

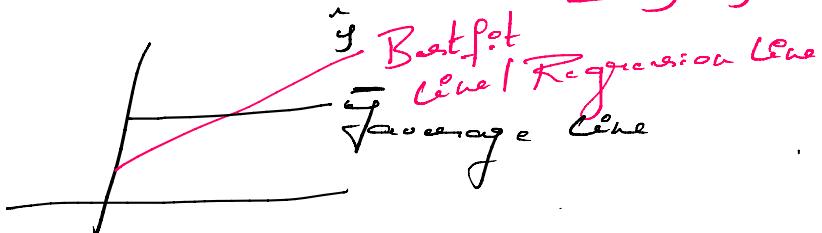
Metrics to measure the performance of the model.

① R^2 -Score / R^2 -Score / R-squared.

\checkmark It is used to measure strength of the model

* R^2 Score Compares Regression line with average line.

$$R^2 = 1 - \frac{RSS}{TSS} = 1 - \frac{\sum (y_i - \hat{y}_i)^2}{\sum (y_i - \bar{y})^2}$$



Range of $R^2 = [-1, 1]$

If R^2 is near to "1", we say it's good model.

If R^2 is near to "0", we say bad model.

Python

`R^2_score(y_true, y_pred)`.

② Adjusted R^2 -Score

Disadvantage of R^2

R^2 score increases as the number of independent variables increases which has very less relationship with target variable.

To overcome above issue we use adjusted R^2

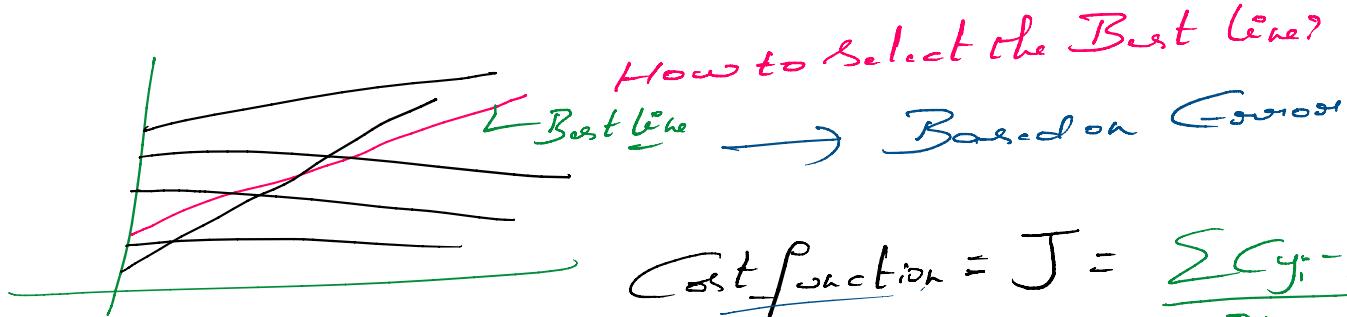
Adjusted R^2 will measure the performance of the model by ignoring columns which has very less relationship with target.

$$\text{adj } R^2 = 1 - \frac{(1-R^2)(n-1)}{n-p-1}$$

$R^2 \rightarrow R^2$ -Score
 $n \rightarrow \text{No of observation}$
 $P \rightarrow \text{No of independent variables.}$

x note

Adjusted R^2 & R^2 -score, when we say it's a good model.

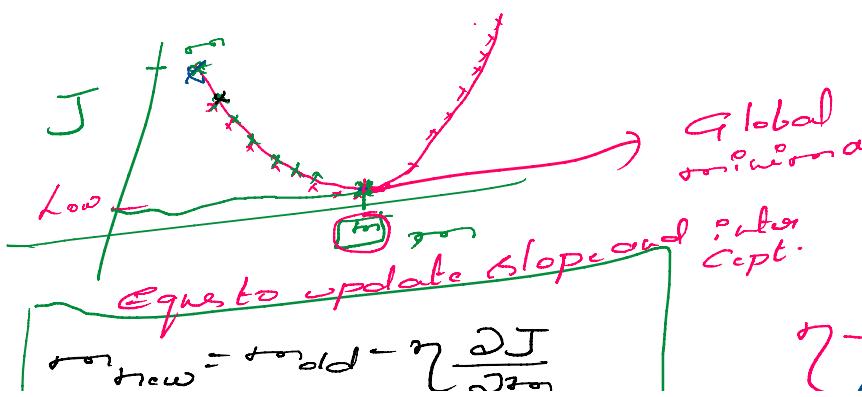
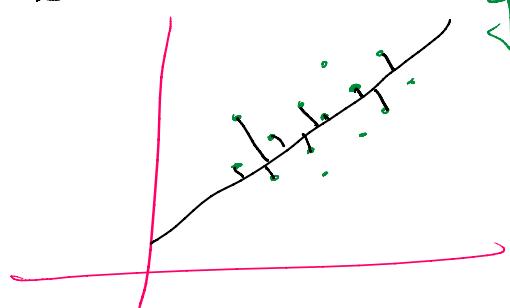


$$\text{Cost function} = J = \frac{\sum (y_i - \hat{y}_i)^2}{n}$$

* what if error very high?

Gradient Descent

Gradient Descent is one of the optimization techniques which minimizes error / loss by choosing optimal values for slope and intercept.



* Gradient Descent starts with random slope and works iteratively to reach global minima.

Global minima is a point where derivative is zero

$\eta \rightarrow \text{Learning rate}$
the model has

EQUATION -

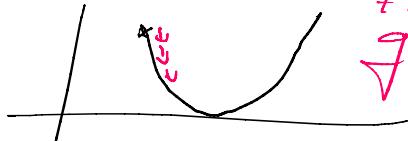
$$C_{\text{new}} = C_{\text{old}} - \eta \frac{\partial J}{\partial C}$$

$$C_{\text{new}} = C_{\text{old}} - \eta \frac{\partial J}{\partial \theta}$$

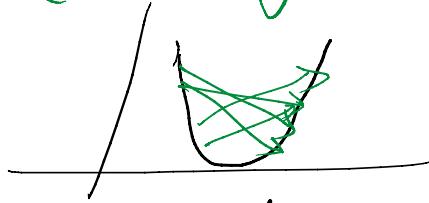
$\eta \rightarrow$ Learning rate

How many steps model has taken to reach global minima.

η is small \rightarrow It will take more time to reach global minima.



η is large \rightarrow It will overshoot that means it will never reach global minima.



How to choose η ? Larger steps and initially take larger steps and take smaller steps when near to global minima.

