

ridge - regression. best score -

→ WSS Fxn → Cross-Entropy

# NAIVE BAYES CLASSIFICATION

→ Based on Bayes Theorem

$$P(A/B) = \frac{P(A) * P(B/A)}{P(B)}$$

---→ This model o/p's probability so, it is very small.

so, we need to apply Normalization

$$\begin{aligned} &\hookrightarrow \text{cur} \geq 0.5 \Rightarrow 1 \\ &\quad < 0.5 \Rightarrow 0 \end{aligned}$$

---→ For e.g

$$\hookrightarrow P(\text{Yes}/x_i) = 0.13$$

$$P(\text{No}/x_i) = 0.05$$

Normalization

$$P(\text{Yes}/x_i) = \frac{0.13}{0.13 + 0.05} = 0.72 \Rightarrow 1$$



- ★ Use when feature are mostly indepe-  
-ndent.
- ★ Not great for complex Relationship.
- ★ Extremely Fast.

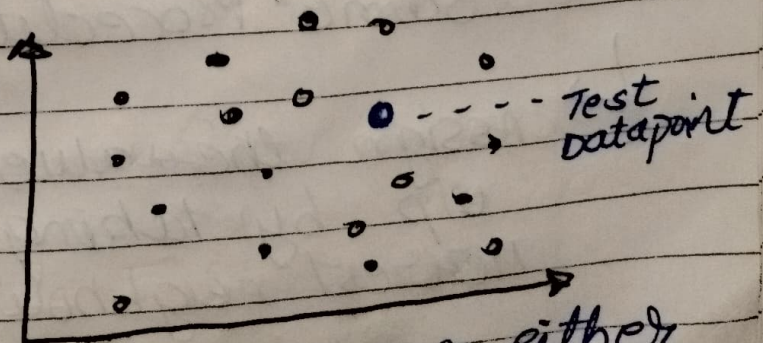
## KNN (CLASSIFICATION & REGRESSION)

↳ K Nearest Neighbour.

### KNN CLASSIFICATION

Very Easy

- i) select Hyperparameter K.
- ii) Put the Test Data point on the Graph.



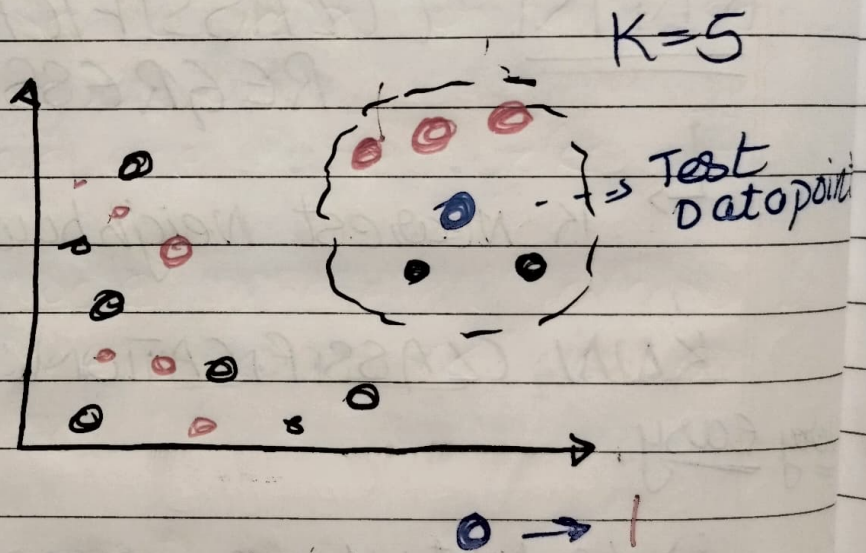
- iii) Check K Nearest Neighbour either



on the basis of Euclidean Distance or Mahhatten Distance

- (iv) From that  $K$  nearest neighbour which are of more class, we will assign the label to our datapoint of that class.

●  $\Rightarrow 1$   
●  $\Rightarrow 0$



## KNNI REGRESSION

↳ Same Procedure

↳ Assign the value to new datapoint o/p by taking average of  $K$  nearest neighbour o/p.



## Problem with KNN

- ① Should not have outliers.
- ② Should not have imbalanced dataset

## Eucledian Distance

$$\hookrightarrow \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$

## Manhattan Distance

$$\hookrightarrow |x_2 - x_1| + |y_2 - y_1|$$

## KNN Regression

$\hookrightarrow$  Loss  $\Rightarrow$  ~~Use~~ No loss fcn

## KNN Classification

$\hookrightarrow$  Don't have formal loss function