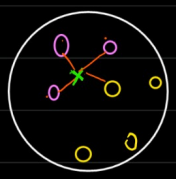
K-Nearest Neighbour

① KNN classifier
② KNN Regressor

* KNN Classifier

f_1	f_2	y
—	—	0
—	—	1
—	—	0
		0
		1
		1

Scen-1



Scen-2

instead of 4 nearest dp, take 7 nearest dp.

* → yellow.

Scen-1

x (Assign a class)

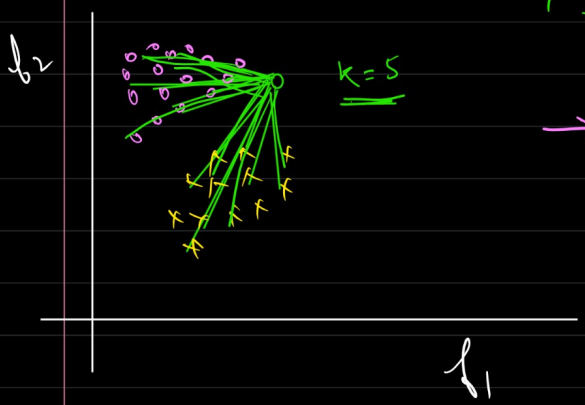
* 4 nearest point

Pink 3 | yellow

majority ⇒ Pink

* ⇒ Pink class

K → nearest dp.

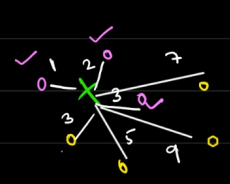


→ As K changes, the class of new dp might also change.

⇓

K is a hyperparameter.

4 nearest dp



0	1
0	2
□	3
0	3
0	5
0	7
0	9

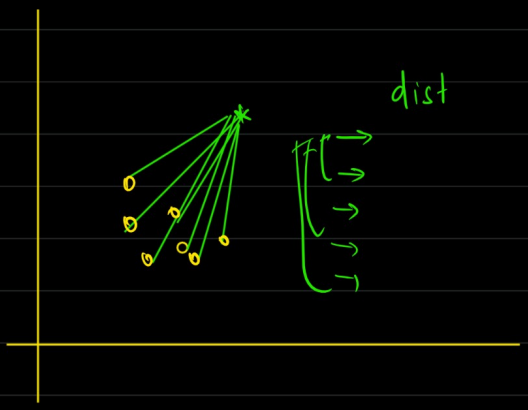
→ Prediction will be 0

binary class. → K as odd value.

↳ 5, 7, 9, 11, 13

2 class:

How K?? ⇒ K is hyperparameter.



Algorithm

① Plot the dp in n-d space.

② Initialise the K-value. (No of neighbours you want to consider)

K ∈ I ⟶ ∞ (Generally k > 3)

→ Calculate the distance of new dp wrt to all dp's.

→ Sort the distance

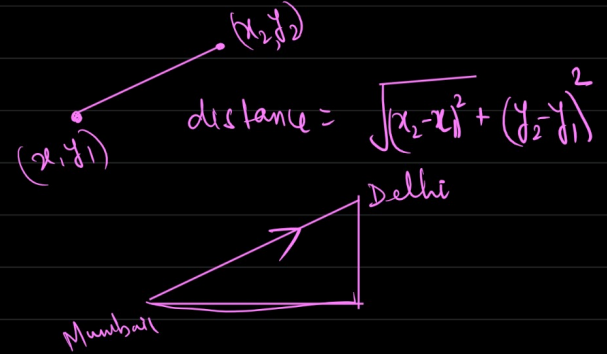
→ based on k find class of that k nearest dp's

③ Find the mode of the class

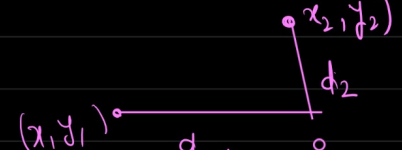
④ Assign the class.

distance

① Euclidean distance.

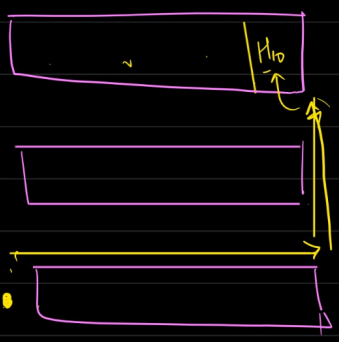


② Manhattan distance.

