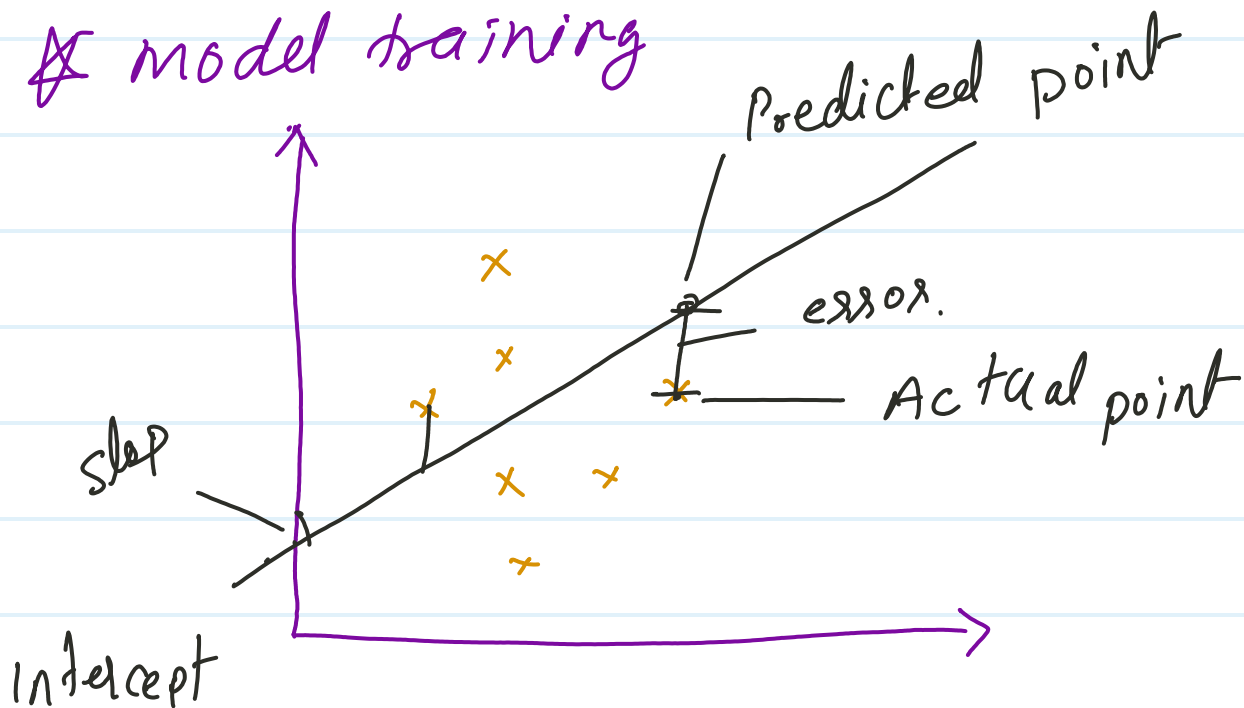


# ★ Linear Regression

## ★ model training



$$y = mx + c$$

$y$  = Actual data  
 $\hat{y}$  = Predicted data

$y$  = Dependent variable

$x$  = Independent variable

$m$  = slop

$c$  = Intercept

$y - \hat{y}$  = Residual error

base eqn line

$$y = mx + c$$

or

$$y = h_{\theta}(x)$$

or

$$h_{\theta}(x) = \theta_0 + \theta_1 x_1$$

or

$$h_{\theta}(x) = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_3 - \dots - \theta_n x_n$$

