

Special Topics Comp Stat & Pro MAT5999 and Computational Stats & Prob. AIM 5002  
Written Assignment 10 (12.5 points)

5/2/22

Solutions to be returned by the beginning of class on Tuesday, 5/9.

1. A response  $Y$  is a function of three independent variables  $x_1$ ,  $x_2$ , and  $x_3$  that are related as follows:

$$Y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \epsilon,$$

Our goal is to fit this model to the  $n = 7$  data points shown in the accompanying table.

y	$x_1$	$x_2$	$x_3$
1	-3	5	-1
0	-2	0	1
0	-1	-3	1
1	0	-4	0
2	1	-3	-1
3	2	0	-1
3	3	5	1

- (a) (3 points) Find the least squares estimates of the model parameters  $\beta_i$ ,  $i = 1, 2, 3$ . Hint: You can use R to perform the necessary computations.
  - (b) (3.5 points) Estimate  $\mathbb{E}(Y)$  when  $x_1 = 1, x_2 = -3, x_3 = -1$ . Compare with the observed response in the original data. Why are these two not equal?
  - (c) (3 points) Give a 95% prediction interval for  $Y$  when  $x_1 = 1, x_2 = -3, x_3 = -1$ .
2. (3 points) **Correlation does not mean causation.** Assume that in a simple linear model, we can reject the null hypothesis  $\beta_1 = 0$  at very high confidence level. That is, we are confident that  $Y$  is correlated with  $X$ . It is important to remember that this does not necessarily mean that the increase of  $X$  *causes*  $Y$  to increase. E.g., if  $X$  is the number of sunglasses sold and  $Y$  is the amount of ice cream sold, we can easily find a correlation between  $X$  and  $Y$  but an increase in  $X$  does not cause an increase in  $Y$ . Indeed, there is a hidden variable, a.k.a. confounding variable (in this example being hot sunny weather) that causes both  $X$  and  $Y$  to increase.

Think about this and give 3 more examples where correlation does not mean causation (no need to give data, just examples. You can look up examples online).

Just for fun: <https://imgs.xkcd.com/comics/correlation.png>