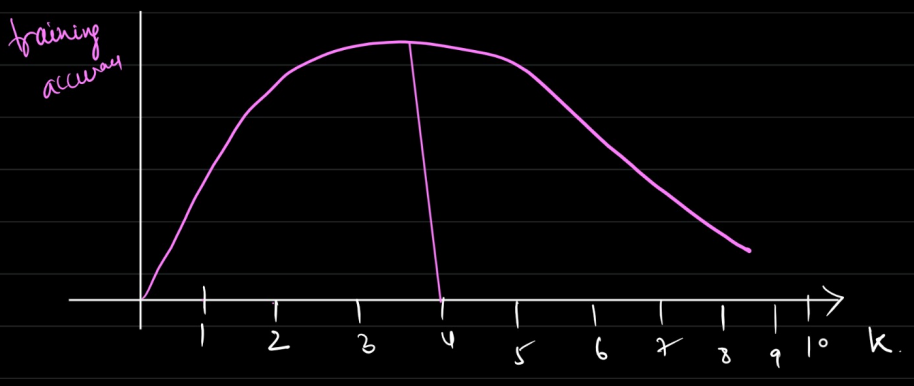


distance = $d_1 + d_2$

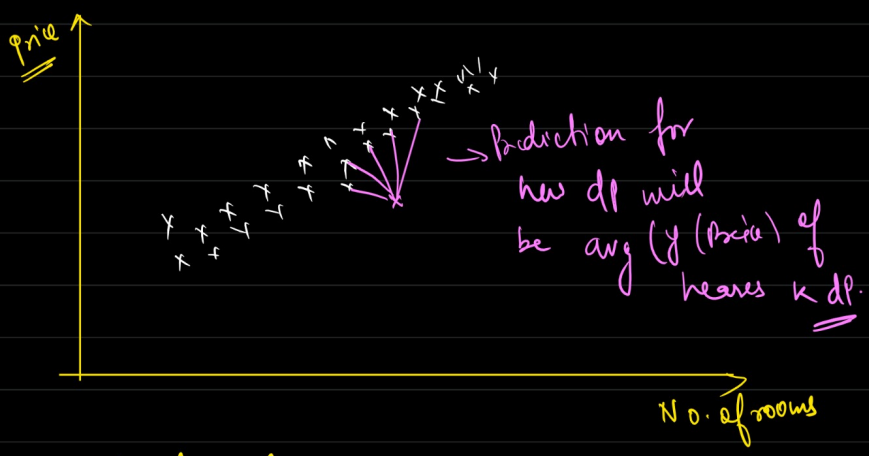
$$d(x,y) = \sqrt{(x_1-y_1)^2 + (x_2-y_2)^2 + \cdots (x_n-y_n)^2}$$



How to decide K.



* KNN Regressor



→ Prediction for new dp will be avg (y (price) of nearest K dp.

* Advantage of KNN
  - → Easy to understand / very intuitive.
  - → Performance of model in terms evaluation metric is good

* disadvantage
  - → Lazy learner.
    - ⇓
    - Computation time is more.

ML
- eager learner
- lazy learner
  - • fit → Coeff's
  - • predict
  - → you are calculating Parameter of model in real time.
    - ⇓
    - KNN.

$\beta_0 + \beta_1 x_1 + \beta_2 x_2$

* All the model parameter are calculated while training & used for prediction.

Some more lazy learners

1. Locally weighted learning (LWL)
2. Case-Based Reasoning
3. Lazy Bayesian rules

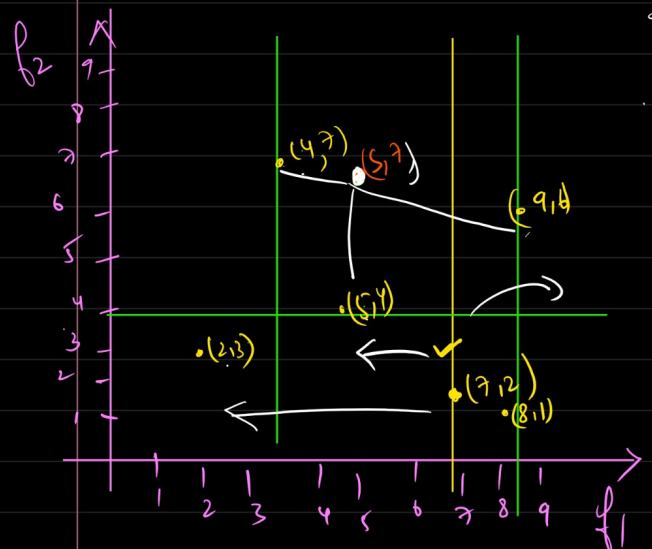
* Variants of KNN.

1. K-D tree
2. Ball tree.

Why KNN is lazy learner ?

→ In real time the distance of test data from each of dp is calculated ⇒ Brute force

* Variants of KNN



f_1	f_2
7	2
5	4
9	6
2	3
4	7
8	1

* K-D-tree

↳ Partition the data in a binary tree.

Step-1 sort the feature f_1 & f_2

Step-2 Calculate the median of f_1 and f_2.

Step-3 Partion the data based on median

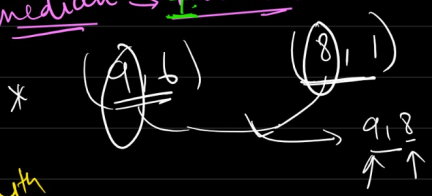
Step-4 Partition all the data recursively in each of subpartion.