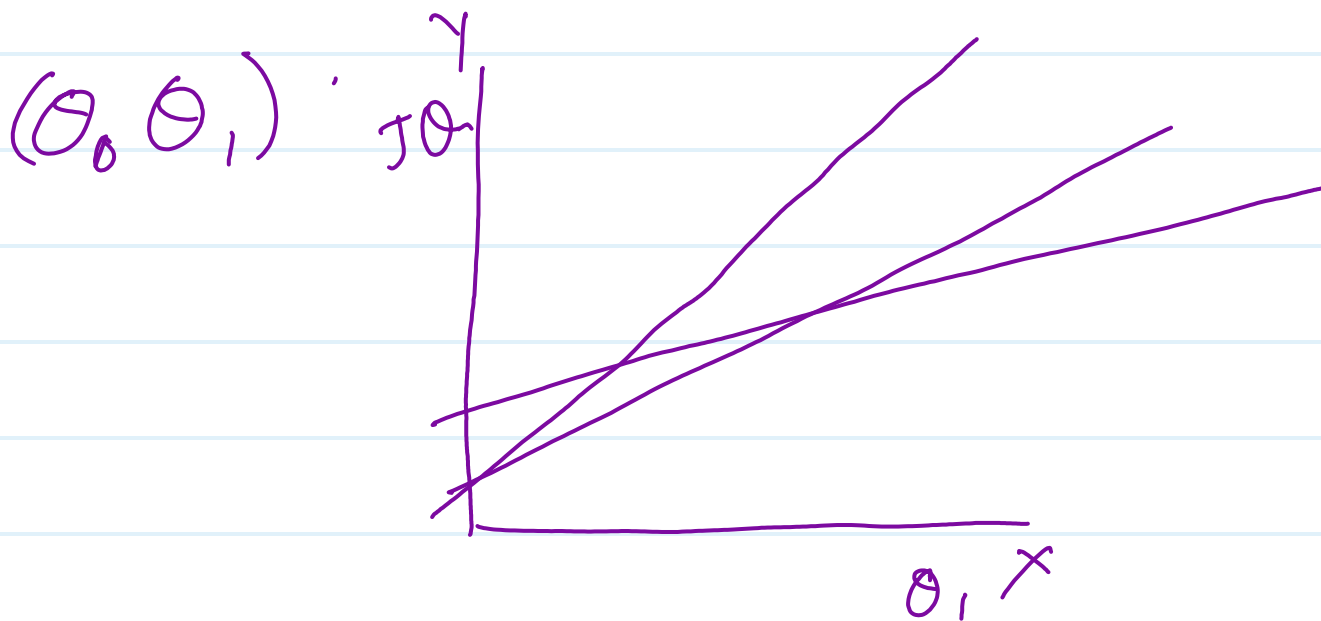
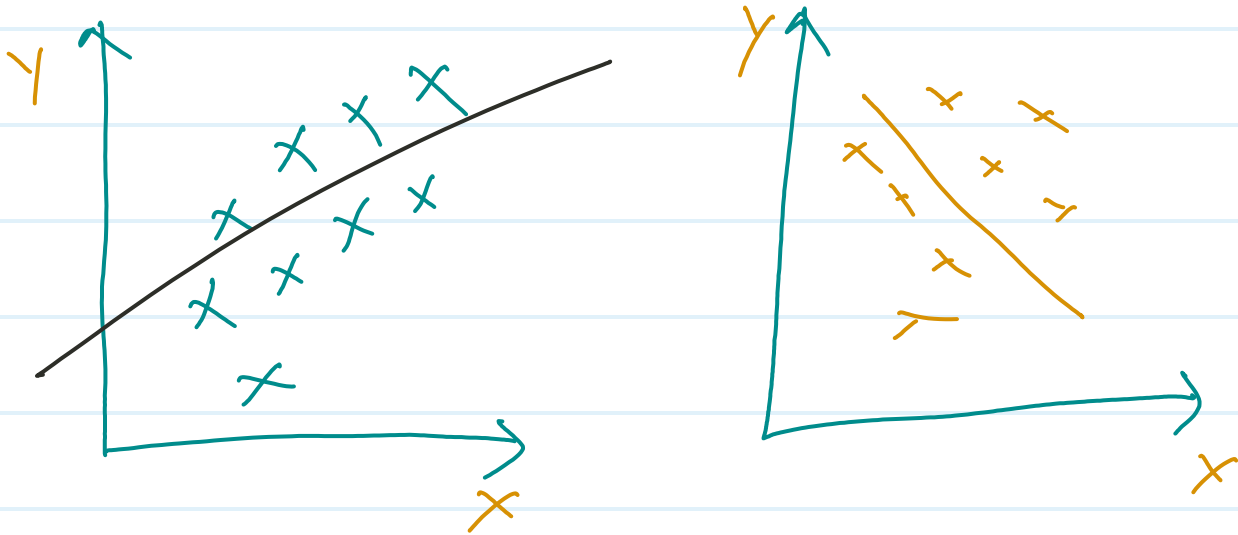
# cost function

$$J = y - h_{\theta}(x)$$

$$J = J(\theta_0, \theta_1)$$

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

This is cost function to min. error by changing value of  $\theta_0, \theta_1$



Repeat convection Theorem

$$\theta_j = \theta_j - \alpha \frac{d}{d\theta_j} [J(\theta_j)]$$

$$\alpha = 0.05, 0.025, 0.1$$

$$\theta_0 = \theta_0 - \alpha \frac{1}{n} \sum_{i=1}^n (h_\theta(x^i) - y^i)$$

$$\theta_1 = \theta_1 - \alpha \frac{1}{n} \sum_{i=1}^n (h_\theta(x^i) - y^i) x^i$$

