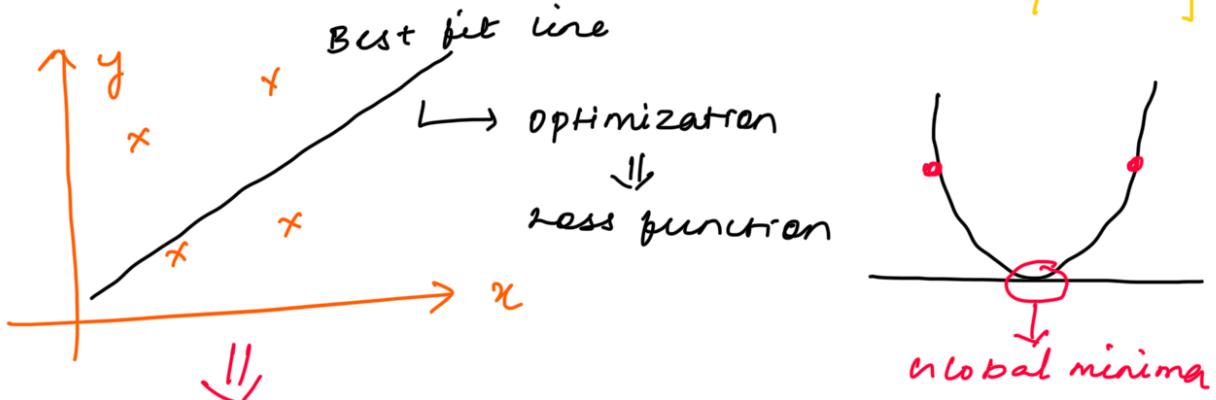


Linear Regression using OLS { ordinary least square }



OLS — can I apply some formula
and calculate B_0 (Intercept)
and B_1 (Coefficient)

Aim of OLS is to reduce the errors between the true points and predicted points

Ordinary Least Square \uparrow Error \downarrow MSE

$$S(B_0, B_1) = \frac{1}{n} \sum_{i=1}^n (y_i - B_0 - B_1 x_i)^2$$

\Downarrow

find B_0 and B_1

$$\frac{\partial S}{\partial B_0} (B_0, B_1) = \frac{2}{n} \sum_{i=1}^n (y_i - B_0 - B_1 x_i)^{2-1}$$

\Downarrow

with respect to B_0

$$= (0 - 1 - 0)$$

$$= -2 \sum_{i=1}^n (y_i - B_0 - B_1 x_i) = 0 \quad \textcircled{1}$$

$$n \quad i=1$$

$$\frac{\partial s}{\partial \beta_1} (\beta_0, \beta_1) = \frac{2}{n} \sum_{i=1}^n (y_i - \beta_0 - \beta_1 x_i)$$

↓,

with respect to β_1
 $(-x_i)$

$$= \frac{2}{n} \sum_{i=1}^n (y_i - \beta_0 - \beta_1 x_i) (-x) = 0$$

$$= -\frac{2}{n} \sum_{i=1}^n (y_i - \beta_0 - \beta_1 x_i) (x_i) = 0 \rightarrow ②$$

Eq → ①

$$-\frac{2}{n} \sum_{i=1}^n (y_i - \beta_0 - \beta_1 x_i) = 0$$

↓,

$$-\sum_{i=1}^n (y_i - \beta_0 - \beta_1 x_i) = 0$$

$$-\sum_{i=1}^n y_i + n \beta_0 + \beta_1 \sum_{i=1}^n x_i = 0$$

$$n \beta_0 + \beta_1 \sum_{i=1}^n x_i = \sum_{i=1}^n y_i$$

$$n - \sum_{i=1}^n y_i = \beta_1 \sum_{i=1}^n x_i$$

$$P_0 = \frac{\sum_{i=1}^n y_i}{n}$$

$$\boxed{P_0 = \bar{y} - P_1 \bar{x}} \Rightarrow \text{Intercept}$$

Eq — ②

$$-\frac{2}{n} \sum_{i=1}^n (y_i - P_0 - P_1)(x_i) = 0$$

↓

$$\sum_{i=1}^n (y_i - P_0 - P_1)(x_i) = 0$$

↓

$$\sum_{i=1}^n x_i y_i - P_0 \sum_{i=1}^n x_i - P_1 \sum_{i=1}^n (x_i)^2 = 0$$

↓/

$$\sum_{i=1}^n (x_i y_i - P_0 x_i - P_1 x_i^2) = 0$$

↓,

$$\text{Replace } P_0 = \bar{y} - P_1 \bar{x}$$

$$\sum_{i=1}^n (x_i y_i - (\bar{y} - P_1 \bar{x}) x_i - P_1 x_i^2) = 0$$

↓↓

$$\sum_{i=1}^n \left(y_i - \bar{y} + \beta_1 \bar{x} x_i - \beta_1 x_i^2 \right) = 0$$

↓↓

$$\sum_{i=1}^n \left(y_i - \bar{y} + \beta_1 \bar{x} - \beta_1 x_i^2 \right) x_i = 0$$

↓↓ .

$$\sum_{i=1}^n \left((y_i - \bar{y}) + \beta_1 (\bar{x} - x_i) \right) = 0$$

↓↓

$$\sum_{i=1}^n (y_i - \bar{y}) + \sum_{i=1}^n \beta_1 (\bar{x} - x_i) = 0$$

↓↓

$$\sum_{i=1}^n \beta_1 (\bar{x} - x_i) = - \sum_{i=1}^n (y_i - \bar{y})$$

↓↓

$$\beta_1 = \frac{- \sum_{i=1}^n (y_i - \bar{y})}{\sum_{i=1}^n (\bar{x} - x_i)}$$

↓↓

$$\beta_1 = \frac{\sum_{i=1}^n (y_i - \bar{y})}{\sum_{i=1}^n (\bar{x} - x_i)}$$

coefficient \Rightarrow

$$\beta_1 = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

OLS \approx Linear Regression (we learn)