

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

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

$$\text{let } u = 3 - m - b, \text{ then } \frac{du}{dm} = -1$$

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

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

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

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

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

$$3 + 1 - 1 = 3$$

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

$$\text{let } u = 3 - m - b, \frac{du}{db} = -1$$

Chain rule:

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

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

chain rule:

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

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

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

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

$$= -1[9 - 4m - 2b]$$

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

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

$$= -1 - 0.1 \cdot (-27)$$

$$= -1 + 2.7 = 1.7$$

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

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

$$= 1 + 1.1 = 2.1$$

Therefore the values of m and b are:

$$m = 1.7$$

$$b = 2.1$$

Second Iteration:
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$$m = 1.7$$

$$b = 2.1$$

Computation of Prediction:  
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Points - Provided: (1,3) and (3,6)

For $x_1 = 1$
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$$\hat{y}_1 = mx_1 + b = (1.7)1 + 2.1 = 3.8$$

for  $x_2 = 3$   
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$$\hat{y}_2 = mx_2 + b = (1.7)3 + 2.1 = 7.2$$

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

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

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

- $n = 2$ (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

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

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

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

Chain rule:

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

Differentiating Term 2:

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

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

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

$$= -6(6-3m-b)$$

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

Substitution

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

Partial derivative with respect to b:  
 $\frac{d}{db} (3-m-b)^2$ , 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:

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

$$\begin{aligned} \frac{dJ}{db} &= \frac{1}{2} [-2(3-m-b) + (-2)(6-3m-b)] \\ &= \frac{1}{2} [-2(3-1.7-2.4) + (-2)(6-3(1.7)-2.4)] \\ &= 2. \end{aligned}$$

Updating m and b

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

(4)

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

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

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

$$m = 1.26 \quad b = 1.9$$



### Third Iteration:

Given:

Model  $\Rightarrow y = mx + b$

Initial values  $\left\{ \begin{array}{l} m = 1.26 \\ b = 1.9 \end{array} \right.$

$\alpha = 0.1$  | Provided Points  
(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  $m$ :

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

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:

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

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:

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

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  $b$ :

• let  $u = 3 - m - b$ ;  $\frac{du}{db} = -1$

Using the chain rule:

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

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

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

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  $b$ .

$$\left. \begin{aligned} M_{\text{new}} &= M_{\text{old}} - \alpha \frac{dJ}{dm} \\ b_{\text{new}} &= b_{\text{old}} - \alpha \frac{dJ}{db} \end{aligned} \right\}$$

$$\begin{aligned} \textcircled{*} M_{\text{new}} &= 1.26 - (0.1)(-0.8) \\ &= 1.34 \\ &= \underline{\underline{1.34}} \end{aligned}$$

$$\begin{aligned} \textcircled{*} b_{\text{new}} &= 1.9 - (0.1)(-0.16) \\ &= 1.916 \\ &= \underline{\underline{1.916}} \end{aligned}$$

$$\therefore \boxed{\begin{aligned} m &= 1.34 \\ b &= 1.916 \end{aligned}}$$



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

$$m = 1.34 \quad \text{Learning rate} = 0.1$$

$$b = 1.916 \quad \text{Data points} = (1/3)(36)$$

Step 1: computing predictions

$$\hat{y}_1 = mx_1 + b = (1.34)(1) + 1.916$$

$$= 3.256$$

$$\hat{y}_2 = mx_2 + b = (1.34)(5) + 1.916$$

$$= 5.936$$

Step 2: computing errors

$$e_1 = y_1 - \hat{y}_1 = 3 - 3.256 = -0.256$$

$$e_2 = y_2 - \hat{y}_2 = 6 - 5.936 = 0.064$$

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

using formula:

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

Substitute:

$$m = 1.34, b = 1.916$$

First term:

$$3 - m - b = 3 - 1.34 - 1.916$$

$$= 0.256$$

Second term

$$6 - 3m - b = 6 - 3(1.34) - 1.916$$

$$= 6 - 4.02 - 1.916 =$$

$$0.064$$

$$\text{So: } \frac{\partial J}{\partial m} = \frac{1}{2} [-2(-0.256) -$$

$$6(0.064)]$$

$$= \frac{1}{2} (0.512 - 0.384)$$

$$= \frac{1}{2} (0.128) = 0.064$$

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

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

Substituting  $m$  and  $b$

$$m = 1.34, b = 1.916$$

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

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

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



## 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

$$m = 1.3336$$

$$b = 1.8968$$

## SUMMARY OF ALL ITERATIONS

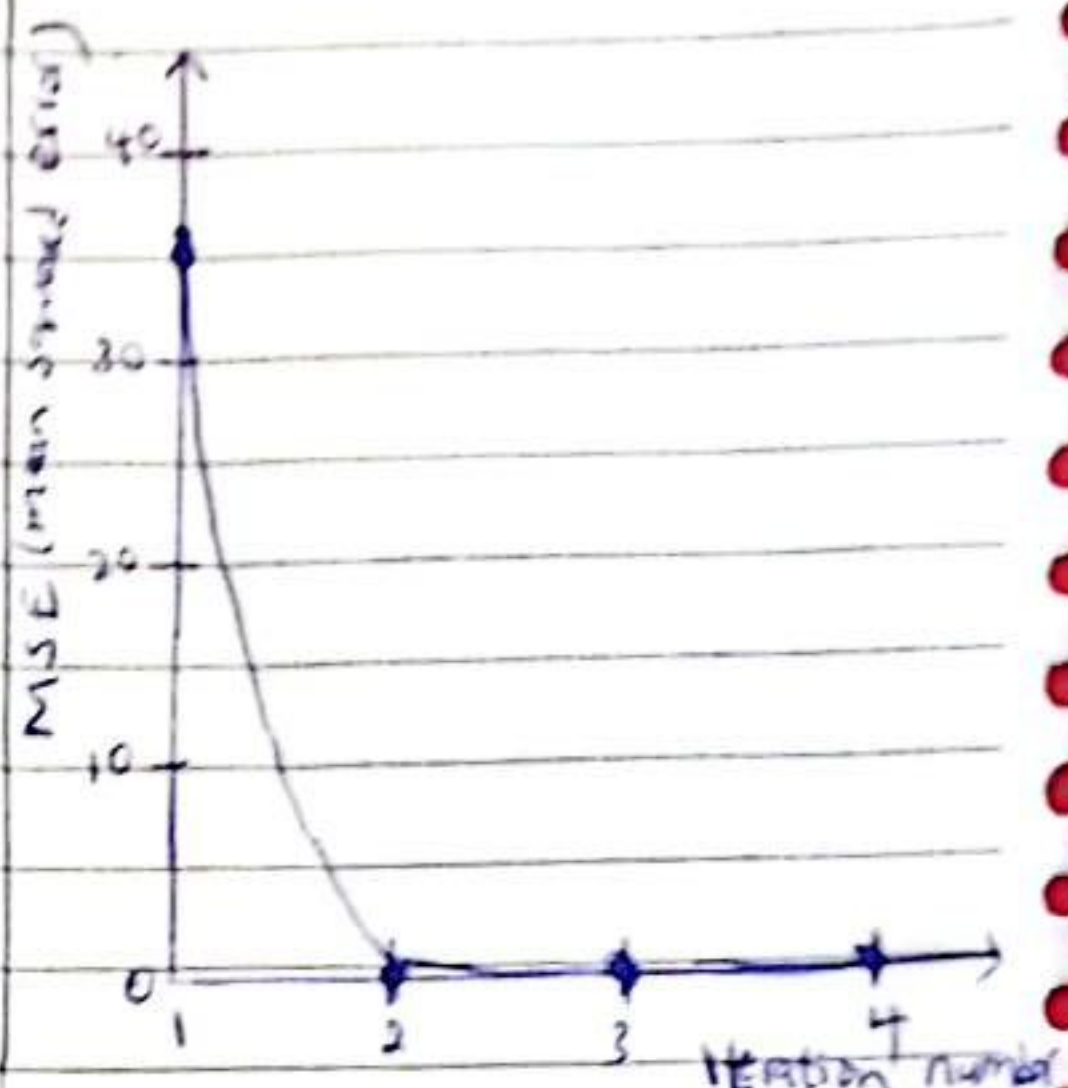
1st iteration:  $m = 1.7$   $b = 2.1$

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.



## KEY OBSERVATION

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

Parameter evolution graph (m and b)

