

# K-Nearest Neighbours (KNN)

KNN :-

↳ classification

$k=3$

$P_1 \Rightarrow$

we need to take

3 nearest points to  $P_1$

$P_1 \Rightarrow [+ve, +ve, +ve]$

↳ all the data points near to  $P_1$  are

+ve's; So  $P_1$  belongs to +ve data point

(Note!:- Maximum Voting)

$P_2 \Rightarrow$

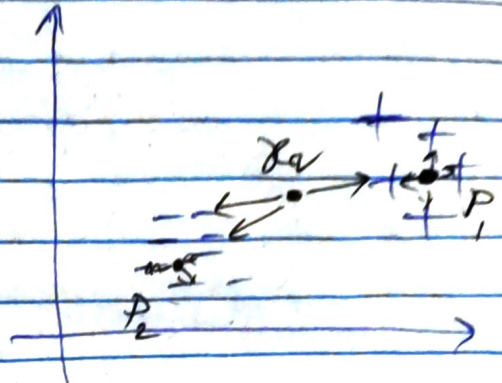
$P_2 = [-ve, -ve, -ve]$

all are -ve points. i.e. it has maximum votes for -ve point (Majority Voting)

$\Rightarrow P_2$  belongs to -ve data point

Task!:- Given any query point  $x_q$  we want to find its class by looking at its K-nearest neighbours.

↳ We will consider K-nearest datapoints to  $x_q$  (which are nearest to  $x_q$ ) <sup>Euclidean distance</sup> and we will do Majority Voting.



Algorithm:-

↳ Given  $\rightarrow D_{\text{Train}}, K, x_q$   
↓                      ↓                      ↓  
Training data    # of Neighbours    find the class for the query point

Lets  $K=3$  (How to find best  $K$  we will discuss later)

Step 1:-  $K=3 \Rightarrow 3$  nearest neighbours ~~for~~ to  $x_q$

(Here we are not training any data, directly we are testing it)

$\Rightarrow$  We will get  $K$ -Nearest Neighbours to  $x_q$

Step 2:-

we will get the class for

$K$ -Nearest points & do the

Majority Voting

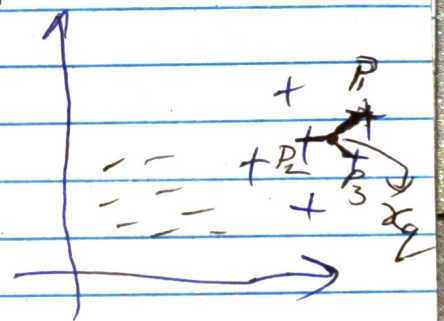
Here

$K$ -Nearest Neighbours are  $P_1, P_2$  and  $P_3$ .

$\Rightarrow P_1, P_2, P_3 \rightarrow$  belongs to +ve class.

$\Rightarrow x_q = [+ve, +ve, +ve]$   
+ve is having 3

As per Majority Voting  $x_q$  is positive





Q) How to find the K-Nearest Neighbours

A  $\rightarrow$  Euclidean distance/...

Here Computer has to know which are the nearest neighbours.

$\rightarrow$  So it will calculate distance from each query point to each data point.

$\rightarrow$  Now i will sort it <sup>the</sup> distance in Ascending order.

$\rightarrow$  Now i know which data points having minimum distance to the query point.

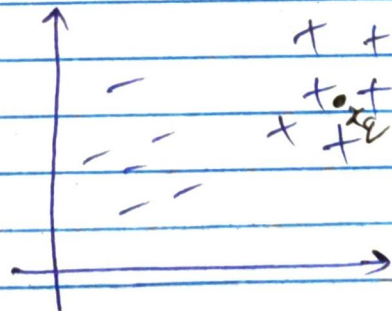
$\rightarrow$  Now identify the class it belongs to and do the Majority Voting.

Ex:-

1  $\rightarrow$  150  
2  $\rightarrow$  5  
3  $\rightarrow$  0.5  
4  $\rightarrow$  100  
5  $\rightarrow$  50

999  $\rightarrow$  0.1  
1000  $\rightarrow$  0.9

sort asc

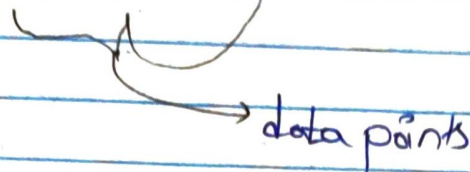


For each datapoint, i am calculating the distance from  $x_q$ .

Here i am having 1000 datapoints

$$= [0.1, 0.5, 0.9, 5, \dots, 150]$$

$\downarrow$        $\downarrow$        $\downarrow$        $\downarrow$   
 999    3      1000      1

  
 data points

$$= [999, 3, 1000]$$

$\downarrow$        $\downarrow$        $\downarrow$   
 [ +ve   +ve   +ve ]

$$\Rightarrow \begin{matrix} +ve = 3 \\ -ve = 0 \end{matrix} \rightarrow \text{Majority Voting}$$

Voting  
 $\downarrow$

$$\underline{\underline{x_q = +ve}}$$

KNN!  $\rightarrow$

Given  $\rightarrow D_{\text{train}}, K, x_q$

1. Initialize distance = [ ]

2. for each datapoint in  $D_{\text{train}}$ :

Compute distance of  $x_q$  from dataset

(and we will store in an array)

distance.append(distance, index of datapoint)

O/p of 2.  $\Rightarrow$

$$[(150, 1), (5, 2), (0.5, 3), (100, 4), \dots, (0.1, 999), (0.9, 1000)]$$

(refer to the prev page)

3. Sort the distance list according to distances

$$[(0.1, 999), (0.5, 3), (0.9, 1000), \dots, (150, 1)]$$

$\swarrow$        $\searrow$   
 Distance of  $x_q$  from Data point in  $D_{\text{train}}$       Index of  $D_{\text{train}}$



4. pick first K tuples from distance list

K nearest neighbors = distance[:K]  
(lets K=3)

5. Initialize

pos-count = 0

neg-count = 0

for neighbour in nearest-neighbors:

if class is +ve

↳ pos-count += 1

else

neg-count += 1

6. if (pos-count > neg-count):  
+ve class

else

-ve class.

**\*\*** Note:- So, in this algorithm there is no test data.

↳ It completes training data (memorizes the complete training data)

↳ No, need of train-test separation.

entire training data → KNN → memorizes the complete data & stores it in Ram

Testing  $m_c$  ← Logistic Reg → Min  $\log \sum$  Signed distance

Testing  $m_c$  ← Linear Reg → Min Sum of Squared error