

KNN - Algorithm  
ab Assignment 3:- Write a program to implement K-Nearest Neighbor algorithm to classify the Iris Data set. Print both correct and wrong predictions. Java/Python ML Library classes can be used for this problem.

### Algorithm Steps

Step 1:- Load/Read/Scan the training Data set, and Testing data set

Step 2:- Next, Choose/Input the value of  $K$  i.e. the nearest data points.  $K$  can be any integer

Step 3:- For each point in the test data do the following

3.1 → Calculate distance b/w test data & each row of training data with the help of ~~any~~ Euclidean Distance ~~method~~

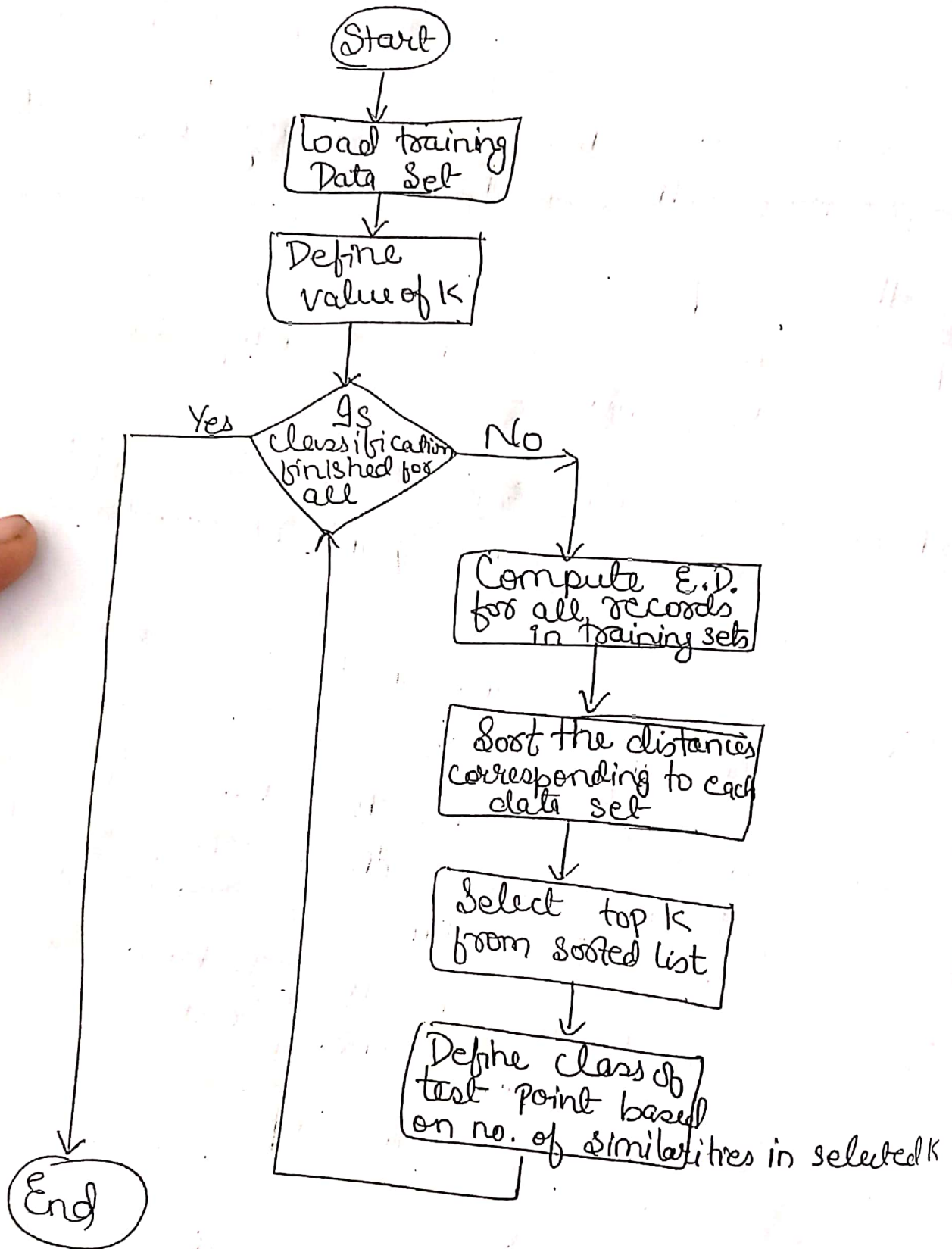
3.2 → Now, based on the distance value, sort them in ascending order.

3.3 → Next, choose the top  $K$  rows from the sorted array.

3.4 → Now, Assign a class to the test point based on most frequent class of these rows.