\* Model evaluation or performance metrics

- (i) MSE (mean squared error)
- (ii) RMSE (Root mean square error)
- (iii) MAE (mean Absolute error)
- (iv)  $R^2$
- (v) Adj.  $R^2$

(i) MSE

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

↓ lower

## ② RMSE

$$RMSE = \frac{1}{n} \sqrt{\sum_{i=1}^n (y - \hat{y})^2}$$

Y  
2.  
3.5  
4  
6.  
7.1

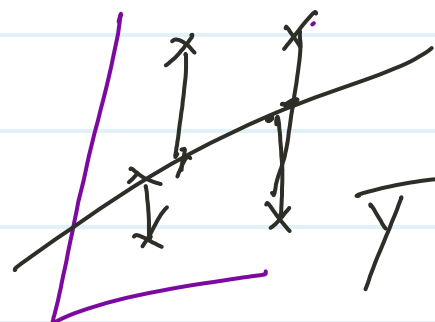
## ③ MAE

$$MAE = \frac{1}{n} \sum_{i=1}^n |y - \hat{y}|$$

lower value better.

## \* Accuracy matrix

### ① $R^2$



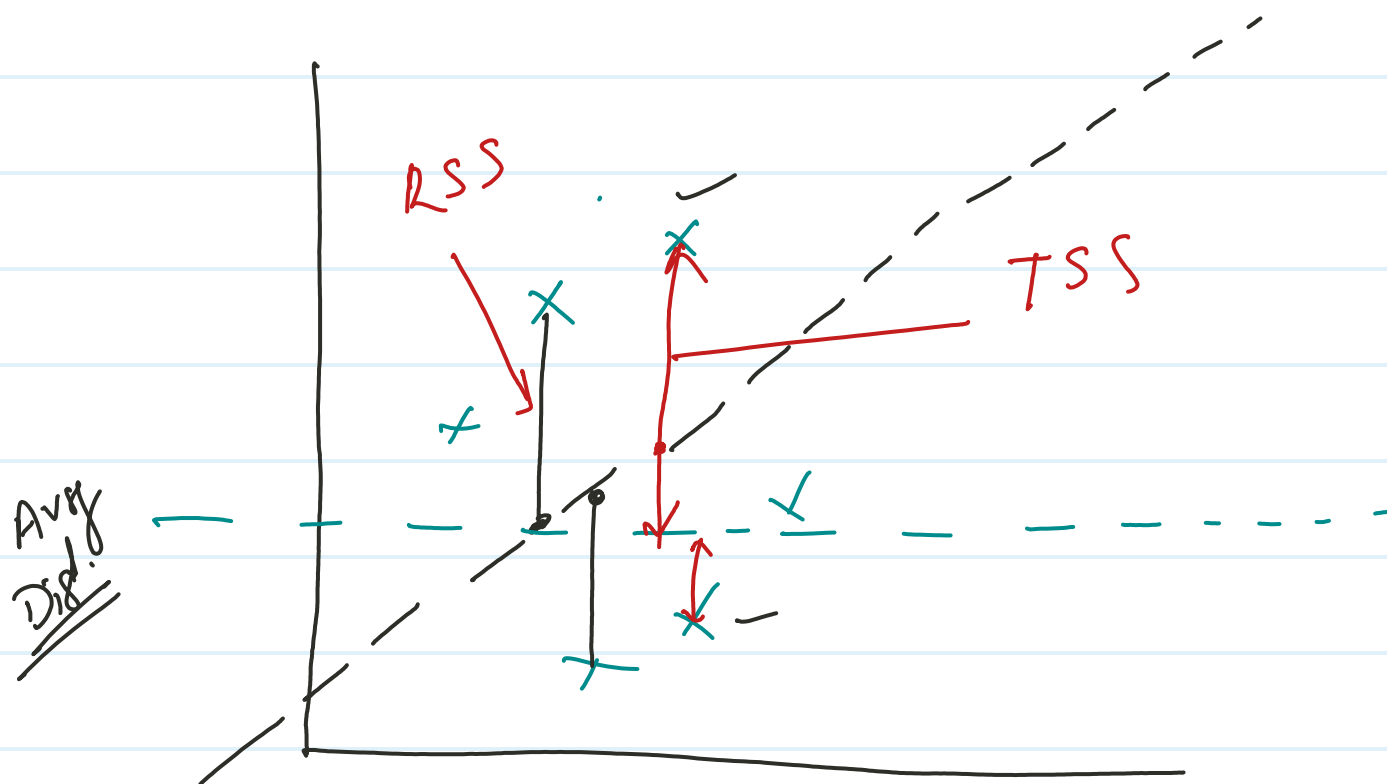
$$R^2 = 1 - \frac{RSS}{TSS} \quad 0 - 1$$

