

Linear Regression

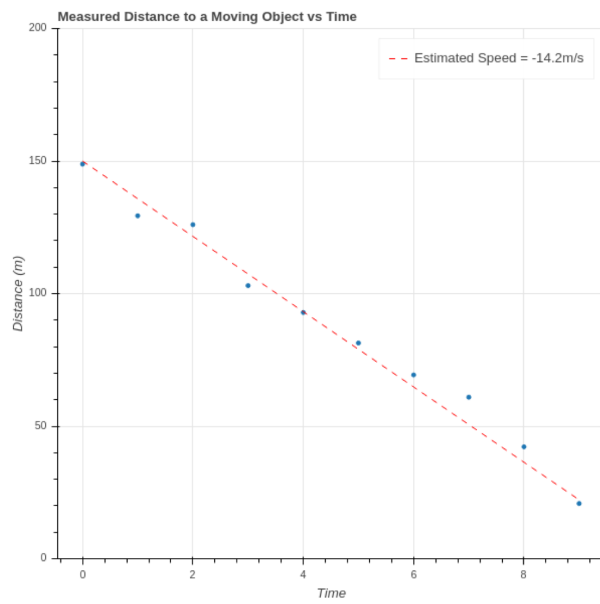
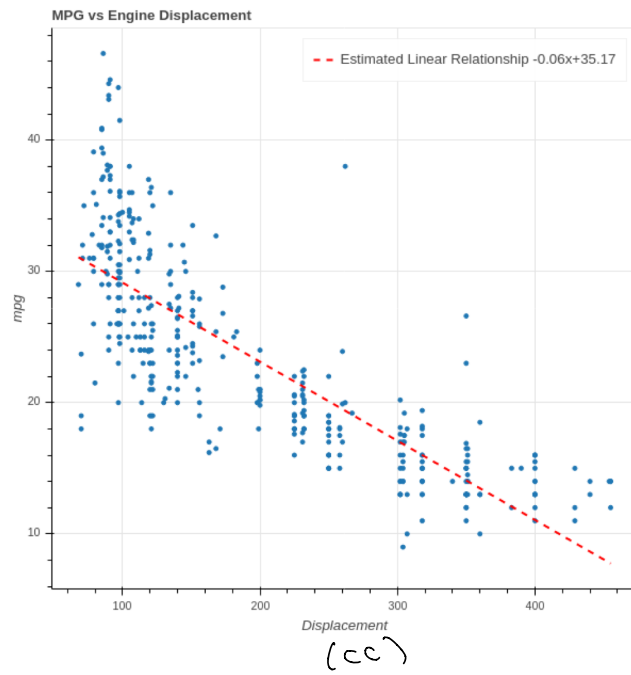


Figure 1: Physics Experiment



Linear Regression

Least Squares (Ordinary Least Squares or OLS)

Given data points $\{(x_i, y_i)\}_{i=1}^N$

Find slope m and intercept b so that

$$y_i \approx mx_i + b \quad \text{for } i=1, \dots, N$$

Mean squared error

$$MSE(m, b) = \frac{1}{N} \sum_{i=1}^N (y_i - mx_i - b)^2$$

'Best' choice of m and b are the values that minimize the MSE.

Use Calc to find them. Let $E = MSE(m, b)$

$$\text{Solve } \frac{\partial E}{\partial m} = 0 \quad \frac{\partial E}{\partial b} = 0 \quad b$$

find the minimum.

$$MSE = \frac{1}{N} \sum_{i=1}^N (y_i - mx_i - b)^2$$

a quadratic fn of m and b

$$MSE = Am^2 + Bmb + Cb^2 + Dm + Eb + F$$

$$\frac{\partial E}{\partial m}$$

$$E = \frac{1}{N} \sum (y_i - mx_i - b)^2$$

$$\begin{aligned} \frac{\partial E}{\partial m} &= \frac{1}{N} \sum_{i=1}^N 2(y_i - mx_i - b)[-x_i] \\ &= -2 \cdot \frac{1}{N} \sum_{i=1}^N (y_i x_i - mx_i^2 - bx_i) \\ &= -2 \cdot \left[\frac{1}{N} \sum y_i x_i - m \left(\frac{1}{N} \sum x_i^2 \right) - b \left(\frac{1}{N} \sum x_i \right) \right] \end{aligned}$$

