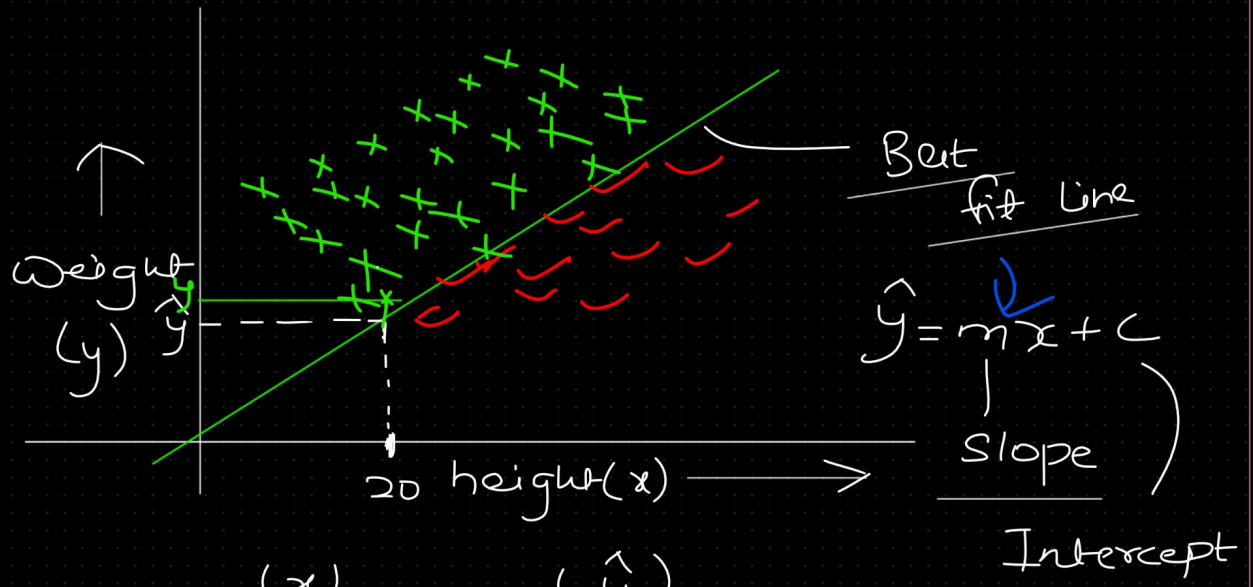


Linear Regression & Gradient Descent



(x)	(y)
Height	Weight
<div> <div>1</div> <div>1000</div> </div>	

Continuous Real number / numeric value

Error or Residual

$$(\hat{y} - y)$$

Predicted value

actual value

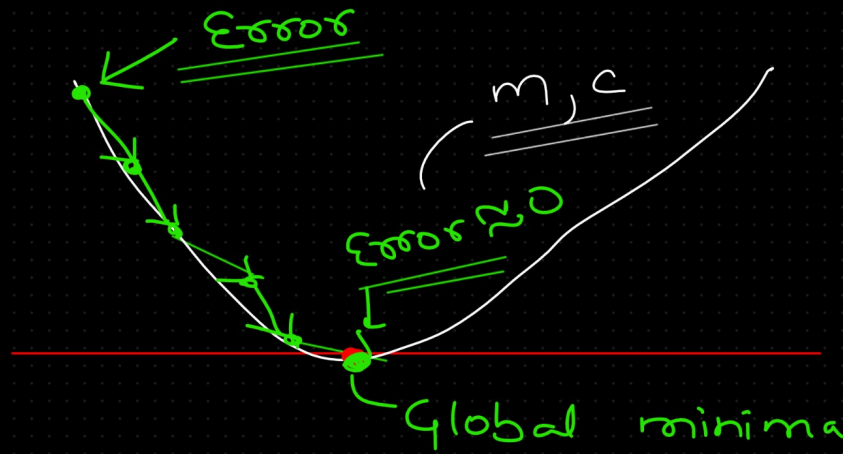
Best fit line

Predicted
value

$$\hat{y} = mx + c$$

$m, c \rightarrow$ quite accurate

Gradient Descent Optimizer



$\eta = 0.001$

Learning Rate

$$m_{\text{new}} = m_{\text{old}} - \eta \frac{\partial E}{\partial m}$$

$$c_{\text{new}} = c_{\text{old}} - \eta \frac{\partial E}{\partial c}$$

$$(a-b)^2 = a^2 + b^2 - 2ab$$

$$\underline{\underline{\text{Error} = \sum_{i=1}^n (y_i - \hat{y})^2 \downarrow \downarrow}}$$

$$\text{Error} = \sum_{i=1}^n (y_i - (mx + c))^2$$

$$\text{Error} = \sum_{i=1}^n (y_i^2 + (mx + c)^2 - 2y_i(mx + c))$$

$$\frac{\partial E}{\partial m} = \sum_{i=1}^n 2 (mx + c) \cdot x - 2y_i x$$

$$\frac{\partial E}{\partial m} = \sum_{i=1}^n 2 x_i (c + mx - y_i)$$

$$\frac{\partial E}{\partial m} = \sum_{i=1}^n 2 x_i (\hat{y} - y_i)$$

$$\frac{\partial E}{\partial c} = \sum_{i=1}^n (2(mx + c) - 2y_i)$$

$$\frac{\partial E}{\partial c} = \sum_{i=1}^n 2 (mx + c - y_i)$$

$$\frac{\partial E}{\partial c} = \sum_{i=1}^n 2 (\hat{y} - y_i)$$

$$\frac{\delta x^2}{\delta x} = 2x^{2-1} = \underline{2x}$$