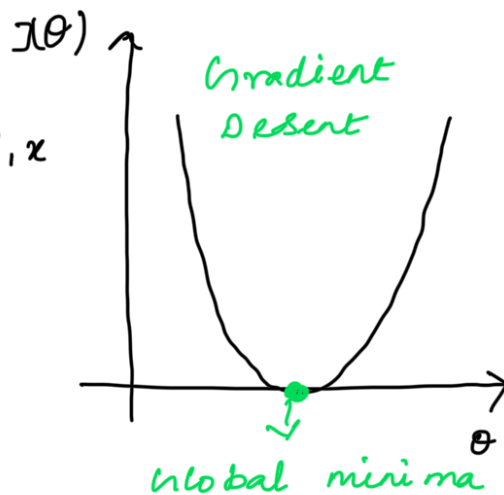
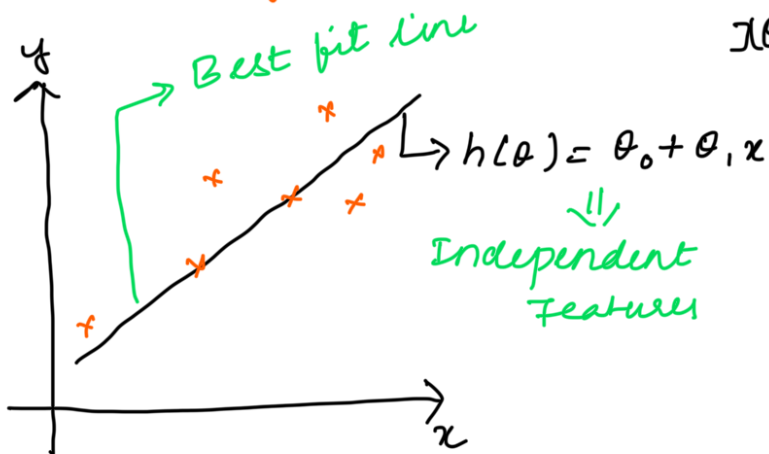


# Ridge Regression, Lasso Regression,

## Elasticnet Regression

### Linear Regression

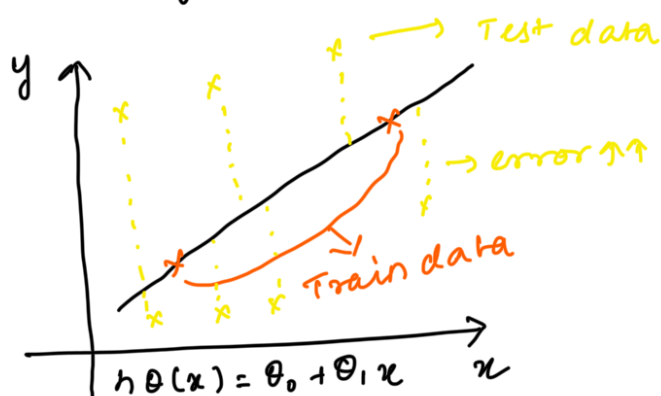


$$\text{cost fn} = \frac{1}{2m} \sum_{i=1}^m (h_{\theta}(x^i) - y^i)^2$$

$\Downarrow$   
mean squared error

### ① Ridge regression

Let say I have the below data points



### Over fitting

Train data  $\rightarrow$   
accuracy  $\uparrow \rightarrow$  low Bias

Test data  $\rightarrow$   
accuracy  $\downarrow \rightarrow$  high Variance.

here my model is overfitting.  
here the model performed well  
with the train data, however  
with the test data, error  
increases, this is the problem of

### note:

we should not  
get 100% accuracy on  
training data.

## Overfitting.

the training data, as this will lead to overfitting.

Ridge regression is also called as  $L_2$  regularization which helps to reduce overfitting.

Suppose, my linear regression has some overfitting. in order to reduce the overfitting in linear regression we use Ridge regression.

