

# PART 3: GRADIENT DESCENT MANUAL CALCULATION

Given information

model:  $y = mx + b$

initial  $m = -1$

initial  $b = 1$

Learning rate  $\alpha = 0.1$

Data points:  $(1, 3)$  and  $(3, 6)$

COST FUNCTION (mean square error)

$$J(m, b) = \frac{1}{n} \sum_{i=1}^n (y_i - (mx_i + b))^2$$

where  $n = 2$

Step 1: computing predictions with  $m$  and  $b$

$$\hat{y}_1 = mx_1 + b = (-1)(1) + 1 = 0$$

$$\hat{y}_2 = mx_2 + b = (-1)(3) + 1 = -2$$

Step 2: computing error

$$e_1 = y_1 - \hat{y}_1$$

$$e_2 = y_2 - \hat{y}_2$$

$$e_1 = 3 - 0 = 3$$

$$e_2 = 6 - (-2) = 8$$

Step 3: computing partial derivative  
of  $J$  with respect to  $m$

$$J = \frac{1}{2} ((3 - m - b)^2 + (6 - 3m - b)^2)$$

Differentiating the first term

$$\frac{\partial}{\partial m} (3 - m - b)^2$$

Let  $u = 3 - m - b$ , then  $\frac{du}{dm} = -1$   
chain rule:

$$\frac{d}{dm} (u^2) = 2u \cdot \frac{du}{dm} =$$

$$2(3 - m - b)(-1)$$

$$\therefore \frac{\partial}{\partial m} (3 - m - b)^2$$

$$= -2(3 - m - b)$$

Differentiating 2nd term

$$\frac{\partial}{\partial m} (6 - 3m - b)^2$$

Let  $v = 6 - 3m - b$ , then  $\frac{dv}{dm} = -3$

chain rule:

$$\frac{d}{dm} (v^2) = 2v \cdot \frac{dv}{dm} = 2(6 - 3m - b)(-3)$$

$$\frac{\partial}{\partial m} (6 - 3m - b)^2 = -6(6 - 3m - b)$$

Sum it up and multiply by  $\frac{1}{2}$

$$\frac{\partial J}{\partial m} = \frac{1}{2} [-2(3 - m - b) + (-6)(6 - 3m - b)]$$

$$\frac{1}{2} (-2(3 - m - b) - 6(6 - 3m - b))$$

Step 4: Substitute

First term

$$3 - m - b = 3 - (-1) - 1 = 2$$

$$3 + 1 - 1 = 3$$

2nd term

$$6 - 3m - b = 6 - 3(-1) - 1 = 6 + 3 - 1 \\ = 8$$

$$\therefore \frac{\partial J}{\partial m} = \frac{1}{2} [-2(3) - 6(8)]$$

$$= \frac{1}{2} [-6 - 48] = \frac{1}{2} (-54) = -27$$

Step 5: computing partial derivative of  $J$  with respect to  $b$

$$\text{let } v = 3 - m - b, \frac{dy}{db} = -1$$

chain rule:

$$2V \cdot \frac{dy}{db} = 2(3 - m - b)(-1) = \\ -2(3 - m - b)$$

$$\text{let } v = 6 - 3m - b, \frac{dy}{db} = -1$$

chain rule:

$$2V \cdot \frac{dv}{db} = 2(6 - 3m - b)(-1) = \\ -2(6 - 3m - b)$$

Sum and multiply by  $\frac{1}{2}$

$$\frac{\partial J}{\partial b} = \frac{1}{2} [-2(3 - m - b) - 2(6 - 3m - b)]$$

$$= \frac{1}{2} [-2(3 - m - b + 6 - 3m - b)]$$

$$= \frac{1}{2} [9 - 4m - 2b]$$

Substituting  $m$  and  $b$

where  $m = -1$  and  $b = 1$

$$9 - 4(-1) - 2(1) = 9 + 4 - 2 = 11$$

$$\therefore \frac{\partial J}{\partial b} = -11$$

Step 6 updating  $m$  and  $b$

$$m_{\text{new}} = m_{\text{old}} - \alpha \cdot \frac{\partial J}{\partial m} = \\ -1 - 0.1 \cdot (-27) \\ = -1 + 2.7 = 1.7$$

