the  $RSS$  given in (b), derive and write down the *least squares* estimates  $\hat{\beta}_0$  and  $\hat{\beta}_1$ .

$$\begin{aligned} \frac{\partial RSS}{\partial \beta_0} &= 2[\beta_0 + \beta_1] + 2[3 - (\beta_0 + 2.5\beta_1)](-1) + 2[3 - (\beta_0 + 4\beta_1)](-1) \\ &= -12 + 6\beta_0 + 15\beta_1 = 0. \end{aligned}$$

$$\begin{aligned} \frac{\partial RSS}{\partial \beta_1} &= 2[\beta_0 + \beta_1] + 2[3 - (\beta_0 + 2.5\beta_1)](-2.5) + 2[3 - (\beta_0 + 4\beta_1)](-4) \\ &= -39 + 15\beta_0 + 46.5\beta_1 = 0. \end{aligned}$$

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$$\begin{aligned} -39 + 15\left(2 - \frac{15}{6}\beta_1\right) + 46.5\beta_1 &= 0 \\ -9 + 9\beta_1 &= 0 \end{aligned}$$

$\rightarrow$

$$\hat{\beta}_1 = 1$$

$$\hat{\beta}_0 = -0.5$$