Step 4:- End

## Flow Chart



## K-NN Program

blame:-

<u>X</u>	<u>Y</u>	<u>Class</u>
2.78	2.55	0
1.46	2.36	0
3.39	4.04	0
1.38	1.85	0
3.06	3	0
7.62	2.75	1
5.33	2.08	1
6.92	1.77	1
8.67	-0.24	1
7.67	3.5	1

- ① KNN Theory
- ② KNN Algo steps
- ③ KNN flowchart
- ④ Problem solving
- ⑤ Code Explanation along with Python basics
- ⑥

code → Using Library:-

```
from sklearn.neighbors import KNeighborsClassifier  
knn = KNeighborsClassifier(n_neighbors=1)  
knn.fit(X_train, Y_train)  
pred = knn.predict(X_test)  
pred.
```



## Code

```
import numpy as np
import pandas as pd
from math import sqrt

dt = pd.read_csv("KNN1.csv")
data set = np.array(dt)
dt.head()
X = data set → training
Y = data set [1:] → testing
```

```
# print("X=", X)
```

```
# print("Y=", Y)
```

```
→ k = 3
```

```
ds = []
```

```
for i in range(len(X) - 1):
```

```
    d = np.subtract(X[i], Y)
```

```
    t_s = np.square(d)
```

```
    ss = np.sum(t_s)
```

```
    ds = np.append(ds, [ss])
```

```
# print(ds)
```

```
as_d = np.argsort(ds)
```

```
print(as_d)
```

```
sd = []
```

```
for i in range(k):
```

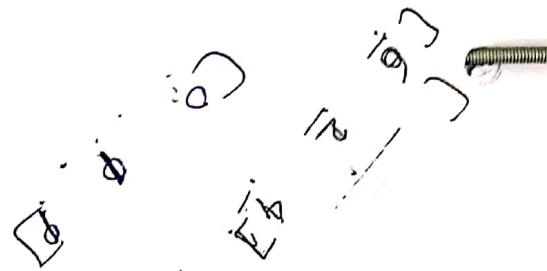
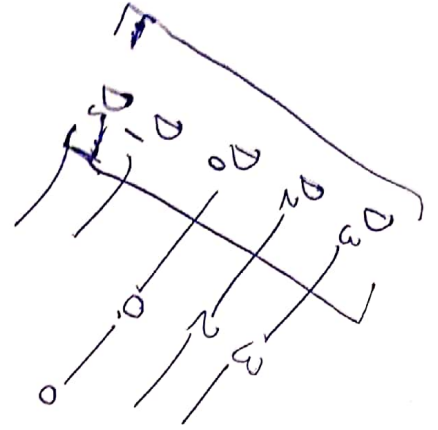
```
    sd.append(X[as_d[i]], -1)
```

```
# print(sd)
```

```
wx = sd.count(2.0)
```

```
wy = sd.count(1.0)
```

```
wz = sd.count(0.0)
```



$$X = (\text{sum}(X1 - X2)^2)^{0.5}$$

```
if (wx > wy and wx > wz)
```

```
    print('class label is 2.0')
```

```
elif (wy > wz):
```

```
    print('class label is 1.0')
```

```
else:
```

```
    print('class label is 0.0')
```

# Assignment-3

**Objective:** Write a program to implement K-Nearest Neighbor Algorithm to classify the Iris Data Set. Print both correct and wrong predictions. Java/python Library Classes can be used for this problem.

## Step-1: Importing necessary libraries.

- numpy for numerical operations.
- pandas for reading csv.

```
In [1]: import numpy as np
import pandas as pd
```

## Step-2: Reading csv file.

```
In [6]: df = pd.read_csv("kNN1.csv")
```

## Step-3: Inspection of data and understanding it.

```
In [7]: df.head()
```

```
Out[7]:
```

	S1	S2	S3	S4	Flower type	Flower Name
0	5.1	3.5	1.4	0.2	0	Iris-setosa
1	4.9	3.0	1.4	0.2	0	Iris-setosa
2	4.7	3.2	1.3	0.2	0	Iris-setosa
3	4.6	3.1	1.5	0.2	0	Iris-setosa
4	5.0	3.6	1.4	0.2	0	Iris-setosa

About Code (Just for knowledge):

- This df.head() code prints first 5 lines of your dataframe.

About Data:

- Total 5 columns are there.
- Sepal length, Sepal width, Petal length, Petal width are features.
- Elower type classification will be done for Iris Setosa as 0, Iris versicolor as 1 and Iris-viginica as 2,in our target column.
- Output of the program would be the flower type classification

```
In [8]: df.shape
```

```
Out[8]: (30, 6)
```

This shows that in our dataframe total 30 rows and 6 columns are available.

```
In [9]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 30 entries, 0 to 29
Data columns (total 6 columns):
```

```
] (30, 6)
```

This shows that in our dataframe total 30 rows and 6 columns are available.

```
] df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 30 entries, 0 to 29
Data columns (total 6 columns):
#   Column          Non-Null Count  Dtype
---  -
0   S1               30 non-null    float64
1   S2               30 non-null    float64
2   S3               30 non-null    float64
3   S4               30 non-null    float64
4   Flower type      30 non-null    int64
5   Flower Name      30 non-null    object
dtypes: float64(4), int64(1), object(1)
memory usage: 1.5+ KB
```

This shows details of your dataframe:

- Data type and non null value of each feature and target.
- memory usage of this dataframe.