$$b_{\text{new}} = b_{\text{old}} - \alpha \cdot \frac{\partial J}{\partial b} =$$

$$1 - 0.1 \cdot (-11) \\ = 1 + 1 \cdot 1 = 2.1$$

Therefore the values of  $m$  and  $b$  are:

$$m = 1.7$$

$$b = 2.1$$

## Second Iteration:

$$m = 1.7$$

$$b = 2.1$$

## Computation of Predictions:

Points Provided: (1, 3) and (3, 6)

for  $x_1 = 1$

$\hat{y}_1 = mx_1 + b = (1.7)1 + 2.1 = 3.8$

for  $x_2 = 3$

$$\hat{y}_2 = mx_2 + b = (1.7)3 + 2.1 = 7.2$$

## Computation of Errors:

$$e_1 = \hat{y}_1 - y_1 = 3.8 - 3 = 0.8$$

$$e_2 = \hat{y}_2 - y_2 = 7.2 - 6 = 1.2$$

## Computation of Gradients:

$$J(m, b) = \frac{1}{n} \sum_{i=1}^n (y_i - (mx_i + b))^2$$

- $n = 2$  (as we have 2 points)

- Due to Constant Multiple rule - we will exempt ourselves from the  $\frac{1}{n}$  part of the formula.

$$\frac{d}{dx} [c \cdot f(x)] = c \cdot f'(x)$$

## Second

## Good term Iteration

$$J(m, b) = \frac{1}{2} [(3-m-b)^2 + (6-3m-b)^2] \quad \textcircled{2}$$

$$\frac{d(3-m-b)^2}{dm}$$

$$\text{let } t = 3-m-b; \quad \frac{dt}{dm} = 0-1-0 = -1$$

Chain rule:

$$\frac{d(t^2)}{dm} = 2t \cdot \frac{dt}{dm} = 2(3-m-b)(-1) = -2(3-m-b)$$

Differentiating Term 2:

$$\frac{d(6-3m-b)^2}{dm}; \quad \text{let } r = 6-3m-b$$

$$\frac{dr}{dm} = 0-3-0 = -3$$

$$\frac{d(r^2)}{dm} = 2r \cdot \frac{dr}{dm} = 2(6-3m-b)(-3)$$

$$\Rightarrow -6(6-3m-b)$$

$$\frac{dJ}{dm} = \frac{1}{2} \left[ -2(3-m-b) + (-6)(6-3m-b) \right]$$

Substitution:

$$\begin{aligned} \frac{dJ}{dm} &= \frac{1}{2} \left[ -2(3-1.7-2.1) + (-6)[(6-3(1.7)-2.1)] \right] \\ &= 4.4 \end{aligned}$$

Partial derivative with respect to  $b$ :

$$\frac{d}{db} (3 - m - b)^2, \text{ let } s = 3 - m - b$$

$$\frac{ds}{db} = 0 - 0 - 1 = -1$$

Chain rule:

$$\frac{d(s^2)}{dm} = 2s \cdot \frac{ds}{dm} = 2(3 - m - b)^{-1} = -2(3 - m - b)$$