We can consider the Ridge regression as a new algorithm to hyperparameter tune the linear regression.

cost fn of Ridge regression

$$\text{cost fn} = \frac{1}{2m} \sum_{i=1}^m (h_{\theta}(x^i) - y^i)^2$$

Hyper parameter.

$$+ \lambda \sum_{i=1}^n (\text{slope})^2$$

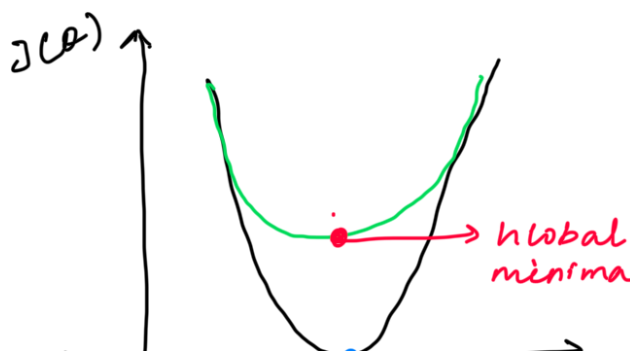
Let  $\lambda = 1$ ,

$$= 0 + 1 \left[ (\theta_1)^2 \right]$$

when adding these two parameters, the situation is going to be in such a way that, not going to get the best fit line that

$\therefore$  my cost function will always  $> 0$  passes exactly on my training data.

This is how ridge regression helps to avoid overfitting.



$$\lambda = 0$$
$$\lambda = 0$$

so when ever,  $\lambda$  value changes gradient descent

global minima  $\theta$

will also change, will also get new global value, and  $\theta$  value reduces.

how Ridge regression reducing overfitting?

→ It reducing the impact by reducing the coefficient of the feature that are not directly related to the output feature.

## ② Lasso Regression

It is also called as  $L_1$  regularization, we specifically use Lasso regression for feature selection.

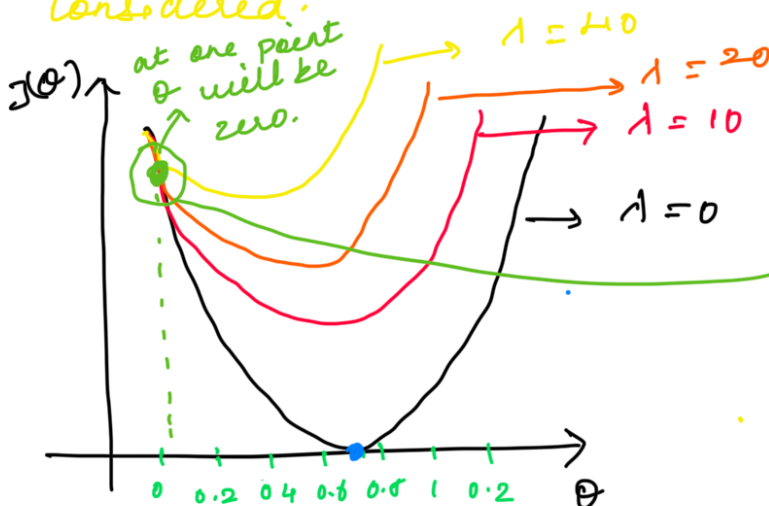
$$\text{cost fn} = \frac{1}{2m} \sum_{i=1}^m \left( h_{\theta}(x^i) - y^i \right)^2 + \lambda \sum_{i=1}^n |\text{slope}|$$

||<sub>1</sub>

used for feature selection.

What is feature selection?

