

세상을 이해하는 통계학의 렌즈

REGRESSION

2020학년도 2학기

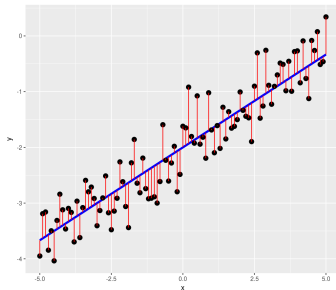
Ordinary Least Squares Regression

$$L(a, b) = \frac{1}{n} \sum_{i=1}^n (y_i - f(x_i))^2$$

$$\arg \min_{a, b} \frac{1}{n} \sum_{i=1}^n (y_i - (ax_i + b))^2$$

- 키워드: OLS 회귀분석, L2 손실함수, Objective Function, Argument, Minimize, a와 b의 함수, Mean Squared Error, 중심선(f(X))을 기준으로 데이터들이 흩어져 있는 정도, $y = ax + b$ 라는 가정.

OLS Regression

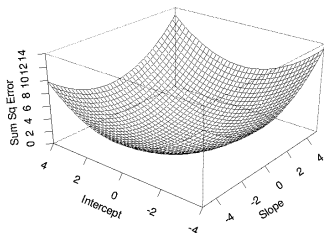


$$\arg \min_{a, b} \frac{1}{n} \sum_{i=1}^n (y_i - (ax_i + b))^2$$

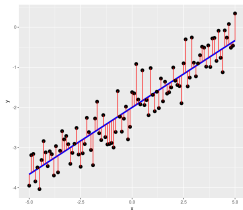
$$y = a^*x + b^*$$

OLS Regression

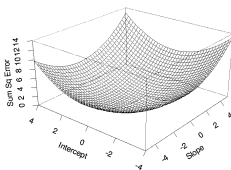
- Visualization of Quadratic Loss given a (slope) and b (intercept).



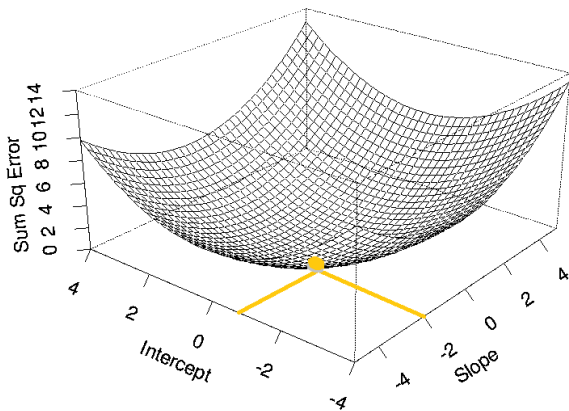
OLS Regression



$$\arg \min_{a, b} \frac{1}{n} \sum_{i=1}^n (y_i - (ax_i + b))^2$$



OLS Regression



OLS Regression

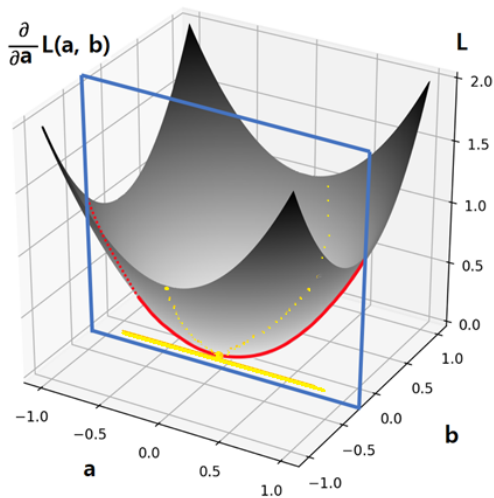
$$L(a, b) = \frac{1}{n} \sum_{i=1}^n (y_i - (ax_i + b))^2$$