Differentiating Term 2:

$$\frac{d}{db} (6 - 3m - b)^2, \text{ let } z = 6 - 3m - b$$

$$\frac{dz}{db} = 0 - 0 - 1 = -1$$

Chain rule:

$$\frac{d(z^2)}{db} = 2z \cdot \frac{dz}{db} = 2(6 - 3m - b)^{-1}$$

$$\frac{dJ}{db} = \frac{1}{2} [-2(3 - m - b) + (-2)(6 - 3m - b)]$$

$$= \frac{1}{2} [-2(3 - 1.7 - 2.4) + (-2)(6 - 3(1.7) - 2.1)] \\ = 2.$$

Updating  $m$  and  $b$

$$m_{\text{new}} = m_{\text{old}} - \alpha \frac{dJ}{dm}$$

(4)

$$b_{\text{new}} = b_{\text{old}} - \alpha \frac{dJ}{db}$$

$$m_{\text{new}} = 1.7 - (0.1) 4.4 = 1.26$$

$$b_{\text{new}} = 2.1 - (0.1) 2 = 1.9$$

$$\begin{array}{ll} m = 1.26 & b = 1.9 \end{array}$$

### Third Iteration :

Given:

$$\text{Model} \rightarrow y = mx + b$$

$$\begin{aligned} \text{Initial values} \\ m &= 1.26 \\ b &= 1.9 \end{aligned}$$

$\alpha = 0.1$	<u>Provided Points</u>
	(1, 3) and (3, 6)

For  $x_1 = 1$ :

$$\hat{y}_1 = mx_1 + b = (1.26)(1) + 1.9 = 3.16$$

For  $x_2 = 3$ :

$$\hat{y}_2 = mx_2 + b = (1.26)(3) + 1.9 = 5.68$$

Error computation:

$$e_1 = \hat{y}_1 - y_1 = 3.16 - 3 = 0.16$$

$$e_2 = \hat{y}_2 - y_2 = 5.68 - 6 = -0.32$$

Gradient computation:

$$J(m, b) = \frac{1}{n} \sum_{i=1}^n (y_i - (mx_i + b))^2 \quad \left| \begin{array}{l} \text{where:} \\ n=2 \text{ (no of points)} \end{array} \right.$$

① Now, let us compute the partial derivative of  $J$  with respect to  $m$ :

$$\bar{J} = \frac{1}{2} [(3-m-b)^2 + (6-3m-b)^2]$$

from differentiating the 1<sup>st</sup> term, we get:

$$\frac{d}{dm} (3-m-b)^2$$

- Let  $t = (3-m-b)$

$$\therefore \frac{dt}{dm} = -1$$

- Using the chain rule:

$$\begin{aligned} \frac{d(t^2)}{dm} &= 2t \cdot \frac{dt}{dm} = 2(3-m-b)(-1) \\ &= -2(3-m-b) \end{aligned}$$

From differentiating the 2<sup>nd</sup> term, we get:

$$\frac{d(6-3m-b)^2}{dm}$$

- Let  $u = 6-3m-b$ ;  $\frac{du}{dm} = -3$

- Using the chain rule:

$$\begin{aligned} \frac{d(u^2)}{dm} &= 2u \cdot \frac{du}{dm} = 2(6-3m-b)(-3) \\ &= -6(6-3m-b) \end{aligned}$$

Now:

$$\begin{aligned} \frac{dJ}{dm} &= \frac{1}{2} \left[ -2(3-m-b) + (-6)(6-3m-b) \right] \\ &= \frac{1}{2} \left[ -2(3-1.26-1.9) - 6(6-3(1.26)-1.9) \right] \\ &= \frac{1}{2} \left[ -2(-0.16) - 6(0.32) \right] \\ &= \boxed{-0.8} \end{aligned}$$

② Let us now compute the partial derivative of  $J$  with respect to  $b$ :

$$\text{Let } u = 3-m-b; \frac{du}{db} = -1$$

Using the chain rule:

$$\begin{aligned} 2u \cdot \frac{du}{db} &= 2(3-m-b)(-1) \\ &= -2(3-m-b) \end{aligned}$$

$$\text{Let } t = 6-3m-b; \frac{dt}{db} = -1$$

$$\begin{aligned} 2t \cdot \frac{dt}{db} &= 2(6-3m-b)(-1) \\ &= -2(6-3m-b) \end{aligned}$$

