

Simple Linear Regression (Supervised Machine Learning Algorithm)

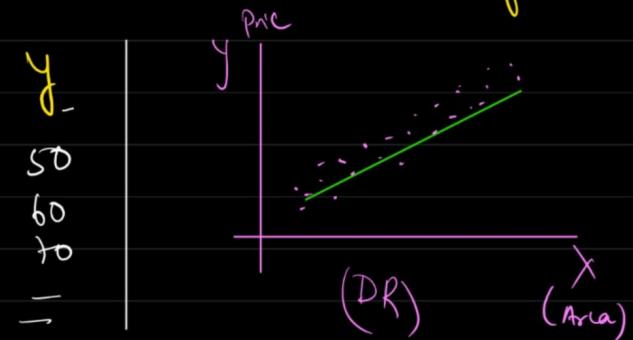
* Regression :- To establish a relationship b/w the two Variables/more than two variables.

* Linear → It establishes a linear relationship.

* SLR attempts to determine the strength and characteristics of the relationship between one dependent Variable (y) and another variable independent Variable (x)

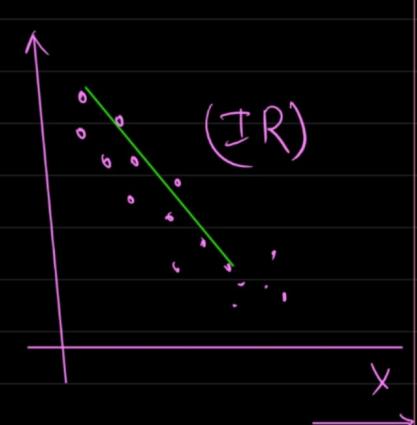
Predict Price of house based on Area of house.

Area of house	y
1100	50
1200	60
1300	70
—	—



e.g. Selling Price of Car wrt age of car.

X age of car	Y Selling Price
10	3.1
9	4.1
7	5.3
—	—



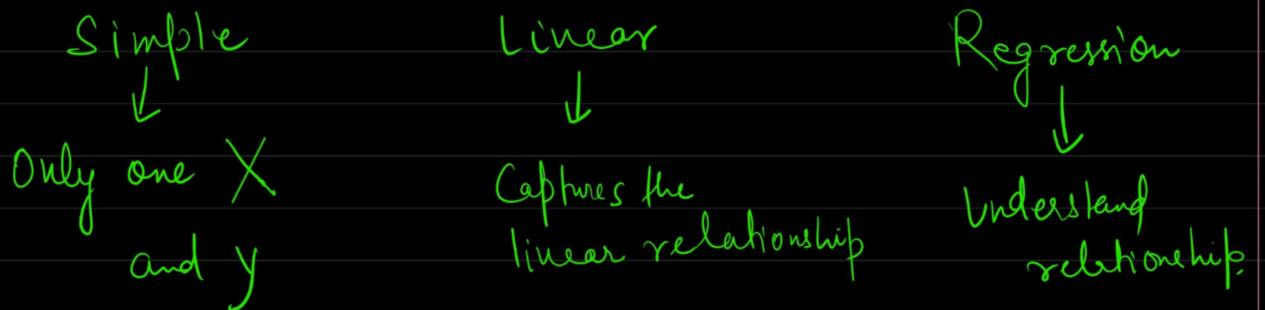
e.g. Age left vs Alcohol consumption

Age
left

(I.R.)

Alcohol consumption





* if multiple IV (multiple X's) \rightarrow multiple linear regression.

