

# Linear Regression In depth

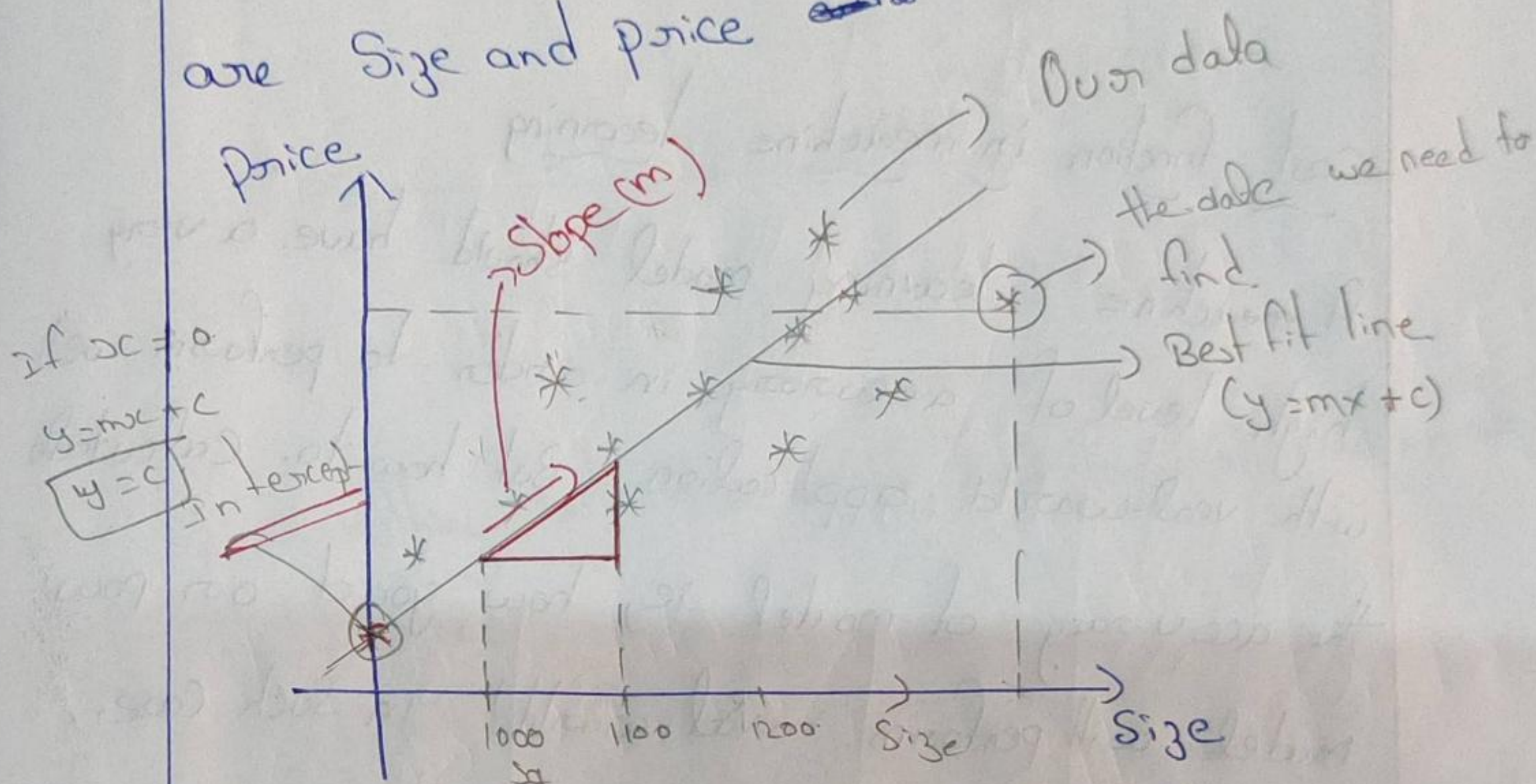
## Maths Intuition

$$y = mx + c \rightarrow \text{Best fit line}$$

$m = \text{slope}$

$c = \text{intercept (or) constant}$

Let's think we have a dataset with 2 features there are Size and price



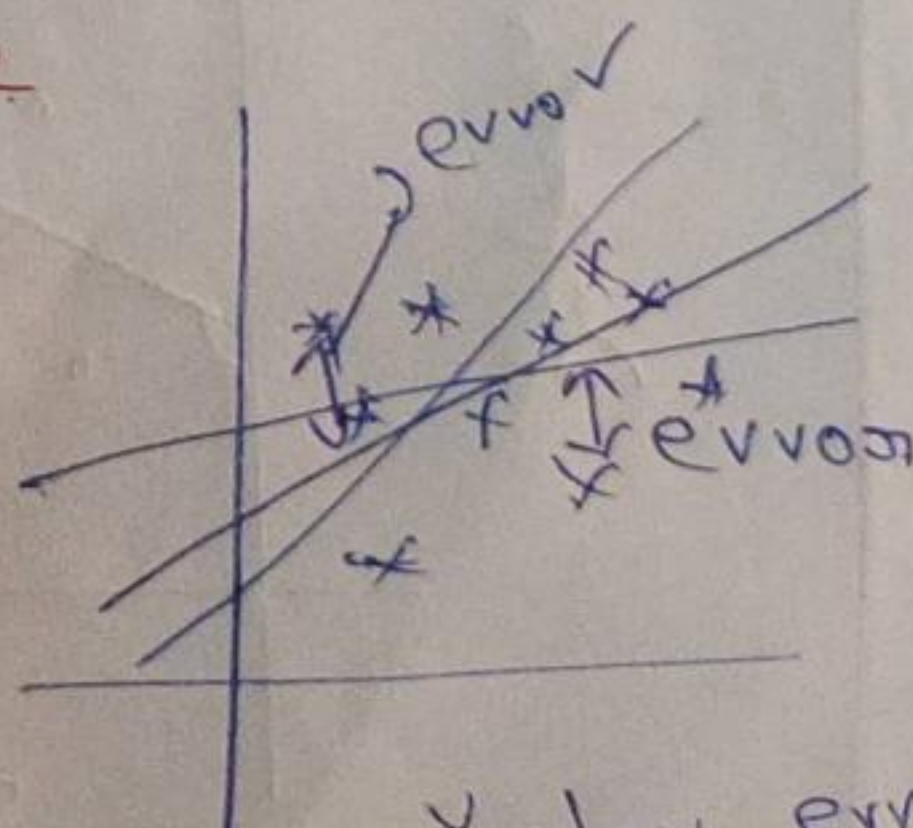
we are doing By Regression line. (or)  
Regression algorithm to create a Best fit line such  
this best fit line does prediction of my feature.

Size.