Now:

$$\begin{aligned} \frac{dJ}{db} &= \frac{1}{2} \left[ -2(3-m-b) + (-2)(6-3m-b) \right] \\ &= \frac{1}{2} \left[ -2(3-1.26-1.9) - 2(6-3(1.26)-1.9) \right] \\ &= \frac{1}{2} \left[ -2(-0.16) - 2(0.32) \right] = \boxed{-0.16} \end{aligned}$$

Let us now update the values of m and b

$$\left. \begin{array}{l} M_{\text{new}} = M_{\text{old}} - \alpha \frac{dJ}{dM} \\ b_{\text{new}} = b_{\text{old}} - \alpha \frac{dJ}{db} \end{array} \right\}$$

$$\textcircled{\#} M_{\text{new}} = 1.20 - (0.1)(-0.8)$$

$$= \underline{1.34}$$

$$\textcircled{\#} b_{\text{new}} = 1.9 - (0.1)(-0.16)$$

$$= \underline{1.916}$$

$$\therefore \boxed{\begin{array}{l} m = 1.34 \\ b = 1.916 \end{array}}$$

Given values from 3<sup>rd</sup> Iteration

$$\begin{aligned} m &= 1.34 & \text{Learning rate} &= 0.1 \\ b &= 1.916 & \text{Data points} &= (1, 3) (3, 6) \end{aligned}$$

Step 1: Computing predictions

$$\begin{aligned} \hat{y}_1 &= mx_1 + b = (1.34)(1) + 1.916 \\ &= 3.256 \end{aligned}$$

$$\begin{aligned} \hat{y}_2 &= mx_2 + b = (1.34)(3) + 1.916 \\ &= 5.936 \end{aligned}$$

Step 2: Computing errors

$$\begin{aligned} e_1 &= y_1 - \hat{y}_1 = 3 - 3.256 = -0.256 \\ e_2 &= y_2 - \hat{y}_2 = 6 - 5.936 = 0.064 \end{aligned}$$

Step 3: Partial derivative of  $J$  with respect to  $m$

using formula:

$$\frac{\partial J}{\partial m} = \frac{1}{2} \left[ -2(3-m-b) - 6(6-3m-b) \right]$$

Substitute:

$$m = 1.34, b = 1.916$$

Second term

$$\begin{aligned} 6 - 3m - b &= 6 - 3(1.34) - 1.916 \\ &= 6 - 4.02 - 1.916 = \\ &0.064 \end{aligned}$$

$$\begin{aligned} \text{so: } \frac{\partial J}{\partial m} &= \frac{1}{2} \left[ -2(-0.256) - \right. \\ &\quad \left. 6(0.064) \right] \\ &= \frac{1}{2} (0.512 - 0.384) \\ &= \frac{1}{2} (0.128) = 0.064 \end{aligned}$$

Step 4: partial derivative of  $J$  with respect to  $b$

$$\frac{\partial J}{\partial b} = -1 \left[ 9 - 4m - 2b \right]$$

Substituting  $m$  and  $b$   
 $m = 1.34, b = 1.916$

$$9 - 4(1.34) - 2(1.916) =$$

$$9 - 5.36 - 3.832 = -0.192$$

$$\text{so: } \frac{\partial J}{\partial b} = -1(-0.192) = 0.192$$

First term:

$$\begin{aligned} 3 - m - b &= 3 - 1.34 - 1.916 \\ &= 0.256 \end{aligned}$$

## MSE VS ITERATION GRAPH

Step 5; updating  $m$  and  $b$

$$m_{\text{new}} = m_{\text{old}} - \alpha \cdot \frac{\partial J}{\partial m}$$

$$= 1.34 - 0.1 \times (0.064)$$

$$= 1.34 - 0.0064$$

$$= 1.3336$$

$$b_{\text{new}} = b_{\text{old}} - \alpha \cdot \frac{\partial J}{\partial b}$$

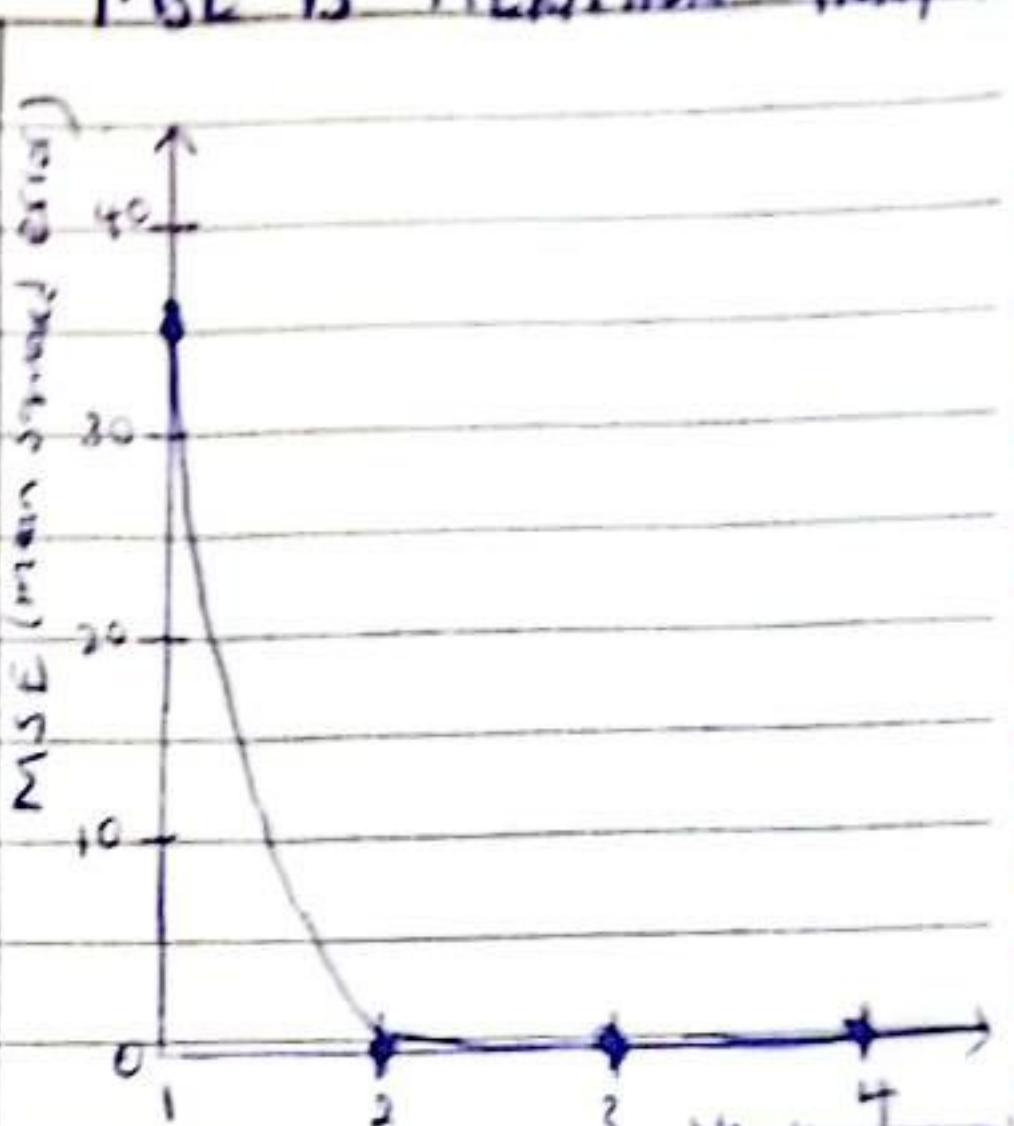
$$= 1.916 - 0.1 \times (0.192)$$

$$= 1.916 - 0.0192 = 1.8968$$

Result after 4th Iteration

$$m = 1.3336$$

$$b = 1.8968$$



KEY OBSERVATION

MSE decreased from 36.5 to 0.034816 - a reduction of over 99%.

