

# #5 Writing Our First Classifier

## L K-Nearest Neighbours (KNN)

Step-1  $\Rightarrow$  Comment Out Imports

We will comment to a classifier which we imported during writing pipeline.

Step-2  $\Rightarrow$  Implement class for classifier

Class ScrappyKNN():

~~def fit(self, x\_train, y\_train):~~

Step-3  $\Rightarrow$  Understand the Interface [under what method we need to implement]

S3 def fit(self, x\_train, y\_train):  
self.x\_train = x\_train  
self.y\_train = y\_train

S3 def Predict(self, x\_test):

predictions = []

for each row in x\_test:

label = self.closest(row)

predictions.append(label)

return predictions.

Step 4: Get pipeline working

accuracy [33%]

for Random Classifier

def closest(self, row):

best\_dist = euc(row, self.x\_train[0])

best\_index = 0

for i in range(1, len(self.x\_train)):

dist = euc(row, self.x\_train[i])

if dist < best\_dist:

best\_dist = dist

best\_index = i

return self.y\_train[best\_index]

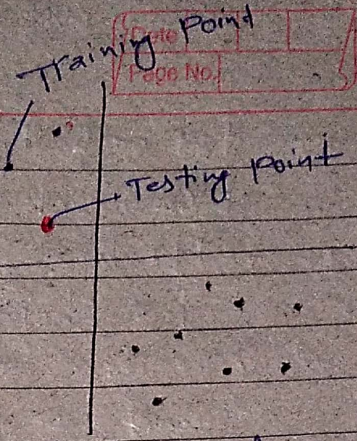
Calculate the distance from test point to 1st training point

REST CODE WILL BE SAME AS PREVIOUS CLASS.



## Step 5 → Intro to K-NN

For given testing point, the algorithm finds the closest training point.



Predictions → The testing point is then predicted to have the same label as its nearest neighbor.

Role of K → If there is a tie in distance or to make predictions more robust, K comes into play. K represents the number of neighbours to consider when making a prediction.

For example →  $K=3$ , the classifier looks at three closest points. But for simplicity we take  $K=1$ .

## Step 6 → Measure distance

- To determine the "closest distance" neighbours, a distance metric is required. The video introduces
- Use 'Euclidean distance' → measure straight line distance b/w two points  $A^2 + B^2 = C^2$ . It can be used by implementing SciPy library.

$$d(a,b) = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$

if you want for all the visualization 1D - 2D - 3D ----

from scipy.spatial import distance

def func(a,b):

a is point from training data  
b is point from testing data

return distance.euclidean(a,b)



## Step 7 → Implement nearest neighbor algorithm

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- For each test point, the classifier calculates the ~~Euc~~ Euclidean distance to all training points.
- It then iterate through these distance to find the shortest one.
- variables are maintained to track the shortest dist found so far and the index of the training point corresponding to that shortest dist.
- finally using the index, the label of the closest training example is retrieved and returned as the prediction for this test point.

## Step 8 Run Pipeline

Accuracy got 90% +

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