* Example

No. of Hours studied | Marks obtained

	Marks
8	72
9	85.5
-	-
-	-
-	-
	\bar{x} of Hours Studying

example

Weight	Height	ht
65	168	-
70	172	-
-	-	-
-	-	wt

* To predict price of a house based on number of rooms

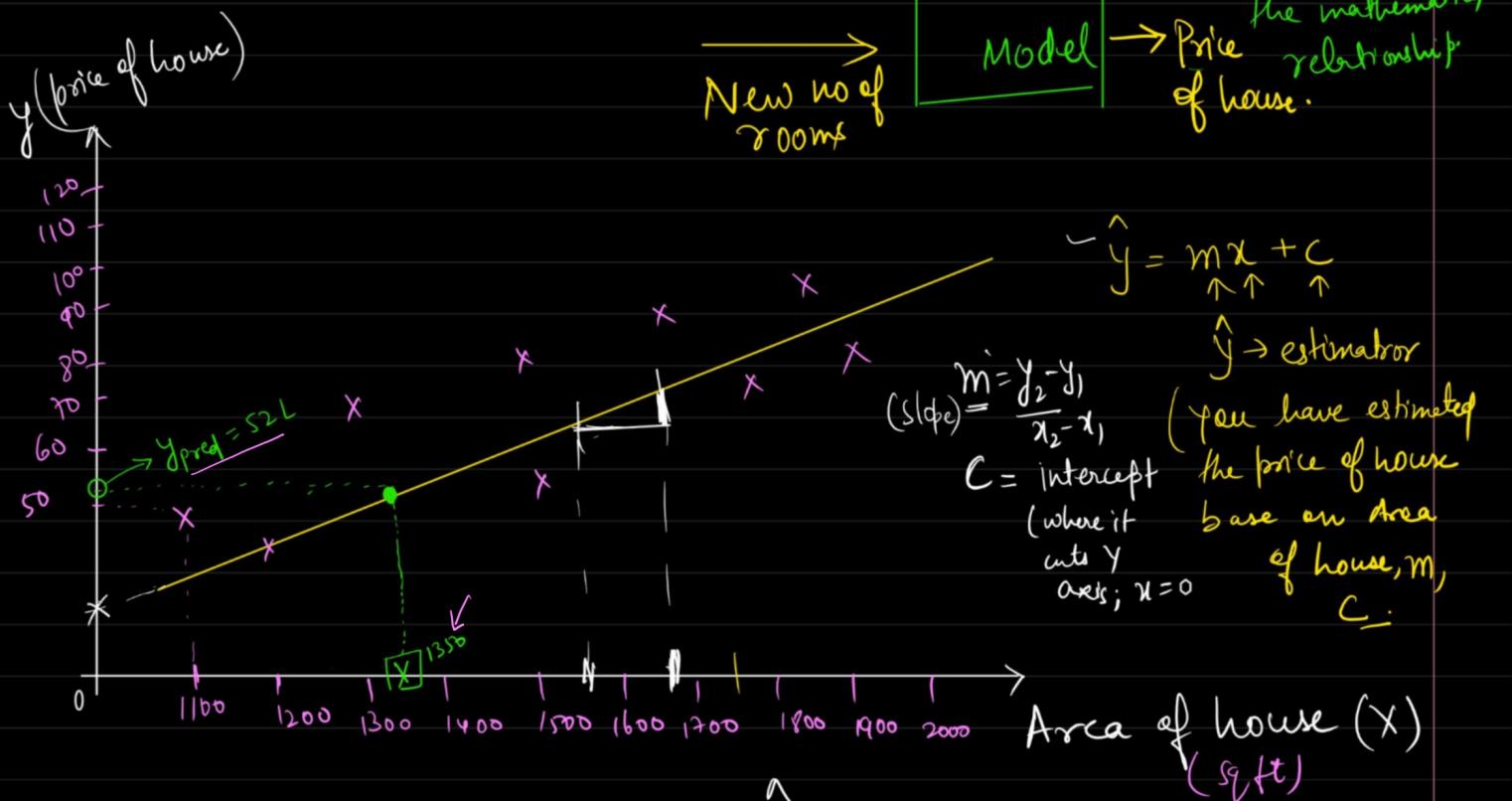
$y = \text{price of house}$

$x \rightarrow \text{Number of rooms}$.

# of rooms (x)	Price of house (y)
1	50
1.5	60
2	65
3	75
$\frac{1}{4}$	$\bar{?} \rightarrow 85$

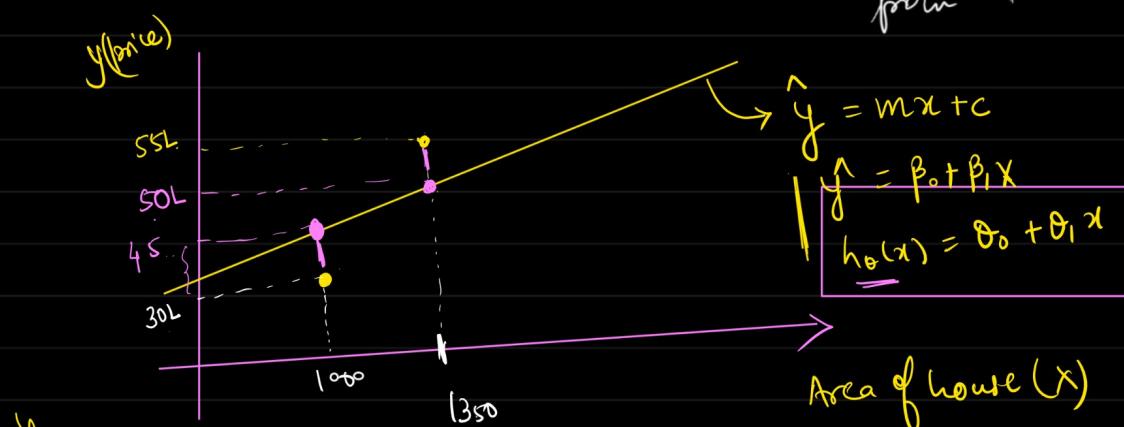
\Rightarrow ML - To understand the mathematical relationship.

