

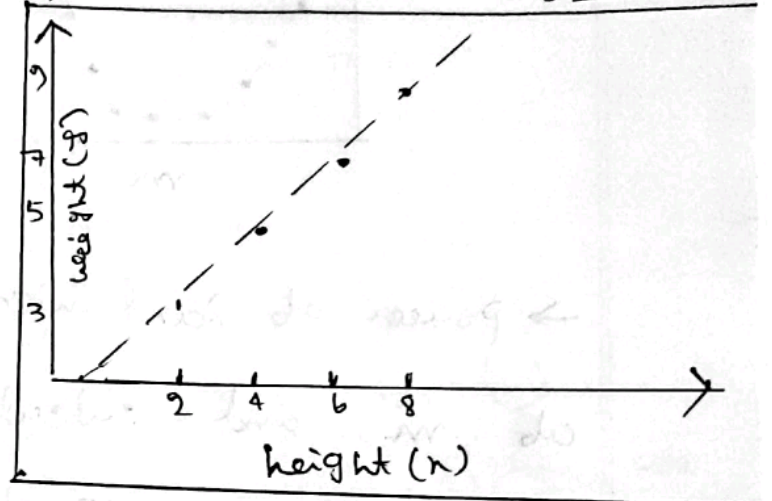
Gradient descent

DT: 30/07/25

Assume a dataset,

<u>height(x)</u>	<u>weight(y)</u>
2	3
4	5
6	7
8	9

Fig 1



Problem 01: The relationship between x and y can be defined by $y = mx$. So, you can predict y (weight) by plugging in x (height) in the equation,

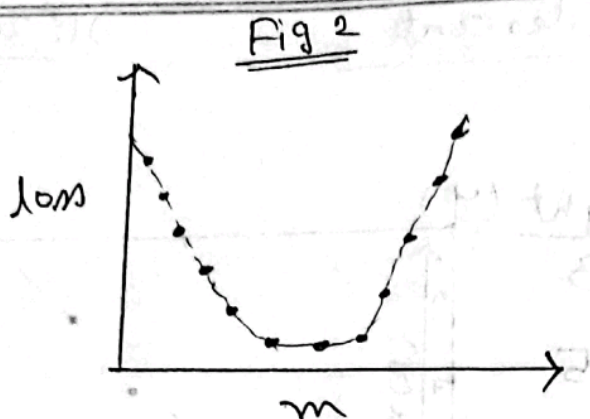
for example, $y(\text{weight}) = m \times 7$

Question is, how can we find m ???

→ Using Least Square Method [STA201]

→ Using Gradient Descent [CSE422/CSE437]

⊗ GD is an opt. algorithm, and have plenty of usage: minimizing error (loss function)



Remember: Our target is to find optimal value of m , so that loss is minimized.

→ power of Randomization: Randomly pick values of m , and calculate error (loss function). Plot them in a graph, and easily get the optimal solution.

⊗ This is how nature works. It always select the best candidate (Survival of the fittest).

But too randomness will not work. What if optimal solution is within the interval $[2, 5]$, but you are randomly trialing within $[7, 10]$?