$$\frac{\partial L(a, b)}{\partial a} \Big|_{a=a^*} = 0$$

$$\frac{\partial L(a, b)}{\partial b} \Big|_{b=b^*} = 0$$

OLS Regression

Visualisation of partial derivative of $L(a, b)$



OLS Regression

$$\arg \min_{a, b} \frac{1}{n} \sum_{i=1}^n (y_i - (ax_i + b))^2$$

$$L(a, b) = \frac{1}{n} \sum_{i=1}^n (y_i - (ax_i + b))^2$$

$$\frac{\partial L(a, b)}{\partial a} \Big|_{a=a^*} = 0$$

$$\frac{\partial L(a, b)}{\partial b} \Big|_{b=b^*} = 0$$

OLS Regression

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

$$b^* = \bar{y} - a^* \bar{x}$$

\bar{x} : x의 평균

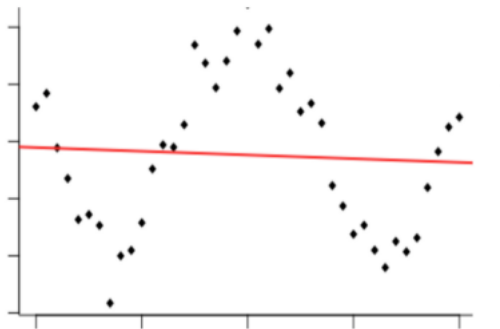
\bar{y} : y의 평균

$$f(\mathbf{x}) = a^* x + b^*$$

Pearson Correlation Coefficient

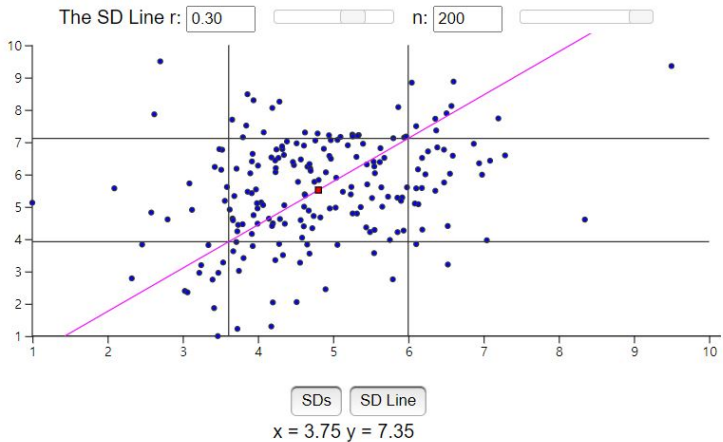
$$r_{XY} = \frac{\frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{n-1}}{\sqrt{\frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n-1}} \sqrt{\frac{\sum_{i=1}^n (y_i - \bar{y})^2}{n-1}}}$$

Pearson Correlation Coefficient



$$r = -0.028$$

Standard Deviation Line



OLS Regression

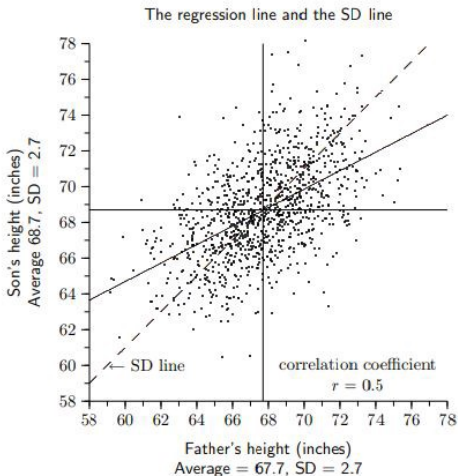
$$a^* = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2} = r_{XY} \cdot \frac{SD_y}{SD_x}$$

\bar{x} : x의 평균

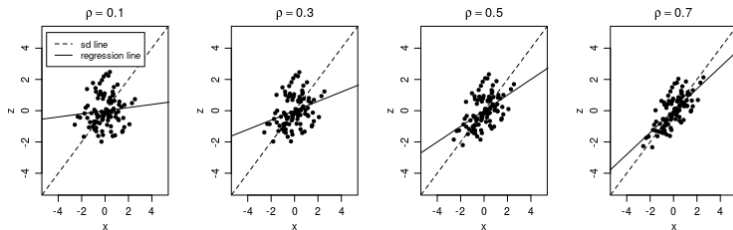
\bar{y} : y의 평균

$$f(\mathbf{x}) = a^*x + b^*$$

Meaning of “a(slope)” in OLS Regression Line



Meaning of “a(slope)” in OLS Regression Line



OLS Regression

$$a^* = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2} = r_{XY} \cdot \frac{SD_y}{SD_x}$$

\bar{x} : x의 평균

\bar{y} : y의 평균

$$f(\mathbf{x}) = a^*x + b^*$$

OLS Regression

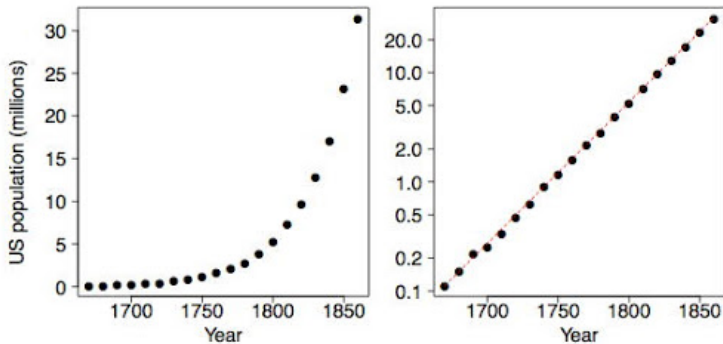
$$y = a^*x + b^*$$

$$\log y = a^*x + b^*$$

$$y = a^*\log x + b^*$$

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Log Transformation for Linearity



OLS Regression

$$y = a^*x + b^*$$

$$\log y = a^*x + b^*$$

$$y = a^*\log x + b^*$$

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