

# Random Variables and Stochastic Process (AI5030)

Soumyajit Chatterjee  
AI22MTECH02005

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## Question 56 (2019)

There are two sets of observations on a random vector  $(X, Y)$ . Consider a simple linear regression model with an intercept for regressing  $Y$  on  $X$ . Let  $\hat{\beta}_i$  be the least square estimate of the regression coefficient obtained from the  $i$ th ( $i=1, 2$ ) set consisting of  $n_i$  observations ( $n_1, n_2 > 2$ ). Let  $\hat{\beta}_0$  be the least square estimate obtained from the pooled sample size  $n_1 + n_2$ . If it is known that  $\hat{\beta}_1 > \hat{\beta}_2 > 0$ , which of the following statements is true ?

1.  $\hat{\beta}_2 < \hat{\beta}_0 < \hat{\beta}_1$
2.  $\hat{\beta}_0$  may lie outside  $(\hat{\beta}_2, \hat{\beta}_1)$  but cannot exceed  $\hat{\beta}_1 + \hat{\beta}_2$
3.  $\hat{\beta}_0$  may lie outside  $(\hat{\beta}_2, \hat{\beta}_1)$  but cannot be negative
4.  $\hat{\beta}_0$  can be negative

## Solution

Let the straight line estimated on a vector  $(X, Y)$  of sample size  $n_1, n_2$  and  $n_1 + n_2$  be:

$$y_1 = \hat{\beta}_1 x_1 \quad (1)$$

$$y_2 = \hat{\beta}_2 x_2 \quad (2)$$

$$y = \hat{\beta}_0 x \quad (3)$$

The linear regression coefficient on the vector  $(X, Y)$  is given by the formula:

$$\hat{\beta} = (X^T X)^{-1} (X^T Y)$$

The term  $(X^T X)^{-1}$  can be written as the square of  $L_2$  norm of  $X$ . Therefore, the above equation can be re-written as:

$$\hat{\beta} = \frac{X^T Y}{\|X\|^2}$$

Therefore, from equation 1 and 2 we can write that:

$$\hat{\beta}_1 = \frac{X_1^T Y_1}{\|X_1\|^2} = \frac{p_1}{q_1} \quad (4)$$

$$\hat{\beta}_2 = \frac{X_2^T Y_2}{\|X_2\|^2} = \frac{p_2}{q_2} \quad (5)$$

Since,  $\hat{\beta}_0$  is the estimator for stacked sample size  $n_1 + n_2$  and can be given by:

$$\hat{\beta}_0 = \frac{[X_1^T \ X_2^T] \begin{bmatrix} Y_1 \\ Y_2 \end{bmatrix}}{\|X_1\|^2 + \|X_2\|^2} \quad (6)$$

Upon simplification we get:

$$\hat{\beta}_0 = \frac{X_1^T Y_1 + X_1^T Y_2 + X_2^T Y_1 + X_2^T Y_2}{\|X_1\|^2 + \|X_2\|^2} \quad (7)$$

Since, all the terms in the numerator are scalars, we can represent the above equation as:

$$\hat{\beta}_0 = \frac{p_1 + p_2 + r + s}{q_1 + q_2}$$

Where r is given as  $X_2^T Y_1$  and s is given as  $X_1^T Y_2$

In the question it is given that  $\hat{\beta}_1 > \hat{\beta}_2 > 0$  which implies:

$$\frac{p_1}{q_1} > \frac{p_2}{q_2} > 0$$

Since,  $q_1$  and  $q_2$  is the square of norm, it is always positive. Therefore, from the above equation we can say that  $p_1 > 0$  and  $p_2 > 0$ .

In the expression for  $\hat{\beta}_0$ ,

$$\hat{\beta}_0 = \frac{p_1 + p_2 + r + s}{q_1 + q_2}$$

it can be greater than  $\hat{\beta}_1$  and  $\hat{\beta}_2$  if r and s are also positive. Therefore, option (1) is incorrect.

If we evaluate the expression of  $\hat{\beta}_1 + \hat{\beta}_2$  it comes out to be:

$$\hat{\beta}_1 + \hat{\beta}_2 = \frac{p_1 + p_2}{q_1 + q_2}$$

But the equation for  $\hat{\beta}_0$  is:

$$\hat{\beta}_0 = \frac{p_1 + p_2 + r + s}{q_1 + q_2}$$

Therefore,

$$\frac{p_1 + p_2 + r + s}{q_1 + q_2} > \frac{p_1 + p_2}{q_1 + q_2}$$

only if  $r$  and  $s$  take positive values. Therefore, option (2) is incorrect.  
In the expression of  $\hat{\beta}_0$

$$\hat{\beta}_0 = \frac{p_1 + p_2 + r + s}{q_1 + q_2}$$

if  $r$  and  $s$  are negative then the numerator of the above expression can be negative.  
Therefore, option (3) is also incorrect.

So, from the above conclusions drawn we can say that option (4) is the correct option.