→ Gradient descent solves this problem

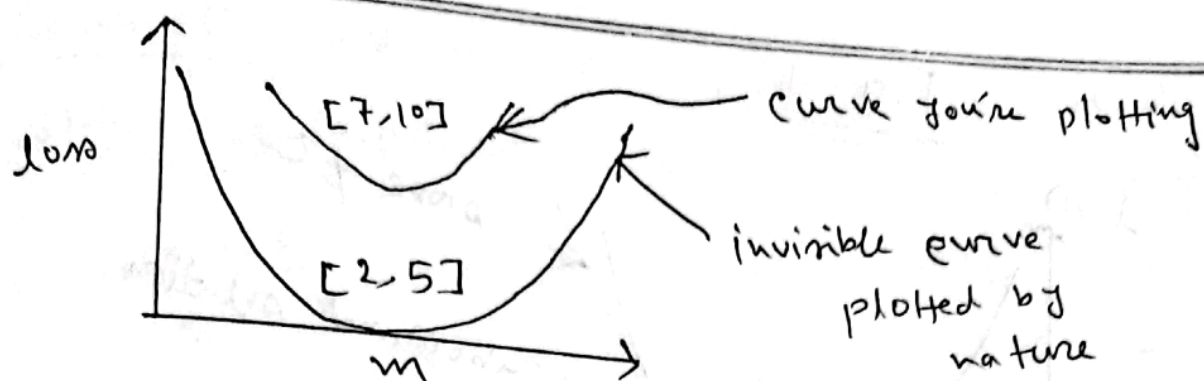


Fig 3

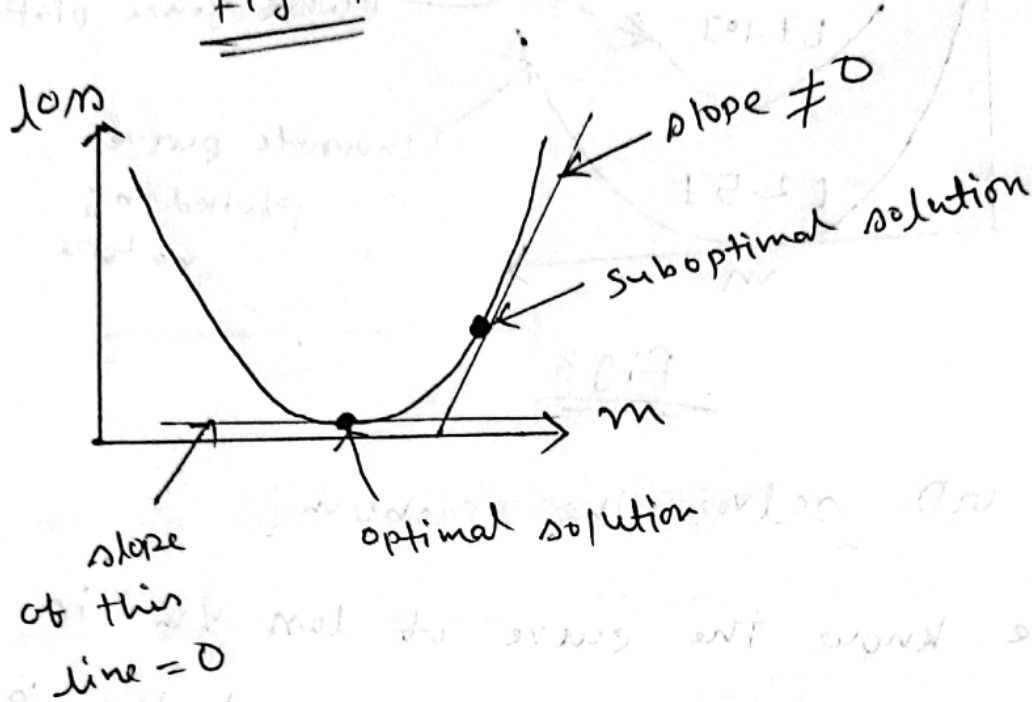
How GD solves the problem?

1. We know the curve of $\text{loss } v_m$ is U
2. If we can somehow reach at the bottom, we get our optimal solution.
3. Slope of a horizontal line is 0 (as $\Delta y = 0$),
so, when we reach our optimal solution:

Derivative of the loss function
with respect to the parameter (m)
is 0.

$$\text{or, } \frac{dl}{dm} = 0$$

Fig 4



Hence, if $\frac{dJ}{dm} \neq 0$, we keep updating m .

But how much do we update m ?

Update rule: $m = m - \frac{dJ}{dm} \times \alpha$, where,

α = learning rate

Simulation If, $m = 2.5$ and $\alpha = 0.1$

$$\text{new } m = 2.5 - (2.5 \times 0.1)$$

$$= 2.25 \text{ (updated by } 0.25)$$

If, $m = 0.15$,

$$\text{new } m = 0.15 - (0.15 \times 0.1)$$

$$= 0.135 \text{ (updated by } 0.015)$$

Exercise 013 Perform one epoch and find the value of (updated value) of m . $\alpha = 0.01$
 Use SSR as loss function.

height	weight
2	3
4	5
6	7
8	9

Soln: $y = mx$; Let, $m = 0.1$

Step 01 - Calculate loss function.

$$y_1 = 0.1 \times 2 = 0.2$$

$$\text{loss} = |(3 - 0.2)| = 2.8$$

← not necessary in math. Just to make it understand

Squared Residual for T1:

$$(3 - m \times 2)^2 = (3 - 2m)^2$$

Squared Residual for T2:

$$(5 - m \times 4)^2 = (5 - 4m)^2$$

for T3: $(7 - 6m)^2$

for T4: $(9 - 8m)^2$

Δ = Sum of squared residuals (SSR)

$$= T1_{sr} + T2_{sr} + T3_{sr} + T4_{sr}$$

$$= (3-2m)^2 + (5-4m)^2 + (7-6m)^2 + (9-8m)^2$$

Step 02: Calculate $\frac{d\Delta}{dm}$

$$\frac{d\Delta}{dm} = \frac{d}{dm} [(3-2m)^2 + (5-4m)^2 + (7-6m)^2 + (9-8m)^2]$$

$$= \frac{d}{dm} (3-2m)^2 + \frac{d}{dm} (5-4m)^2 + \frac{d}{dm} (7-6m)^2 + \frac{d}{dm} (9-8m)^2$$

$$= 2(3-2m) \cdot -2 + 2(5-4m) \cdot -4 + 2(7-6m) \cdot -6 + 2(9-8m) \cdot -8$$

$$= -12 \cdot 2 - 36 \cdot 8 - 76 \cdot 6 - 132 \cdot 8 = -233.6$$

Step 03: Calculate updated m .

