

## MACHINE LEARNING DAY 3

15/OCT/2022

- Ridge Regression
- Lasso Regression
- Elastic Regression.

### ① Ridge Regression (L2 Regularization)

The main aim of Ridge Regression is to reduce overfitting.

Note:-

Overfitting means the model will perform extremely well for Training data but it will not perform well for Test data.

$$\text{Cost Function} = \frac{1}{m} \sum_{i=1}^m \left( h_{\theta}(x^{(i)}) - y^{(i)} \right)^2 + \lambda (\text{Slope})^2$$

where  $\lambda$  = Hyperparameter

If  $h_{\theta}(x) = \theta_0 + \theta_1 x$  is the equation the slope will be  $(\theta_1)^2$

For multiple variables.

$$\text{Cost Function} = \frac{1}{m} \sum_{i=1}^m \left( h_{\theta}(x^{(i)}) - y^{(i)} \right)^2 + \lambda \sum_{i=1}^m (\text{Slope})^2$$

Suppose the equation is  $h_{\theta}(x) = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_3$



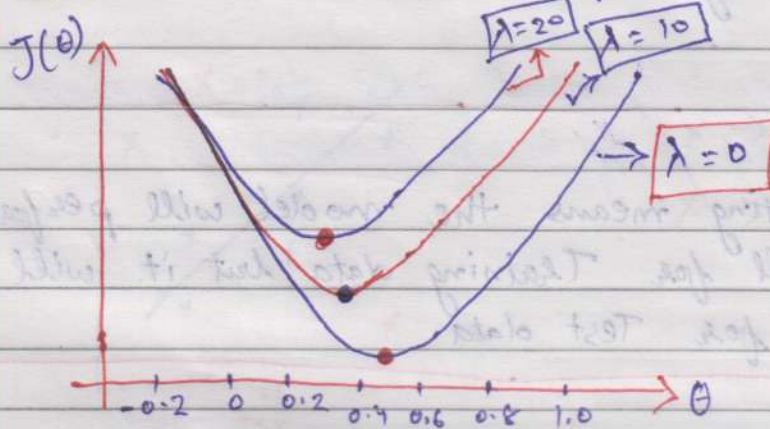
Then Slope will become  
 $(\theta_1 + \theta_2 + \theta_3)^2$

Now talk about  $\lambda \sum_{i=1}^m (\text{Slope})^2$

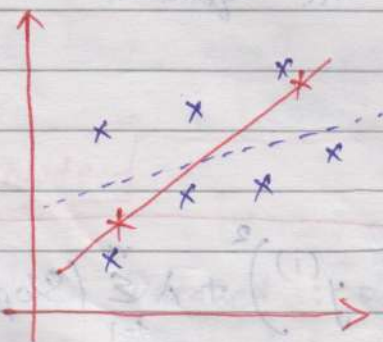
If  $\lambda = 0$ , Cost function will be of linear regression itself.

### Interview Question:

Q = What is the relationship between slope and  $\lambda$ ?



Ans.:- We can say that  $\lambda$  is inversely proportional to slope because we can see from the above graph as the  $\lambda$  value increase there is a decrease in the slope.



$$\text{Cost function} = 0 + \lambda (\text{Slope})^2 = +ve$$

$\therefore$  Slope decrease as  $\lambda$  is +ve

**Note:-** In Ridge Regression, value of  $\theta$  can never be Zero.



## ② Lasso Regression :- (L1 Norm) (L1 Regularization)

This is basically used to reduce the features.  
{ Feature Selection }

Formula:-

$$\boxed{\text{Cost Function}} : \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x)^{(i)} - y^{(i)})^2 + \lambda \sum_{i=1}^m |\text{Slope}|$$

Lasso Regression uses L1 regularization technique. It is used when we have more features because it automatically performs feature selection.

Suppose:-  $h_{\theta}(x) = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_3$   
 $= \theta_0 + 0.54 x_1 + 0.23 x_2 + 0.10 x_3$

Now as value of  $\lambda$  increases the 3<sup>rd</sup> Feature  $0.10 x_3$  decreases and at some value of  $\lambda$ , it will be zero and get removed there fore the model will focus only on important features.

## ③ Elastic Net Regression (L1 and L2 Norm)

It is the combination of both Ridge and Lasso Regression.

$$\text{Cost Func} :- \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x)^{(i)} - y^{(i)})^2 + \lambda_1 \sum_{i=1}^m (\text{Slope})^2 + \lambda_2 \sum_{i=1}^m |\text{Slope}|$$

$\Downarrow$  Ridge Regression       $\Downarrow$  Lasso Regression