This was 4 dp's that's why you understood the trend. If many dp's you



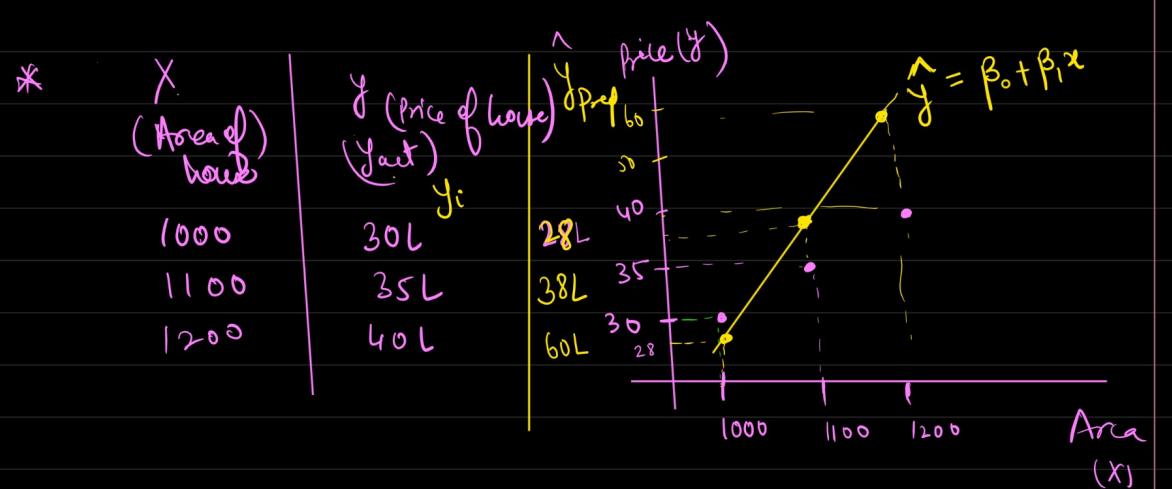
$X \rightarrow \text{is dp}$
 $(m, c), (\beta_0, \beta_1), (\theta_0, \theta_1)$
Coefficients.

Many book, you will find this format
because we consider a S.L.R line to be hypothesis that it passes through these points



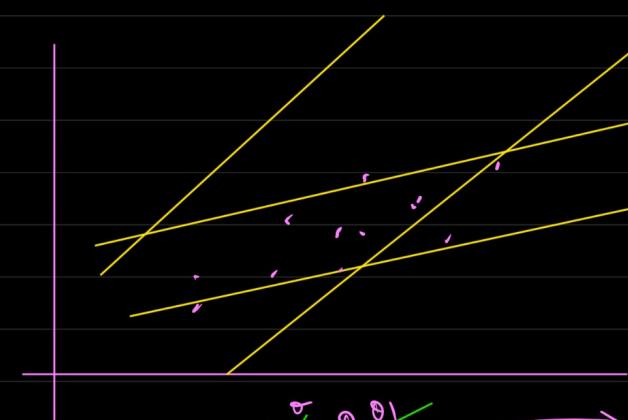
$y_{actual} \text{ for } 1350 \text{ sq} = 55L$
 $y_{pred} \text{ for } 1350 \text{ sq} = 50L$

Error = $y_{actual} - y_{pred}$



Error = $y_{actual} - y_{pred}$

$$\text{Error} = y_i - \hat{y}_i \quad \rightarrow \quad \hat{y}_i = \theta_0 + \theta_1 x_i$$

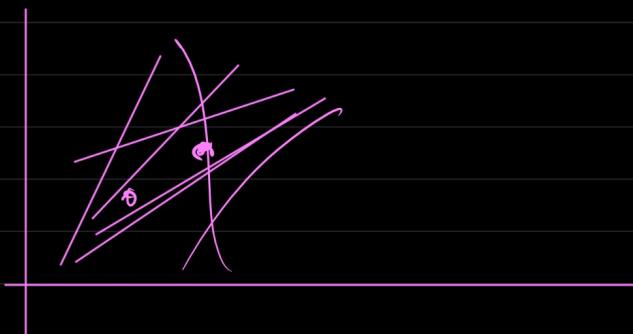


Error $\downarrow (\theta_0, \theta_1), (\beta_0, \beta_1)$
optimal.