$$m = \text{old } m - \frac{d\Delta}{dm} \times \alpha$$

$$= 0.1 - (-233.6 \times 0.01)$$

$$= 2.436 \text{ (Ans)}$$

$$y_1 = 2.436 \quad x_1 = 2.436 \times 2 = 4.872$$

$$\text{loss} = |(3 - 4.872)| = 1.87 < 2.8$$

↗
previous loss

Example 2: The following data can be modeled by: $y = m_1x_1 + m_2x_2 + c$. Find the first updated value of m_1 , m_2 , and c using Gradient Descent. Assume, $m_1 = 0.1$, $m_2 = 0.3$, $c = 1$, $\alpha = 0.00025$. Use SSR as loss f.

<u>age</u>	<u>year</u>	<u>salary (k)</u>
28	4	40
31	7	44
33	8	45

Solⁿ: Step 01: calculate SSR.

$$T1_{ssr} = (40 - 28m_1 - 4m_2 - c)^2$$

$$T2_{ssr} = (44 - 31m_1 - 7m_2 - c)^2$$

$$T3_{ssr} = (45 - 33m_1 - 8m_2 - c)^2$$

$$\begin{aligned} SSR = J = & (40 - 28m_1 - 4m_2 - c)^2 \\ & + (44 - 31m_1 - 7m_2 - c)^2 \\ & + (45 - 33m_1 - 8m_2 - c)^2 \end{aligned}$$

Step 02: Calculate $\frac{dl}{dm_1}$, $\frac{dl}{dm_2}$, $\frac{dl}{c}$

$$\begin{aligned}\frac{dl}{dm_1} &= \frac{d}{dm_1} [(40 - 28m_1 - 4m_2 - c)^2] \\ &+ \frac{d}{dm_1} [(44 - 31m_1 - 7m_2 - c)^2] \\ &+ \frac{d}{dm_1} [(45 - 33m_1 - 8m_2 - c)^2]\end{aligned}$$

$$\begin{aligned}&= 2(40 - 28m_1 - 4m_2 - c) \cdot (-28) \\ &+ 2(44 - 31m_1 - 7m_2 - c) \cdot (-31) \\ &+ 2(45 - 33m_1 - 8m_2 - c) \cdot (-33)\end{aligned} \quad \left. \vphantom{\begin{aligned} &= 2(40 - 28m_1 - 4m_2 - c) \cdot (-28) \\ &+ 2(44 - 31m_1 - 7m_2 - c) \cdot (-31) \\ &+ 2(45 - 33m_1 - 8m_2 - c) \cdot (-33) \end{aligned}} \right\} \text{Partial derivative}$$

$$= -1960 - 2343.6 - 2527.8 = -6831.4$$

Updating m_1 : $m_1' = m_1 - \left(\frac{dl}{dm_1} \alpha\right)$

$$= 0.1 - (-6831.4 \times 0.00025)$$

$$= \boxed{1.80}$$

$$\begin{aligned}\frac{dl}{dm_2} &= 2(40 - 28m_1 - 4m_2 - c) \cdot (-4) \\ &\quad + 2(44 - 31m_1 - 7m_2 - c) \cdot (-8) \\ &\quad + 2(45 - 33m_1 - 8m_2 - c) \cdot (-8)\end{aligned}$$

$$= -280 - 529.2 - 612.8 = -1422$$

$$m_2' = m_2 - \left(\frac{dl}{dm_2} \alpha \right) = 0.3 - (-1422 \times 0.00025) = \boxed{0.655}$$

$$\begin{aligned}\frac{dl}{dc} &= 2(40 - 28m_1 - 4m_2 - c) \cdot -1 \\ &\quad + 2(44 - 31m_1 - 7m_2 - c) \cdot -1 \\ &\quad + 2(45 - 33m_1 - 8m_2 - c) \cdot -1\end{aligned}$$

$$= \cancel{5.05} - 222.2$$

$$c' = c - \left(\frac{dl}{dc} \alpha \right) = 1 - (\cancel{4.05} - 222.2 \times 0.00025) = \boxed{1.05}$$

$$\text{So, updated } (m_1, m_2, c) = (1.80, 0.655, 1.05) \quad (\text{Ans})$$