

Assignment 0515

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Given the formula of least square as $y = Ax + B$, we want to find the smallest of deviation between estimation data and the observed data.

We think the minimum of the distance should be minimum when the distance is squared.

$$\begin{aligned} & \Sigma(y_i - Ax_i - B)^2 \\ &= A^2 \Sigma x_i^2 + nB^2 + \Sigma y_i^2 - 2A \Sigma x_i y_i - 2B \Sigma y_i + 2AB \Sigma x_i \end{aligned}$$

Now, partial derivative with respect to A,

$$2A \Sigma x_i^2 - 2 \Sigma x_i y_i + 2B \Sigma x_i$$

then partial derivative with respect to B,

$$2nB - 2 \Sigma y_i + 2A \Sigma x_i$$

When these two are equal 0, the maximum will be given.

Now solve this simultaneous equation,

$$\begin{cases} 2A \Sigma x_i^2 - 2 \Sigma x_i y_i + 2B \Sigma x_i = 0 & (1) \\ 2nB - 2 \Sigma y_i + 2A \Sigma x_i = 0 & (2) \end{cases}$$

$$\begin{cases} A \Sigma x_i^2 - \Sigma x_i y_i + B \Sigma x_i = 0 & (3) \\ nB - \Sigma y_i + A \Sigma x_i = 0 & (4) \end{cases}$$

Then,

$$B = \frac{\Sigma y_i - A \Sigma x_i}{n} \quad (5)$$

Inserting this to (3),

$$A \Sigma x_i^2 - \Sigma x_i y_i + \frac{\Sigma y_i - A \Sigma x_i}{n} \Sigma x_i = 0$$

$$nA \Sigma x_i^2 - n \Sigma x_i y_i + \Sigma y_i - A(\Sigma x_i)^2 = 0$$

$$A = \frac{n \sum x_i y_i - \sum x_i \sum y_i}{n \sum x_i^2 - (\sum x_i)^2},$$

Inserting this to (5)

$$\begin{aligned} B &= \frac{1}{n} (\sum y_i - A \sum x_i) \\ &= \frac{1}{n} \left(\sum y_i - \sum x_i \left(\frac{n \sum x_i y_i - \sum x_i \sum y_i}{n \sum x_i^2 - (\sum x_i)^2} \right) \right) \\ &= \frac{1}{n} \frac{n \sum x_i^2 \sum y_i - (\sum x_i)^2 \sum y_i - n \sum x_i \sum x_i y_i + (\sum x_i)^2 \sum y_i}{n \sum x_i^2 - (\sum x_i)^2} \\ &= \frac{\sum x_i^2 \sum y_i - n \sum x_i \sum x_i y_i}{n \sum x_i^2 - (\sum x_i)^2} \end{aligned}$$

Therefore given (x_i, y_i) , we can calculate the gradients and intercepts from formula.