$$\frac{1}{N} \sum x_i y_i = S_{xy} \quad \frac{1}{N} \sum x_i^2 = S_{xx} \quad \frac{1}{N} \sum x_i = \bar{x}$$

$$\frac{\partial E}{\partial m} = -2 [S_{xy} - m S_{xx} - b \bar{x}] = 0$$

$$\frac{\partial E}{\partial b}$$

$$E = \frac{1}{N} \sum_{i=1}^N (y_i - mx_i - b)^2$$

$$\begin{aligned} \frac{\partial E}{\partial b} &= \frac{1}{N} \sum_{i=1}^N 2(y_i - mx_i - b)[-1] \\ &= -\frac{2}{N} \left[\sum y_i - m \sum x_i - b \sum_{i=1}^N 1 \right] \\ &= -2 \left[\frac{1}{N} \sum y_i - m \left(\frac{1}{N} \sum x_i \right) - b \right] \end{aligned}$$

$$\bar{y} = \frac{1}{N} \sum y_i$$

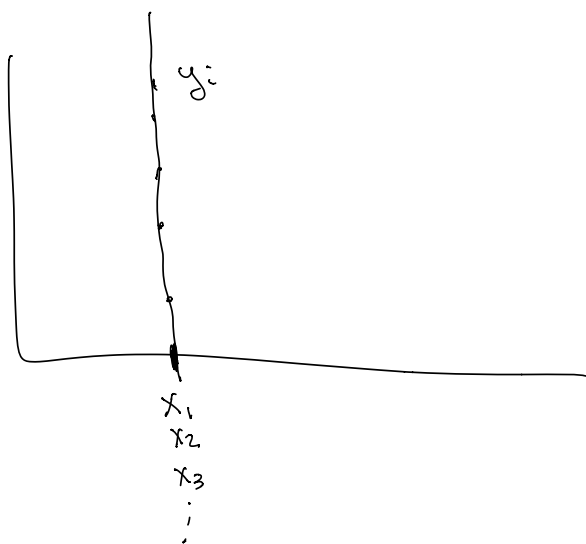
$$\frac{\partial E}{\partial b} = -2 [\bar{y} - m \bar{x} - b] = 0$$

$$\begin{aligned}
 \bar{y} &= m\bar{x} + b \\
 S_{xy} &= mS_{xx} + b\bar{x} \\
 \bar{x}\bar{y} &= m\bar{x}^2 + b\bar{x} \\
 \hline
 S_{xy} - \bar{x}\bar{y} &= m(S_{xx} - \bar{x}^2) \\
 m &= \frac{S_{xy} - \bar{x}\bar{y}}{S_{xx} - \bar{x}^2}
 \end{aligned}$$

$$\begin{aligned}
 \bar{y} &= m\bar{x} + b \\
 S_{xy} &= mS_{xx} + b\bar{x} \\
 \bar{x}S_{xy} &= m\bar{x}S_{xx} + b\bar{x}^2 \\
 S_{xx}\bar{y} &= m\bar{x}S_{xx} + bS_{xx} \\
 \hline
 \bar{x}S_{xy} - \bar{y}S_{xx} &= b(\bar{x}^2 - S_{xx}) \\
 b &= \frac{\bar{y}S_{xx} - \bar{x}S_{xy}}{S_{xx} - \bar{x}^2}
 \end{aligned}$$

$$S_{xx} \neq \bar{x}^2 !!$$

$$\frac{1}{N} \sum x_i^2 \neq \left[\frac{1}{N} \sum x_i \right]^2$$



$$N \sum x_i^2 = (\sum x_i)^2$$

if ^{and only if} all x_i are equal this time.

$$X = (x_1, \dots, x_N)$$

$$E = (1, \dots, 1)$$

$$\|X\|^2 = \sum x_i^2 \quad \|E\|^2 = N$$

$$(\sum x_i)^2 = |X \cdot E|^2$$

Cauchy-Schwarz:

$$|a \cdot b| \leq \|a\| \|b\|$$

with $=$ only if $a = \lambda b$ for some λ .

$$|X \cdot E|^2 \leq \|X\|^2 \|E\|^2$$

$$= \text{only if } X = \lambda E$$

$$(x_1, \dots, x_N) = \lambda (1, 1, \dots, 1)$$

$$\boxed{x_i = \lambda}$$