#### Setp-4: Implementing K-Nearest Neighbor Algorithm

- Input the training data and testing data.
- Choose/Input the value of 'k'.
- Compute euclidean distance between testing data sets and training data sets.
- Sort these distances and select/choose k-distance from the top.
- Assign the flower type to the test data based on most frequent class of these rows.

```
1]: dg = pd.read_csv("C:\\Users\\Bhuvnesh\\Desktop\\kNN3.csv")
dataset=np.array(dg)

X = dataset
Y = dataset[19]
print("X=",X)
print("Y=",Y)
k=3
#ed
ds= []
for i in range (len(X)-1):
    d = np.subtract(X[i],Y)
    t_s = np.square(d)
    rr = np.sum(t_s)
    ds = np.append(ds,[rr])

print('distances=',ds)
asd = np.argsort(ds)
print(asd)
sd=[]
for i in range(k):
    sd.append(X[asd[i],-1])
print(sd)
```

```
wx=sd.count(2.0)
wy=sd.count(1.0)
wz=sd.count(0.0)
```

```
#w=max(wx,wy,wz)
```

```
if (wx>wy and wx>wz):
    print('class label is 2.0')
elif (wy>wz):
    print('class label is 1.0')
else:
    print('class label is 0.0')
```

```
x= [[5.1 3.5 1.4 0.2 0. ]
```

```
[4.9 3. 1.4 0.2 0. ]
```

```
[4.7 3.2 1.3 0.2 0. ]
```

```
[4.6 3.1 1.5 0.2 0. ]
```

```
[5. 3.6 1.4 0.2 0. ]
```

```
[5.4 3.9 1.7 0.4 0. ]
```

```
[4.6 3.4 1.4 0.3 0. ]
```

```
[5. 3.4 1.5 0.2 0. ]
```

```
[4.4 2.9 1.4 0.2 0. ]
```

```
[4.9 3.1 1.5 0.1 0. ]
```

```
[7. 3.2 4.7 1.4 1. ]
```

```
[6.4 3.2 4.5 1.5 1. ]
```

```
[6.9 3.1 4.9 1.5 1. ]
```

```
[5.5 2.3 4. 1.3 1. ]
```

```
[6.5 2.8 4.6 1.5 1. ]
```

```
[5.7 2.8 4.5 1.3 1. ]
```

```
[6.3 3.3 4.7 1.6 1. ]
```

```
[4.9 2.4 3.3 1. 1. ]
```

```
[6.6 2.9 4.6 1.3 1. ]
```

```
[5.2 2.7 3.9 1.4 1. ]
```

```
[6.3 3.3 6. 2.5 2. ]
```

```
[5.8 2.7 5.1 1.9 2. ]
```

```
[7.1 3. 5.9 2.1 2. ]
```

```
[6.3 2.9 5.6 1.8 2. ]
```

```
[6.5 3. 5.8 2.2 2. ]
```

```
[7.6 3. 6.6 2.1 2. ]
```

```
[4.9 2.5 4.5 1.7 2. ]
```

```
[7.3 2.9 6.3 1.8 2. ]
```

```
[6.7 2.5 5.8 1.8 2. ]
```

```
[7.2 3.6 6.1 2.5 2. ]
```

```
[7.2 3.6 6.1 2.5 2. ]
```

```
[7.2 3.6 6.1 2.5 2. ]
```

```
[7.2 3.6 6.1 2.5 2. ]
```

```
[7.2 3.6 6.1 2.5 2. ]
```

```
[7.2 3.6 6.1 2.5 2. ]
```

```
[7.2 3.6 6.1 2.5 2. ]
```

```
[7.2 3.6 6.1 2.5 2. ]
```

```
[7.2 3.6 6.1 2.5 2. ]
```

```
[7.2 3.6 6.1 2.5 2. ]
```

```
[7.2 3.6 6.1 2.5 2. ]
```

```
[7.2 3.6 6.1 2.5 2. ]
```

```
[7.2 3.6 6.1 2.5 2. ]
```

```
[7.2 3.6 6.1 2.5 2. ]
```

```
y= [5.2 2.7 3.9 1.4 1. ]
distances= [ 9.34 8.87 9.7 8.72 9.54 8.32 9.31 8.73 9.37 8.7 4.13 2.06
4.06 0.27 2.2 0.63 2.25 0.7 2.5 0. 8.19 3.05 9.19 5.3
7.03 14.63 1.58 11.37 7.06]
[19 13 15 17 26 11 14 16 18 21 12 10 23 24 28 20 5 9 3 7 1 22 6 0
8 4 2 27 25]
[1.0, 1.0, 1.0]
class label is 1.0
```

```
[ ]:
```