$R^2$  = coeff. of determination

RSS = sum of square of residual

RSS = Distance b/w  $y$  and  $\hat{y}$

TSS = Distance b/w  $y$  and  $\bar{y}$



$$RSS = \sum (y - \hat{y})^2$$

$$TSS = \sum (y - \bar{y})^2$$

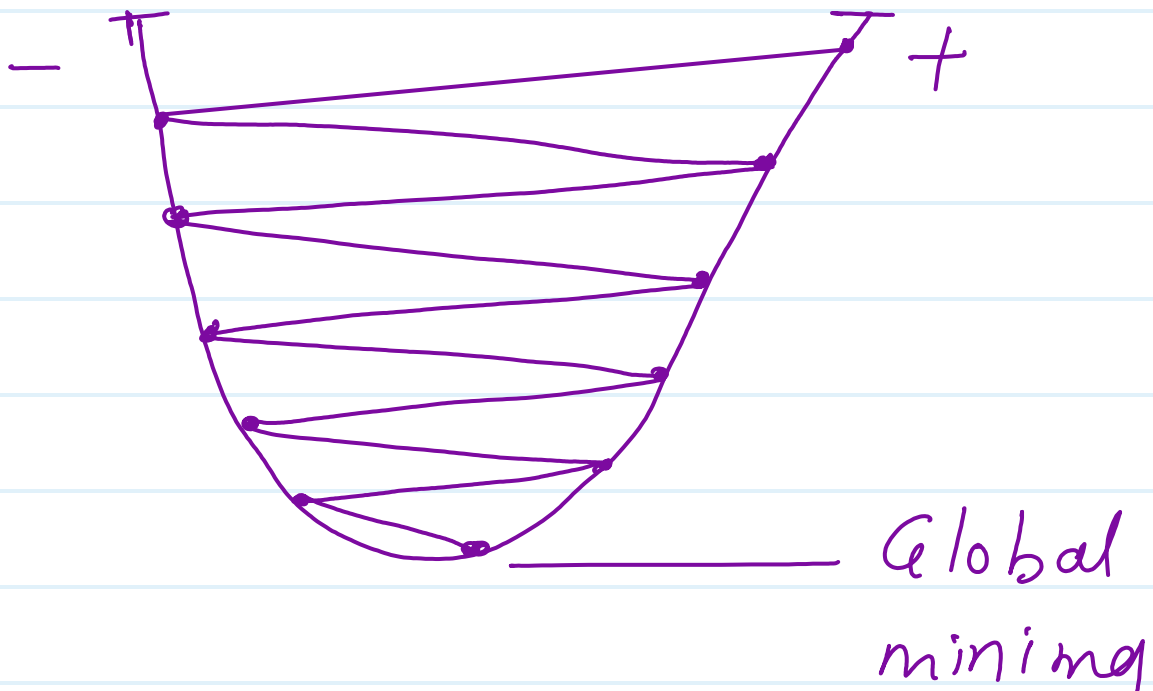
TSS Avg distance

② Adj.  $R^2$

$$Adj. R^2 = 1 - \frac{(1 - R^2)(N - 1)}{N - p - 1}$$

$n$  = no. of datapoint in our dataset

$p$  = no. of independent variable  
( $x_1, x_2, x_3, \dots$ )



\* To find multi co-linearity

$X_1$	$X_2$	$X_3$	$X_4$	$Y$
←————→				↔

\* VIF (variance inflation Factors)

$$VIF = \frac{1}{1 - R^2}$$

VIF = start 1 and it has no limit

IF 1 or less than 5 so no. multicollinearity

If  $> 5$  so there will be co-linearity b/w inde. feature.