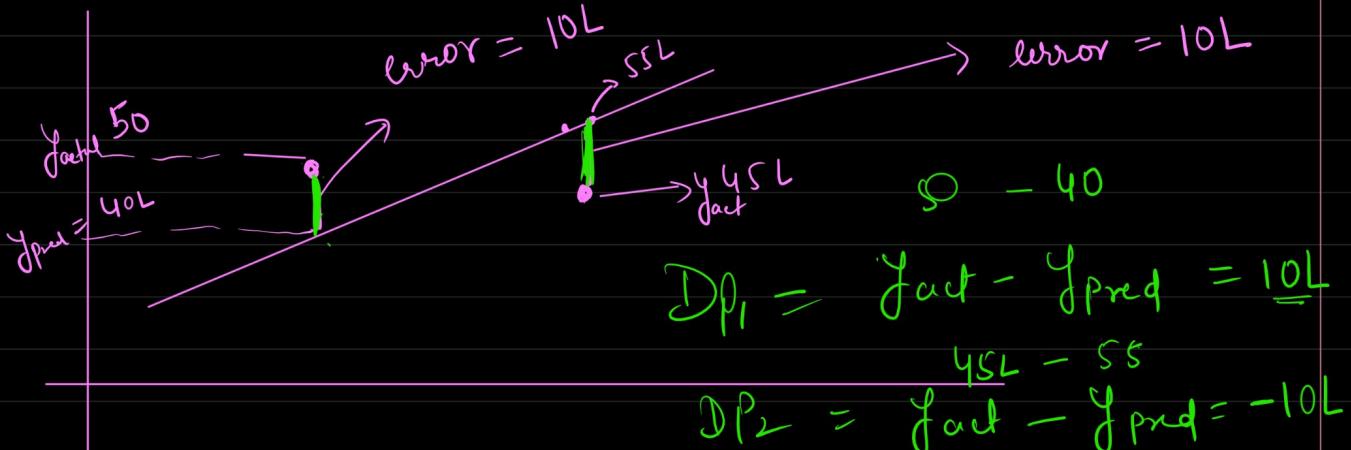
for all lines, we want

Error = $y_{\text{act}} - y_{\text{pred}}$ to be least.

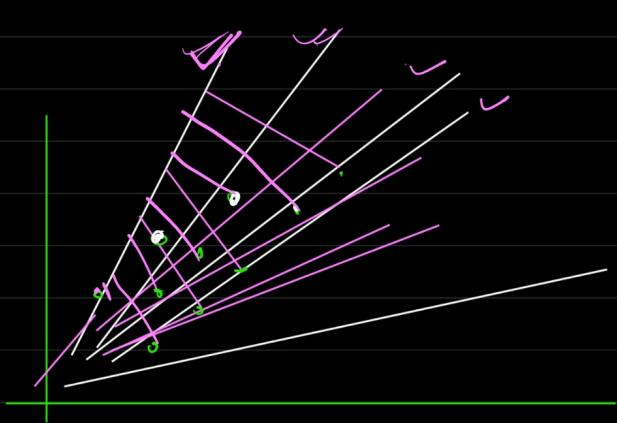
Optimization process



$$\text{Squared error} = \underline{(y_{\text{act}} - y_{\text{pred}})^2}$$



$$\text{for all the } dP's = \underline{10L} + \underline{(-10L)} = 0$$



$$10^2 + (-10)^2 \\ = 100 + 100 = 200$$

$$= \min \underline{(y_{\text{act}} - y_{\text{pred}})^2}$$

$$= \min \sum_{i=1}^n \underline{(y_i - \hat{y}_i)^2}$$

$$y_{\text{actual}} = \min \sum_{i=1}^n (y_i - h_{\theta}(x))^2$$

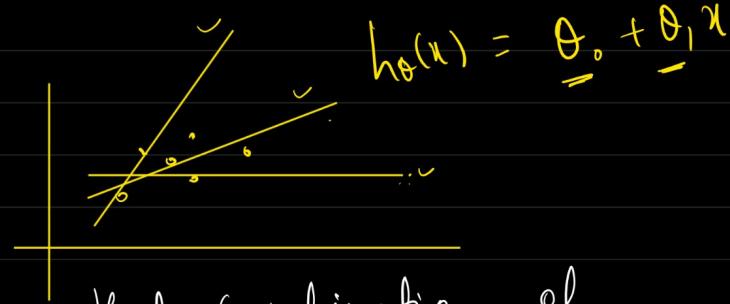
✓ $\min \sum_{i=1}^n (y_i - h_{\theta}(x))^2$

n is no. of datapoint

least square method
or
Ordinary least
square method.