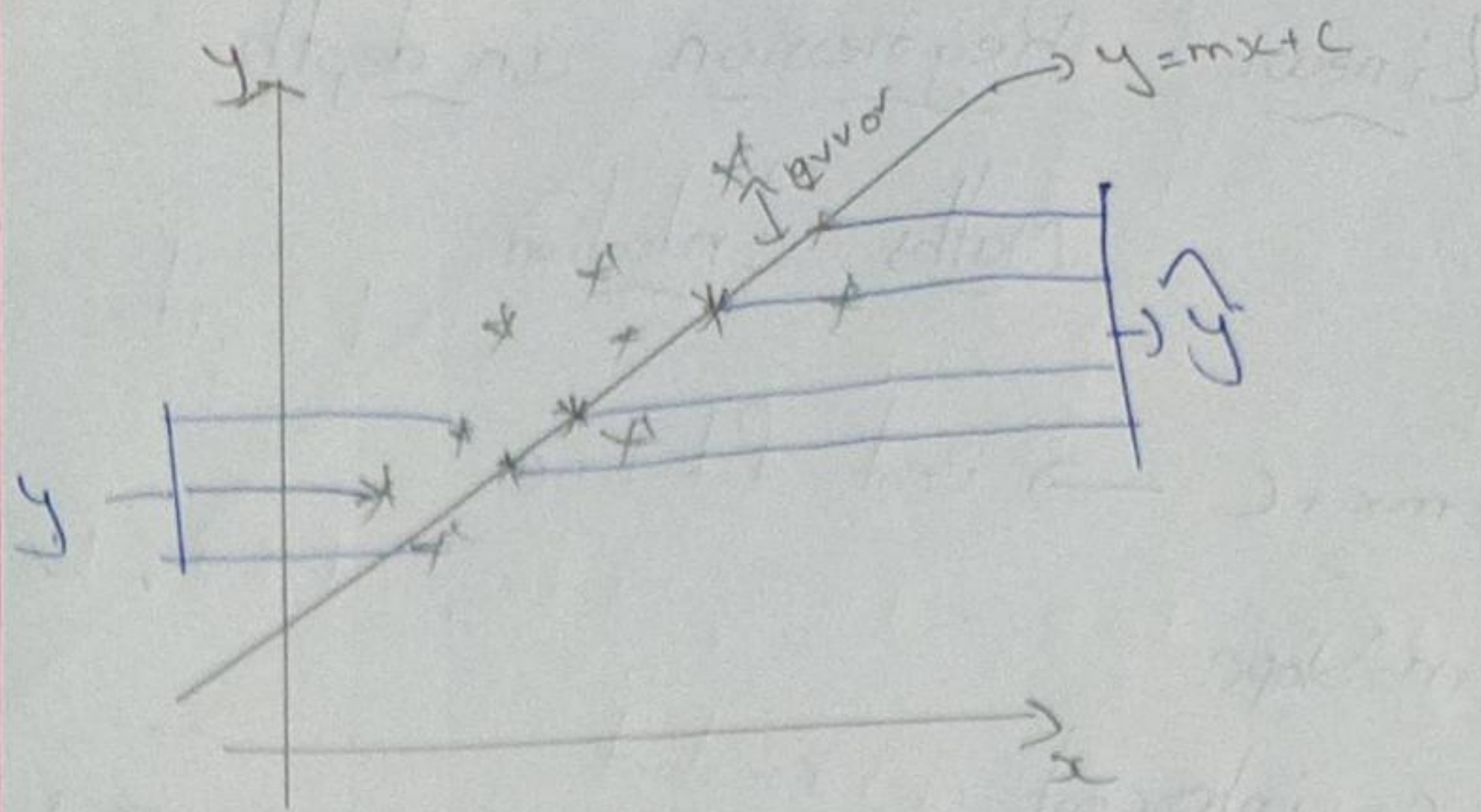
How we can fit the Best fit line

we can draw multi lines also

and we can try to mins. the error and we will add all the error and which best line gives less error that value is called minimal error