Parameter evolution graph (m and b)

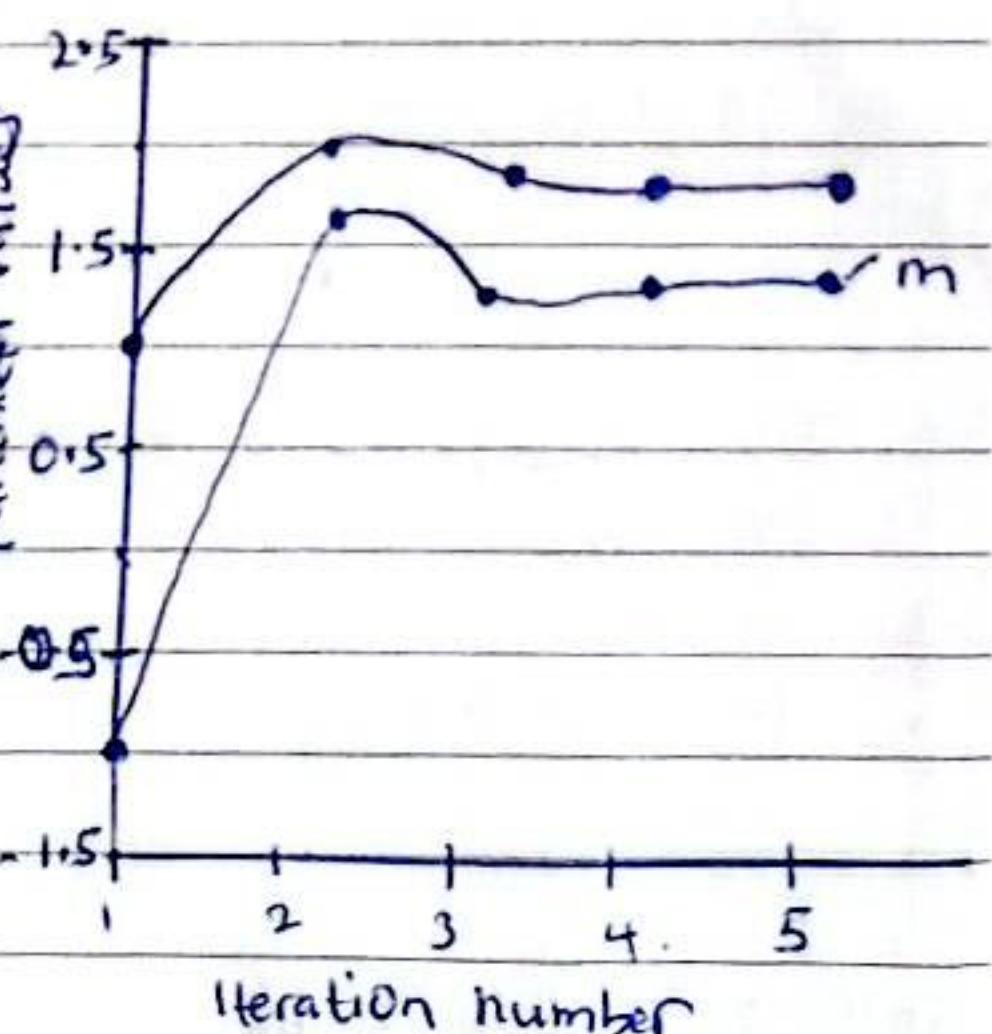
SUMMARY OF ALL ITERATIONS

1st Iteration:  $m = 1.7$   $b = 2.1$

2nd Iteration:  $m = 1.26$   $b = 1.9$

3rd Iteration:  $m = 1.34$   $b = 1.916$

4th Iteration:  $m = 1.3336$   $b = 1.8968$



The parameters are converging towards the true optimal values ( $m = 1.5$ ,  $b = 1.5$ ) and the error is decreasing with each iteration.