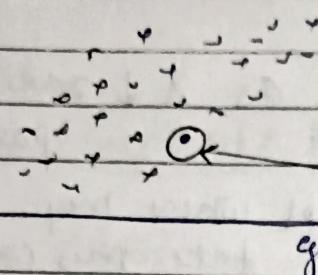


## (Black box model) K-Nearest Neighbour (KNN) (classifying)

19



Blue  $\rightarrow$  placement no gaze (1)

Black  $\rightarrow$  NO placement (0)

new query

gp

Suppose a new query comes then we will calculate its distance (euclidean) from all points. Now we sort it in ascending order now do majority count

lets for pt 1 dist of query  $\rightarrow$  0       $k=3$   
 2  $\rightarrow$  1  
 3  $\rightarrow$  1

so it has been placed

Since among the 3 nearest pt majority has placed so new query also get placed

• How to select  $k$ ?

number

$\rightarrow k = \sqrt{n}$      $n = \text{no. of observations}$

$$\sqrt{n} = \frac{20}{T}$$

don't take even

either take 19 or 21

experimentally  
cross validate

$$n = 1000$$

800

200

apply diff KNN

$$KNN = 1$$

2

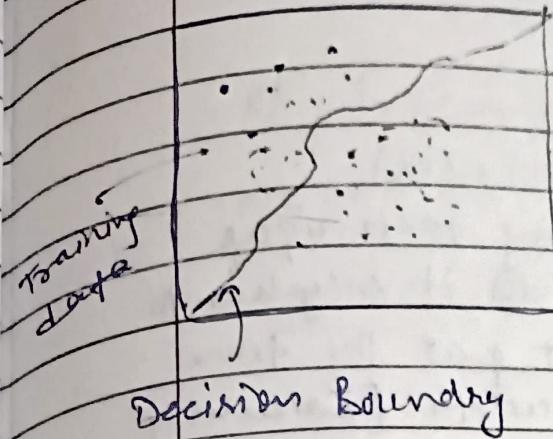
3

1

The one which gives best result take that value of  $K$

## • Decision Surface

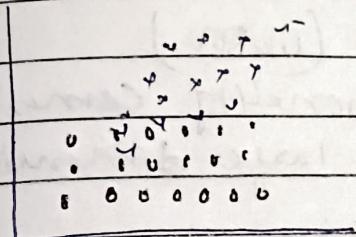
" It is a tool used in classification algorithm



Suppose we consider a 2D data.  
So there are two surfaces one for 0 & other for 1.

Now if any new pt. comes, we can tell whether it will fall in region 0 or region 1.

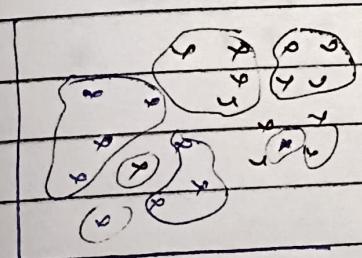
→ now its made -



- we plot the training data
- then generate a numpy meshgrid
- then apply kNN which tells whether 0 or 1
- Then we make pixels & show it as an image.

## • Overfitting and Underfitting in KNN

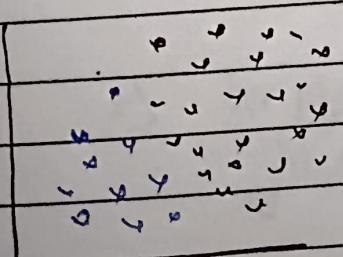
$n = 200$



$K=1$

→ This is overfitting  $\Leftrightarrow$   
"high variance"

$K=200$



" If a new pt comes then now we will see its distance from all pts.

Since black is in majority  $\Leftrightarrow$   
So it will always assign

→ underfitting

low value of  $k \rightarrow$  overfitting  
high  $\cdot \cdot \cdot \rightarrow$  underfitting

## • Limitations of KNN -

- 1) Large dataset  $\Rightarrow$  (500000, 100)

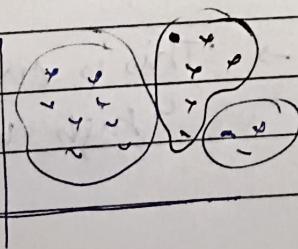
KNN is lazy learning technique  
means training is fast as it only has to  
store the points. But at the time  
of prediction all calculation (Calculating  
dist from all pts, sorting etc) happens so  
it becomes slow

- 2) High Dimensional data (like SV)

The curse of dimensionality comes into play which says at large dimension dist are not reliable.

Since KNN is based on dist. So it is also not reliable

- 3) Doesn't work good with outliers.



- 4) Non homogeneous Scales -

<del>exp</del>	<del>Cal.</del>
0-25	10K - 11ack
<u>small scale</u>	<u>large scale</u>

- 5) Imbalanced data set  
 ↳ 93% Yes  
 ↳ 2% No

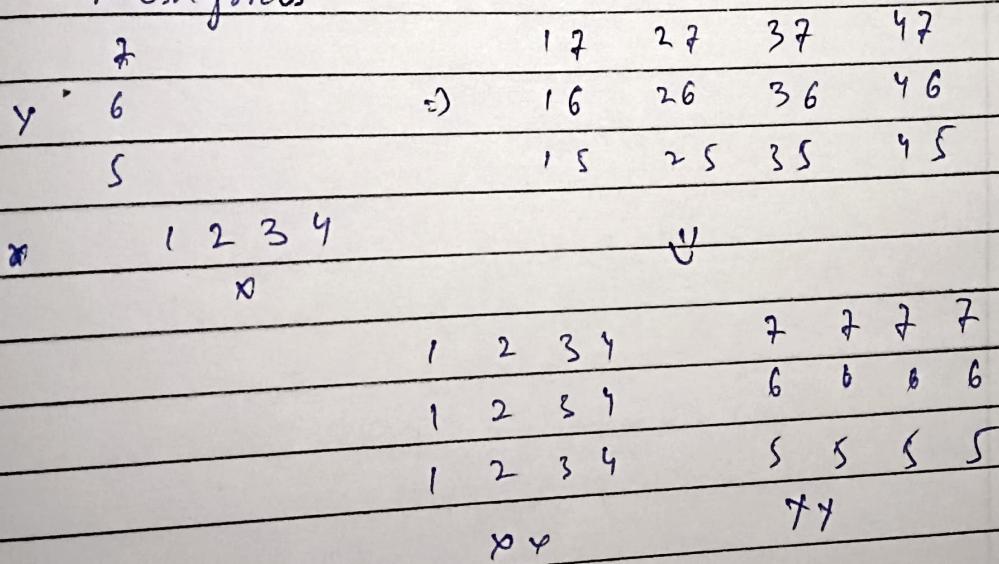
6) ↓ inference  
 not good for  $\therefore$  (Black Box model)

### Decision Boundary:-

- ↳ we can draw decision boundary for all classification algo including Neural Networks
- ↳ It can be both linear or non-linear
- ↳ for high dimension problems, it act as an hyperplane.

$\Rightarrow$  Voronoi Diagram - ✓

### → Meshgrids



$\Rightarrow$  for plotting Decision Boundary  
 just watch lecture