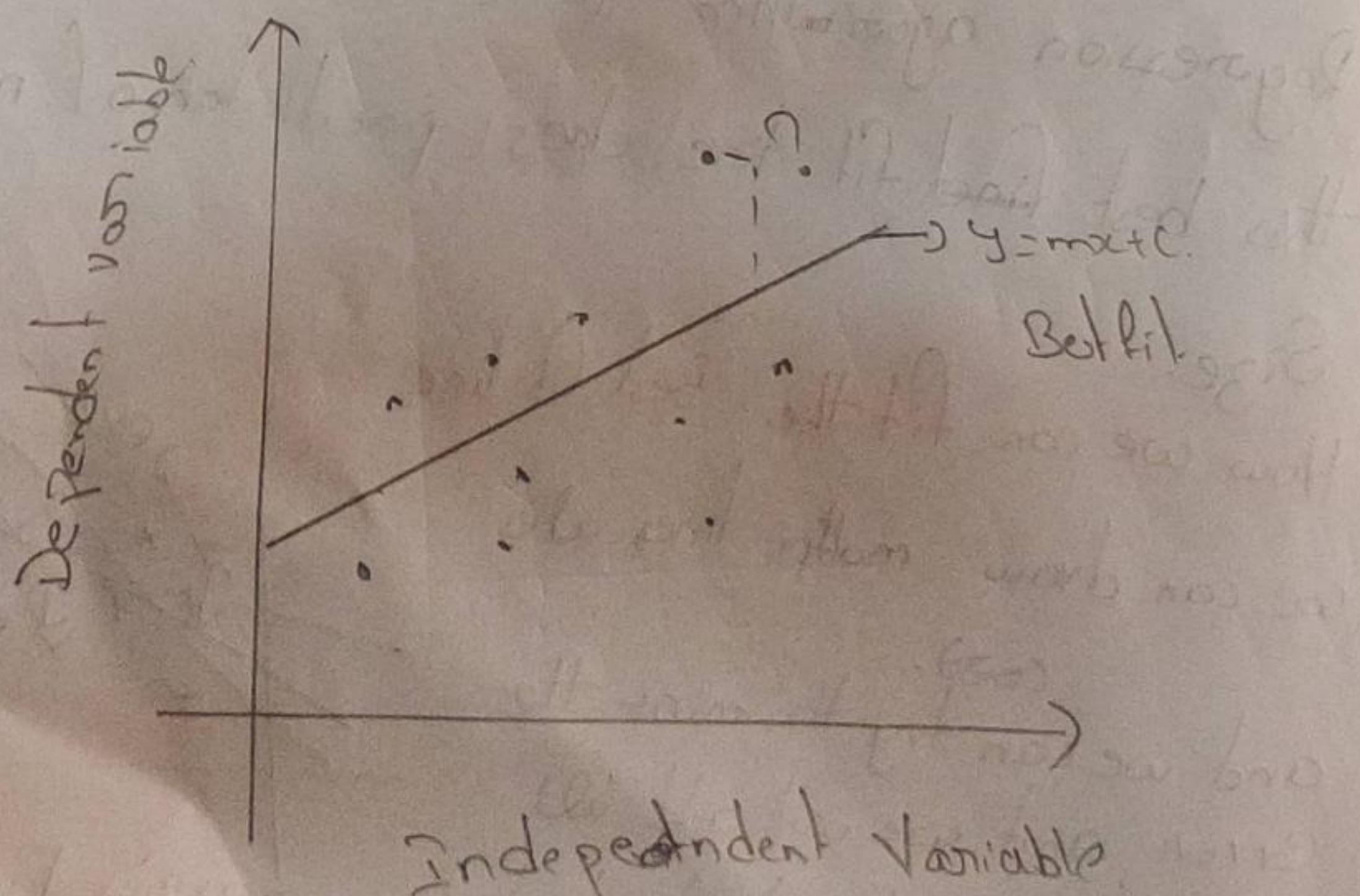




To get the Best fit line. we use the Cost function.

### Cost function in machine learning

A machine learning model should have a very high level of accuracy in order to perform well with real-world application. But how to calculate the accuracy of model. ie, how good or poor model will perform in real world? In such case, the cost function comes in existence.





Cost function also plays a crucial role in understanding that how well your model estimates the relationship between the input and output parameters

$$\text{Cost function} = \frac{1}{2n} \sum_{i=1}^n (\hat{y} - y)^2$$

Basically we need  
↓ minimize the cost function

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

$$y = mx + c$$

①

$\hat{y}$  is nothing but the point that you are finding out.  
best fit line  
 $y$  is nothing but real point