$$y_i - y_{\text{actual}}$$

$$h_{\theta}(x) \rightarrow y_{\text{predicted}}$$



We want to have that combination of m, c (θ_0, θ_1), (β_0, β_1) where the overall error is least and the line will be best fit line.

$$\min \sum_{i=1}^n (y_i - h_{\theta}(x))^2$$

$$\Rightarrow \min \sum_{i=1}^n (y_i - (\theta_0 + \theta_1 x))^2$$

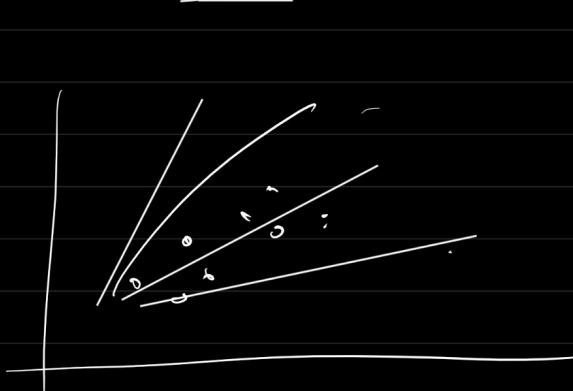
IQ
Technique to find best fit line in SLR
→ OLS.

$$\left\{ \begin{array}{l} \theta_0^1 \quad \theta_1^1 \\ \theta_0^2 \quad \theta_1^2 \\ \theta_0^3 \quad \theta_1^3 \\ \vdots \quad \vdots \\ \theta_0^n \quad \theta_1^n \end{array} \right\} \rightarrow \text{optimal } (\theta_0, \theta_1)$$

S.S.E -

$$= \min \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

Mean square error = $\min \frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2$ $\left(\frac{1+2+3}{3} \right)$ $\rightarrow \underline{\text{mean}}$



* For best fit line

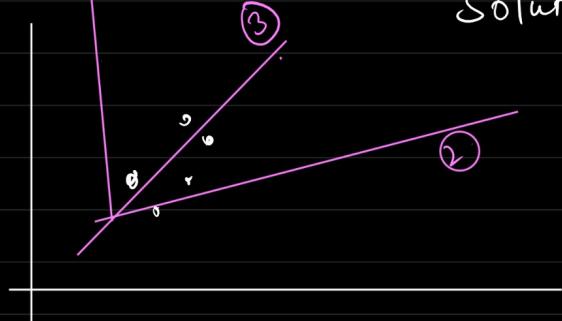
MSE should be least

Penalty

Cost function.

①

MSE is used as CF to get
solution $\rightarrow \beta_0, \beta_1 (\theta_0, \theta_1)$

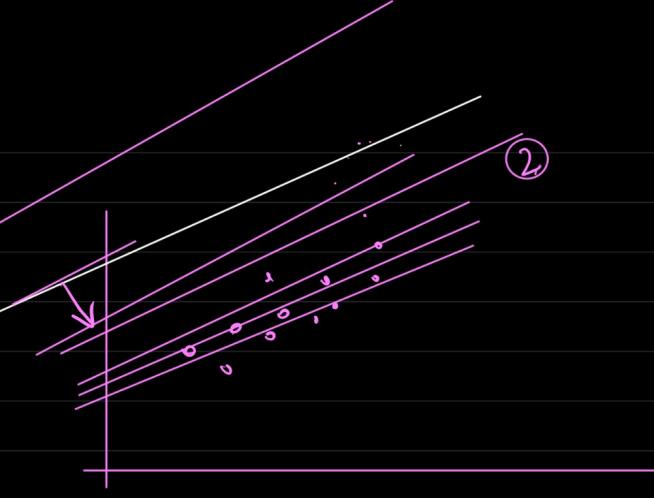


cost function

$$\Downarrow J(\theta_0, \theta_1) = \frac{1}{n} \min \sum_{i=1}^n (y_i - h_\theta(x))^2$$

$$h_\theta(x) = (\theta_0 + \theta_1 x)^2$$





To get optimal line

- ① choose right combination of
- ② Find Mean Square Error line
- ③ best fit line



$$\begin{aligned}
 L_1(m_1, c_1) &= MSE_1 \\
 L_2(m_2, c_2) &= MSE_2 \\
 L_3(m_3, c_3) &= MSE_3 \\
 &\vdots \\
 L_D(m_D, c_D) &= MSE_D
 \end{aligned}$$

MSE_{least}
 best fit line
 $(\text{optimal } \beta_0, \beta_1)$

Optimization process \Rightarrow gradient descent