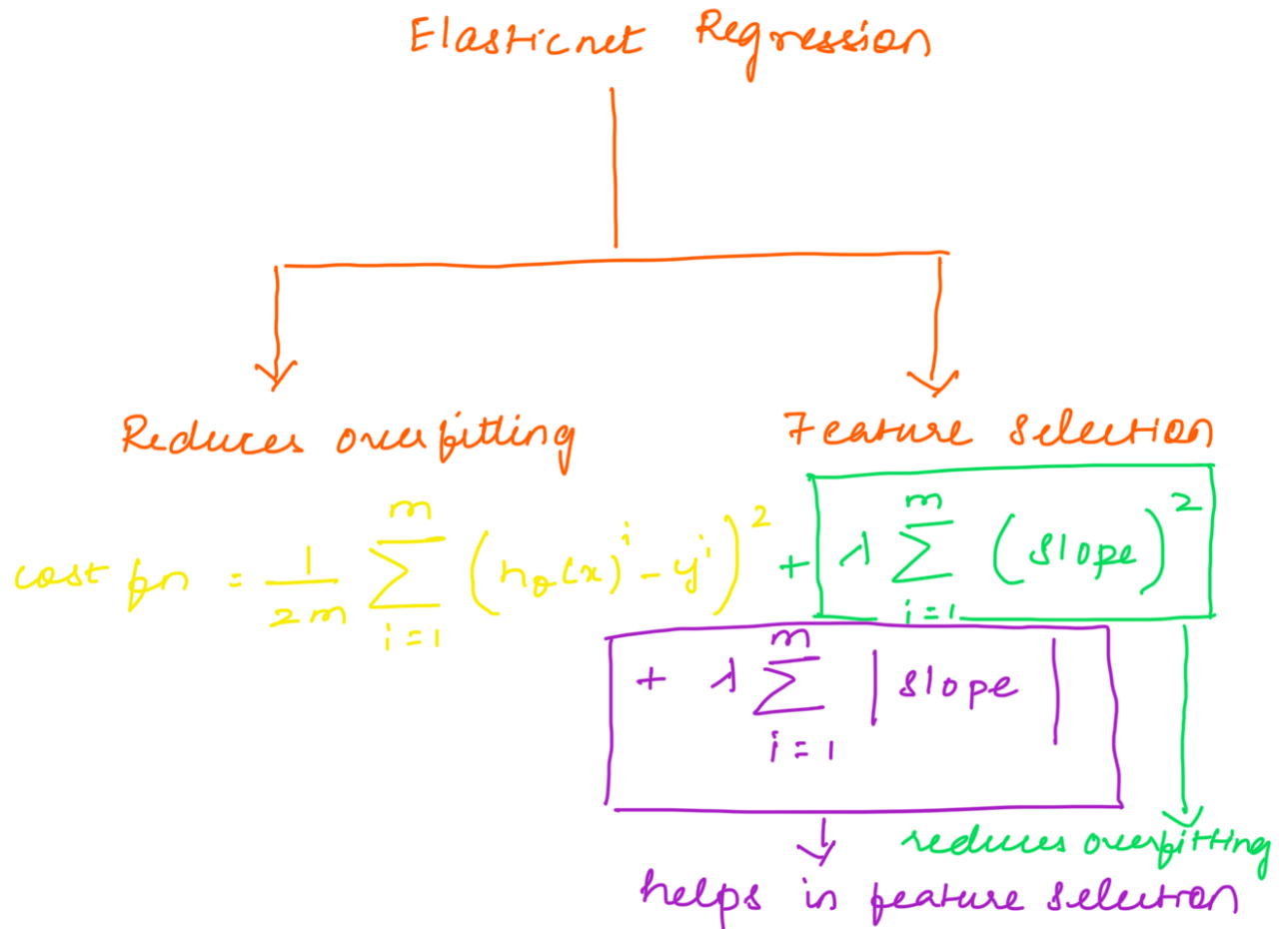
The feature that are not that important will be deleted and the features that are important will be considered.



$\theta$  value becoming zero → coefficient is actually becoming zero. → In short we are trying to remove the specific feature.

### ③ Elasticnet Regression

→ It is the combination of Lasso and Ridge regression



Why we are using Ridge, Lasso and Elasticnet regression

→ Hyperparameter tuning the Linear regression.