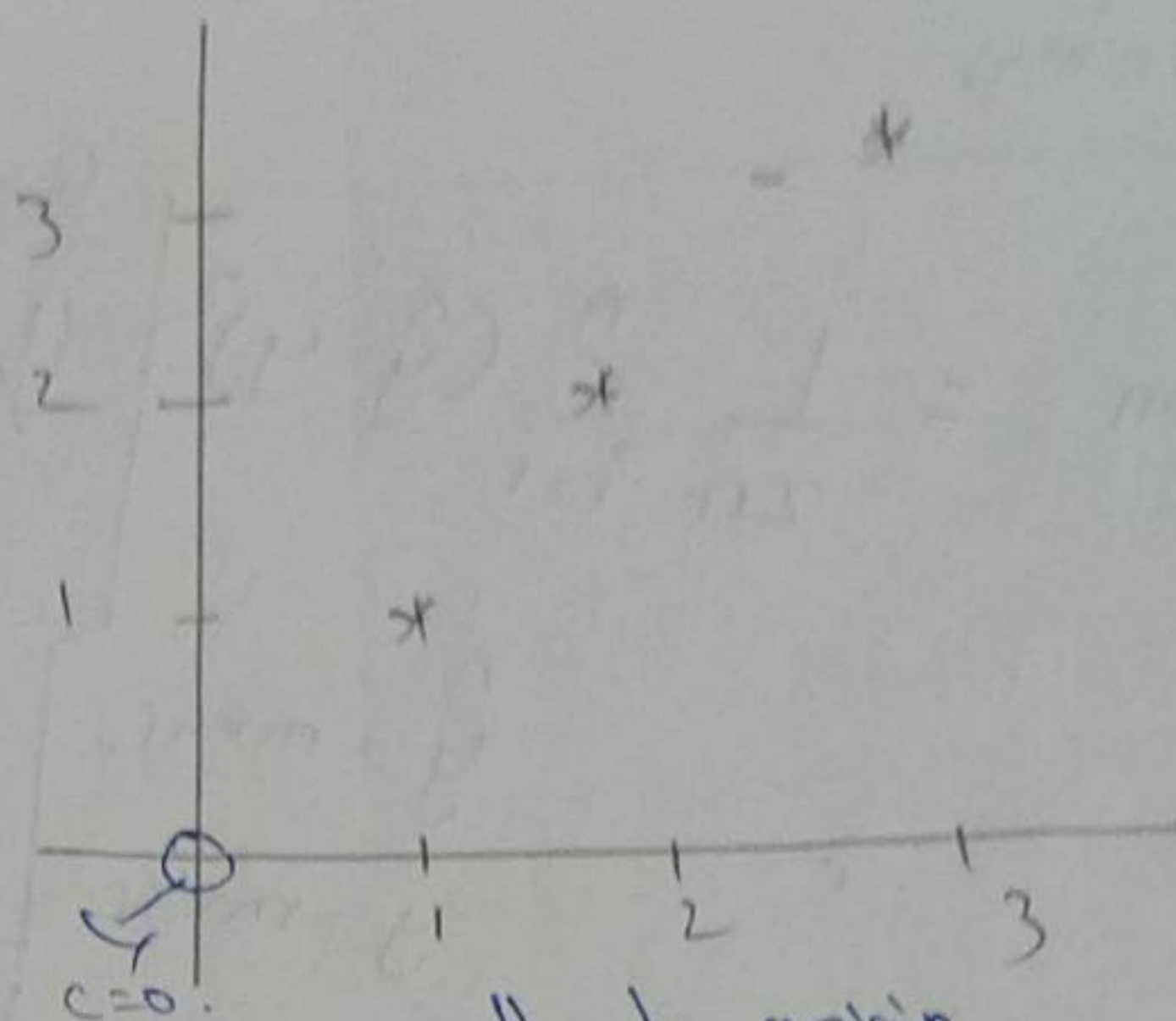
Now

By using the eq ① How we can find the best fit?  
we can have multiple best fit right and we need to compute all the sum of  $y$  that is not a convenient way it will take so much of time to do.



The Best way was

$$\boxed{y=x} \rightarrow 0$$



If  $c$  is passes through origin.

Now we need to find the best fit line.

$$\hat{y} = mx + c \rightarrow \text{if } \boxed{c=0}$$
$$\hat{y} = mx \quad \text{--- (2)}$$