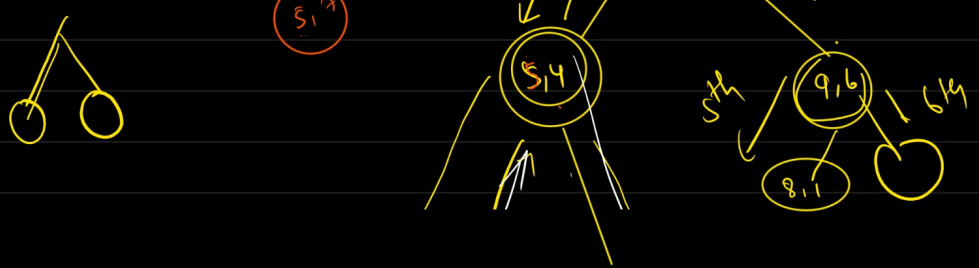
↓

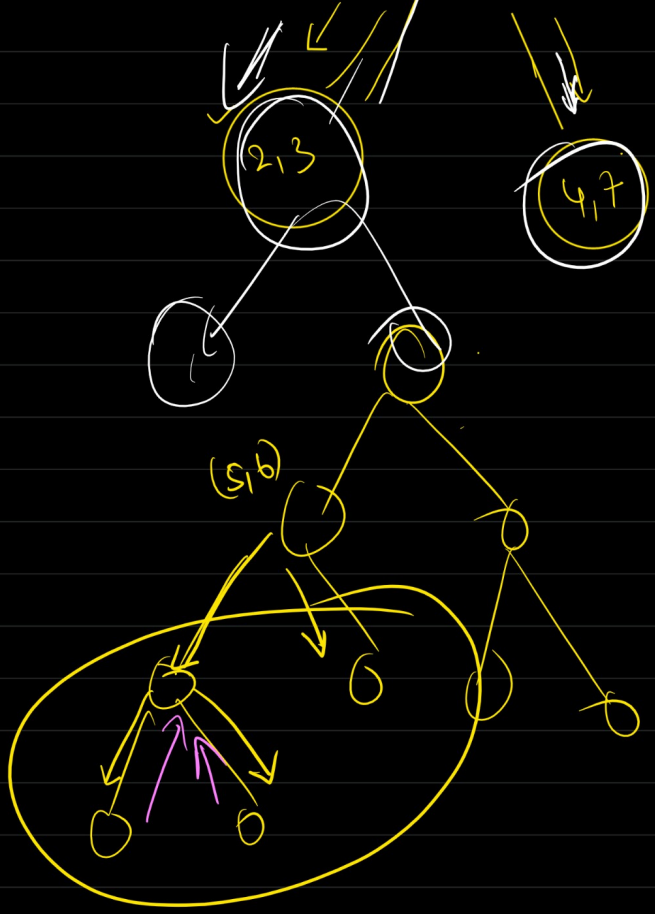
Until in one partion only on dp is there.

→ f_1 - 2, 4, 5, 7, 8, 9.

→ median → 7 as media







BST

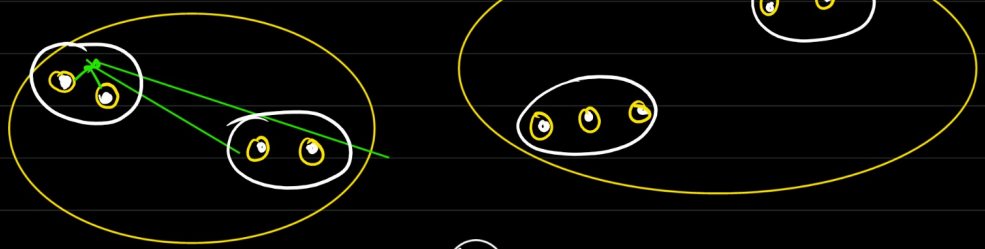
Binary $\Rightarrow O(\log n)$

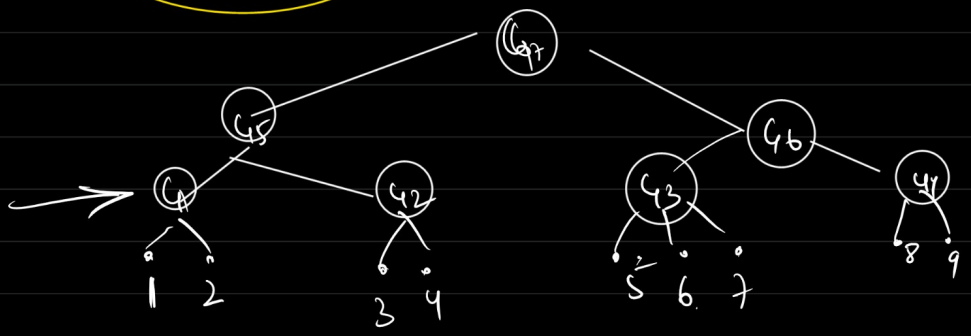
worst = $O(n)$ → Brute force

* Conclusion → reduced the search space.

(2) Ball tree

Advantage
→ only need to calculate distance of the nearest group element.





if small dataset → Use kNN

→ read all csv in one go

*.csv

↓

Concatenate