$$\boxed{m=1}$$

if  $\boxed{xc=1}$

$$\hat{y} = 1 \times 1.$$

$$\hat{y} = 1$$

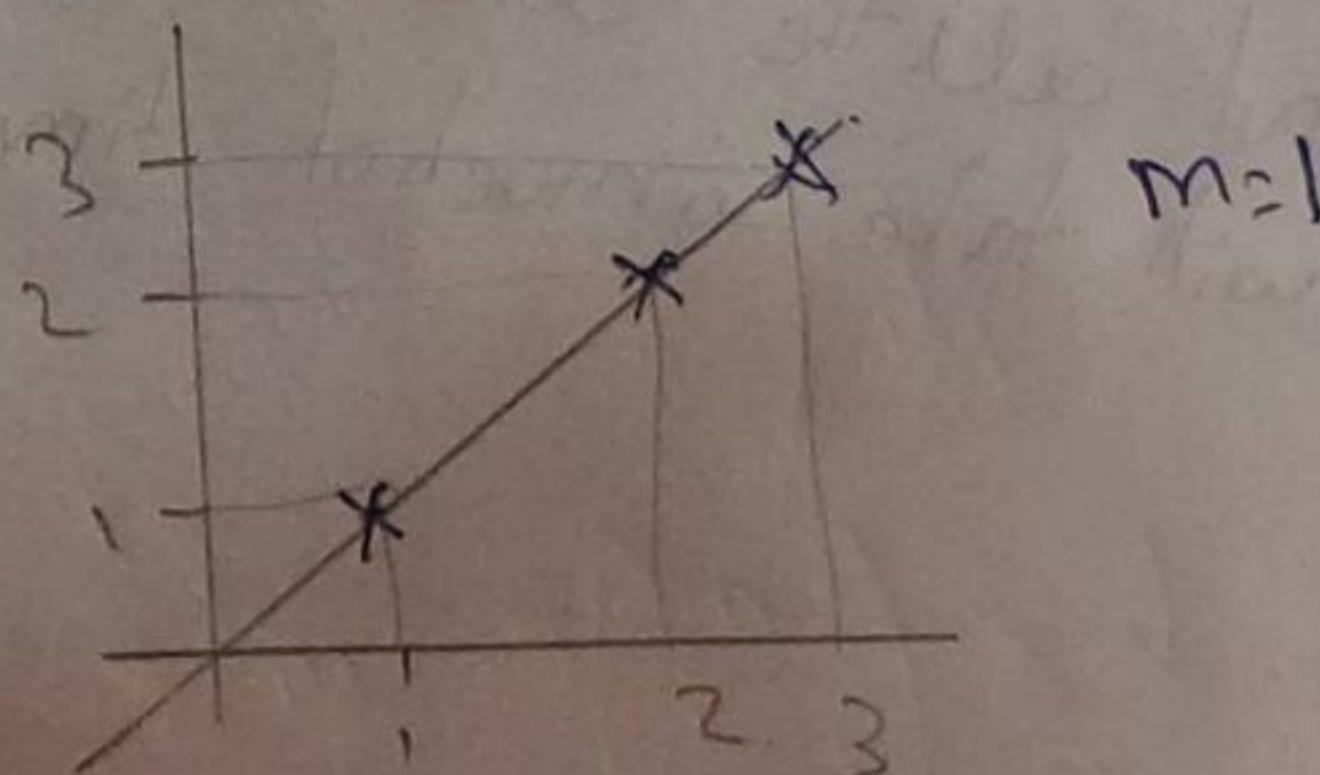
$$x=2$$

$$\hat{y} = 1 \times 2$$
$$\hat{y} = 2$$

$$x=3$$

$$\hat{y} = 1 \times 3$$
$$\hat{y} = 3$$

$\therefore$  the ~~the~~ Best fit line looks like.





we know the Cost function is

$$\text{Cost Function} = \frac{1}{2n} \sum_{i=1}^n (\hat{y} - y)^2$$

$\uparrow$  Best fit line value       $\uparrow$  real value

$$= \frac{1}{2(3)} ((1-1)^2 + (2-2)^2 + (3-3)^2)$$

$$\Rightarrow \frac{0}{6}$$

Cost function  $\Rightarrow 0$

if  $m = 0.5$

$$\hat{y} = mx + c \quad \rightarrow \boxed{c=0}$$

$$\hat{y} = mx$$

if  $\boxed{x=1}$

$$\hat{y} = 0.5 \times 1$$

$$\hat{y} = 0.5$$

$\boxed{x=2}$

$$\hat{y} = 0.5 \times 2$$

$$\hat{y} = 1$$

$\boxed{x=3}$

$$\hat{y} = 0.5 \times 3$$

$$\hat{y} = 1.5$$

$\therefore$  the Best fit line is look like.

we know Cost function is

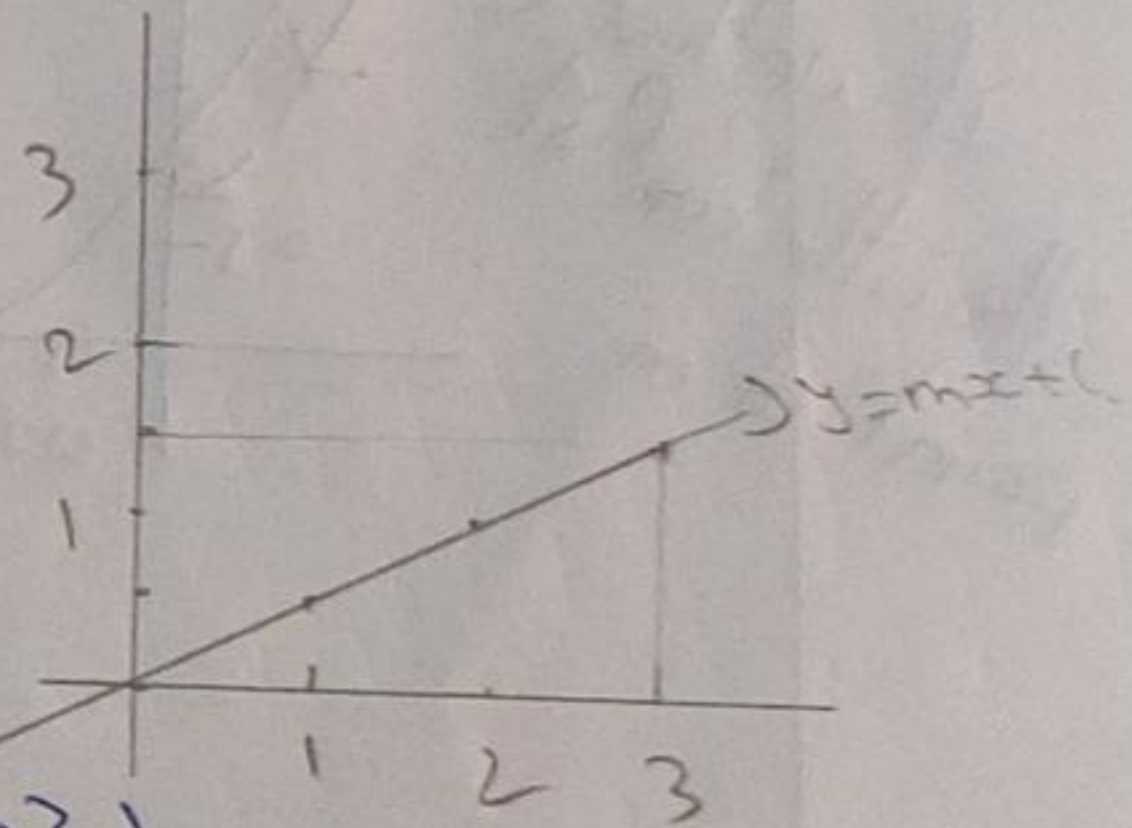
$$\text{Cost function} = \frac{1}{2n} \sum_{i=1}^n (\hat{y} - y)^2$$

$$= \frac{1}{2(3)} ((0.5-1)^2 + (1-1)^2 + (1.5-1)^2)$$

$$\Rightarrow \frac{(0.5)^2 + 0 + (0.5)^2}{6}$$

$$\Rightarrow \frac{0.5}{6}$$

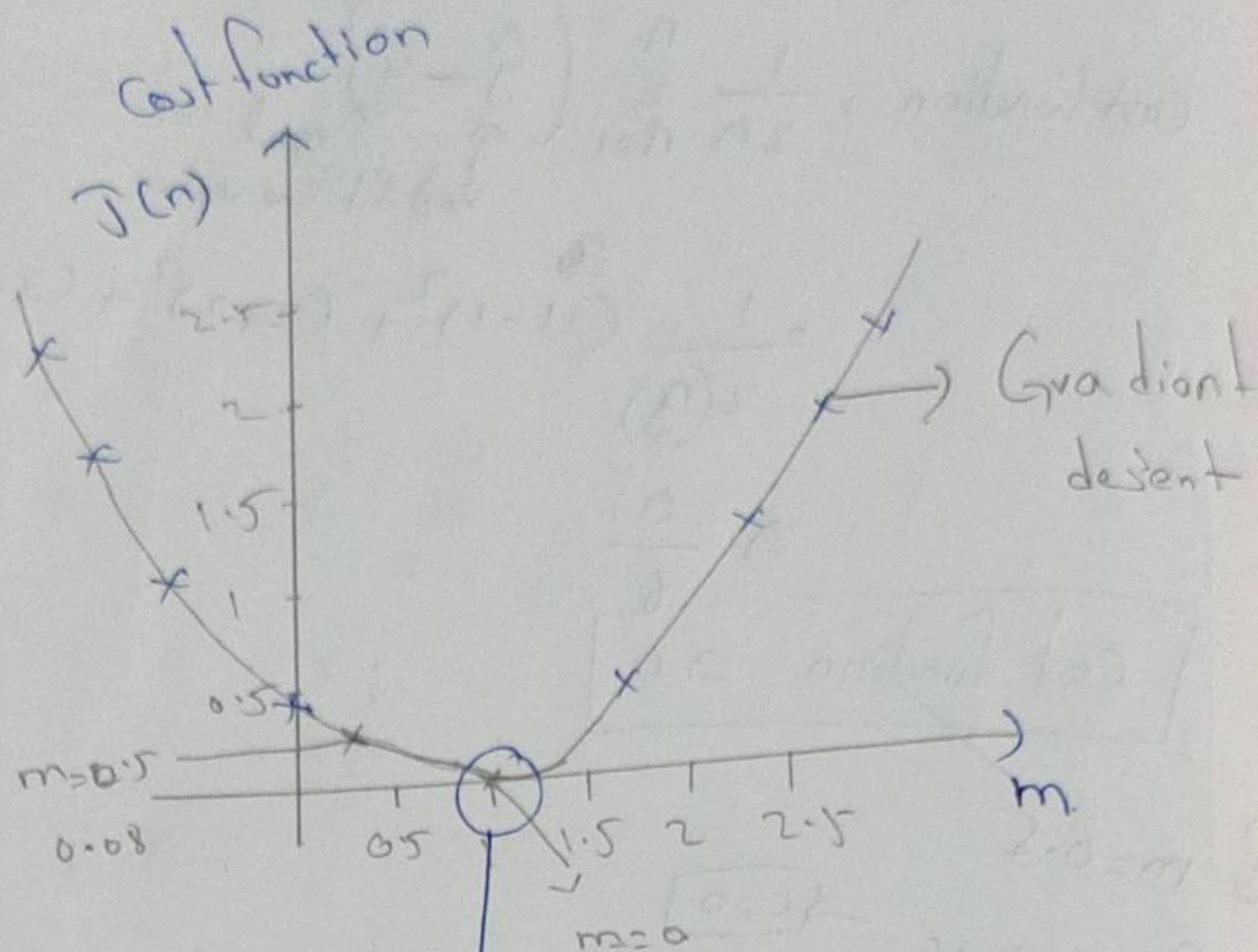
$$\Rightarrow 0.0833 \Rightarrow 0.08$$



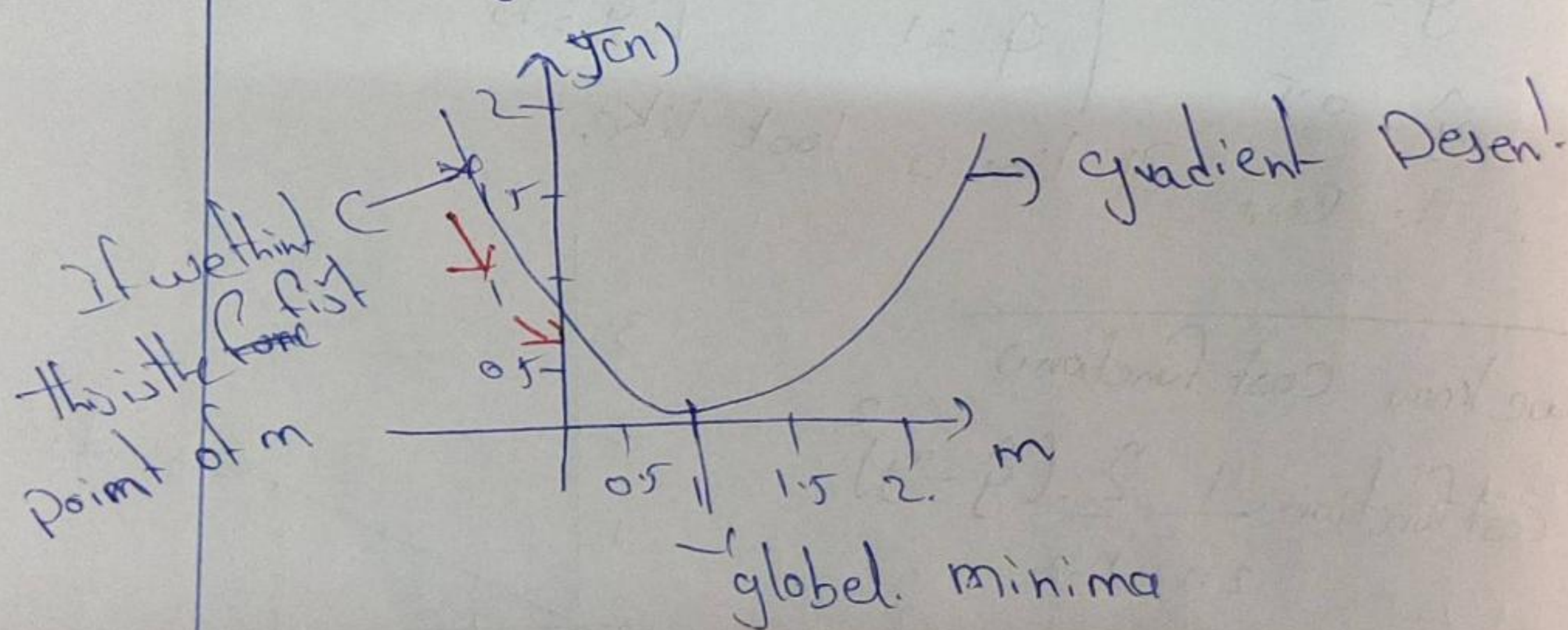
$$\begin{array}{r} 0.5 \times 0.5 \\ 2.5 \\ \hline 0.25 \\ 0.25 \\ \hline 0.50 \end{array}$$

Like that will do for all the "m" values in the graph





global minima.  
global minima.



To go down word we use the theorem called  
Convergence theorem

$$\left\{ \begin{array}{l} m \leftarrow m - \left( \frac{\partial m}{\partial J(m)} \right) \times \alpha \rightarrow \text{learning Rate} \end{array} \right.$$

↓ slope