

Lab Sheet 2

KNN Implementation

S Abhishek

AM.EN.U4CSE19147

Load Dataset

```
import pandas as pd
```

```
data = pd.read_csv("/content/Diabetes.csv")
```

Display First 10 Rows

```
data.head(15)
```

	Pregnancies	Glucose	BloodPressure	...	DiabetesPedigreeFunction	Age	Outcome
0	6	148	72 ...		0.627 50	1	
1	1	85	66 ...		0.351 31	0	
2	8	183	64 ...		0.672 32	1	
3	1	89	66 ...		0.167 21	0	
4	0	137	40 ...		2.288 33	1	
5	5	116	74 ...		0.201 30	0	
6	3	78	50 ...		0.248 26	1	
7	10	115	0 ...		0.134 29	0	
8	2	197	70 ...		0.158 53	1	
9	8	125	96 ...		0.232 54	1	
10	4	110	92 ...		0.191 30	0	
11	10	168	74 ...		0.537 34	1	
12	10	139	80 ...		1.441 57	0	
13	1	189	60 ...		0.398 59	1	
14	5	166	72 ...		0.587 51	1	

[15 rows x 9 columns]

Split Data Set into Train and Test

```
train_percent = round((data.shape[0]*80)/100)
print("Train Data :",train_percent)
```

```
test_percent = round((data.shape[0]*20)/100)
print("Test Data :",test_percent)
```

Train Data : 614

Test Data : 154

```
XTrain = data[:train_percent]
XTest = data[train_percent:]
```

```
print("Train Set :", XTrain.shape, "\nTest Set:", XTest.shape)
```

Train Set : (614, 9)

Test Set: (154, 9)

Split the dataset into 80% training and 20% testing

```
from sklearn.model_selection import train_test_split
```

```
Y = data.Outcome # Only Target Label
```

```
X = data.drop('Outcome',axis=1) # Features
# axis = Whether to drop 0 - labels from the index or 1 - columns.
```

```
x_train,x_test,y_train,y_test=train_test_split(X,Y,test_size=0.2)
```

```
print(x_train.head(), "\n", y_train.head())
```

	Pregnancies	Glucose	BloodPressure	...	BMI	DiabetesPedigreeFunction	Age
473	7	136	90 ...	29.9		0.210	50
465	0	124	56 ...	21.8		0.452	21
72	13	126	90 ...	43.4		0.583	42
127	1	118	58 ...	33.3		0.261	23
204	6	103	72 ...	37.7		0.324	55

```
[5 rows x 8 columns]
```

```
473  0
```

```
465  0
```

```
72   1
```

```
127  0
```

```
204  0
```

```
Name: Outcome, dtype: int64
```

```
from sklearn.preprocessing import StandardScaler
```

```
st_scaler=StandardScaler()
```

```
x_train=st_scaler.fit_transform(x_train)
```

```
x_test=st_scaler.transform(x_test)
```

- We have to execute, `scaler=StandardScaler().fit(X_Test)`

```
from sklearn.neighbors import KNeighborsClassifier
```

```
classifier = KNeighborsClassifier(n_neighbors=5)
```

```
classifier.fit(x_train, y_train)
```

```
y_pred = classifier.predict(x_test)
```

```
print(y_pred)
```

```
[1 0 0 0 0 0 0 0 0 1 1 1 1 0 0 0 1 0 0 0 0 0 1 1 0 0 1 0 0 1 1 1 0 1 0 1  
 1 0 0 0 1 0 0 0 0 1 0 0 0 0 0 1 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0  
 0 0 0 0 1 0 0 0 1 1 0 0 0 0 0 0 1 0 0 1 0 1 0 0 0 0 1 1 0 0 0 1 0 1 1 0 1  
 0 0 0 1 1 0 0 1 1 0 1 1 0 0 0 1 1 0 0 0 1 0 0 0 1 1 1 1 0 0 0 1 0 0 1 0 0  
 0 0 0 0 1 1]
```

```
from sklearn.metrics import classification_report, confusion_matrix
```

```
print(confusion_matrix(y_test, y_pred))
```

```
print()
```

```
print(accuracy_score(y_test, y_pred))
```

```
print()
```

```
print(classification_report(y_test, y_pred))
```

```
[[87 12]
```

```
[30 25]]
```

```
0.7272727272727273
```

	precision	recall	f1-score	support
0	0.74	0.88	0.81	99
1	0.68	0.45	0.54	55
accuracy			0.73	154
macro avg	0.71	0.67	0.67	154
weighted avg	0.72	0.73	0.71	154

Above confusion matrix shows that there are,

- 87 True Negative cases
- 25 True Positive cases
- 12 False Positive cases
- 30 False Negative cases

```
import numpy as np
```

```
print('Actual Outcome 0 :',np.sum(y_test==0))
```

```
print('Predicted Outcome 0 :',np.sum(y_pred==0))
```

```
print()
```

```
print('Actual Outcome 1 :',np.sum(y_test==1))
```

```
print('Predicted Outcome 1 :',np.sum(y_pred==1))
```

```
Actual Outcome 0 : 99
```

```
Predicted Outcome 0 : 117
```

```
Actual Outcome 1 : 55
```

```
Predicted Outcome 1 : 37
```

Total number of correct prediction is $87 + 25 = 112$

```
from sklearn.metrics import accuracy_score
import matplotlib.pyplot as plt
```

```
k = []
accuracy = []
for i in range(2,20,3):

    classifier = KNeighborsClassifier(n_neighbors = i)
    classifier.fit(x_train, y_train)
    y_pred = classifier.predict(x_test)

    print('Now K =',i)
    print()
    print(confusion_matrix(y_test, y_pred))
    print()

    accuracy.append(accuracy_score(y_test, y_pred))

    print(classification_report(y_test, y_pred))
    print()

    k.append(i)

plt.plot(k,accuracy)
plt.xlabel('K Value')
plt.ylabel('Accuracy')
plt.show()
```

Now K = 2

```
[[90  9]
 [32 23]]
```

	precision	recall	f1-score	support
0	0.74	0.91	0.81	99

1	0.72	0.42	0.53	55
---	------	------	------	----

accuracy			0.73	154
macro avg	0.73	0.66	0.67	154
weighted avg	0.73	0.73	0.71	154

Now K = 5

[[84 15]
[22 33]]

	precision	recall	f1-score	support
0	0.79	0.85	0.82	99
1	0.69	0.60	0.64	55

accuracy			0.76	154
macro avg	0.74	0.72	0.73	154
weighted avg	0.75	0.76	0.76	154

Now K = 8

[[88 11]
[27 28]]

	precision	recall	f1-score	support
0	0.77	0.89	0.82	99
1	0.72	0.51	0.60	55

accuracy			0.75	154
macro avg	0.74	0.70	0.71	154
weighted avg	0.75	0.75	0.74	154

Now K = 11

[[86 13]

[27 28]]

	precision	recall	f1-score	support
0	0.76	0.87	0.81	99
1	0.68	0.51	0.58	55
accuracy			0.74	154
macro avg	0.72	0.69	0.70	154
weighted avg	0.73	0.74	0.73	154

Now K = 14

[[89 10]

[31 24]]

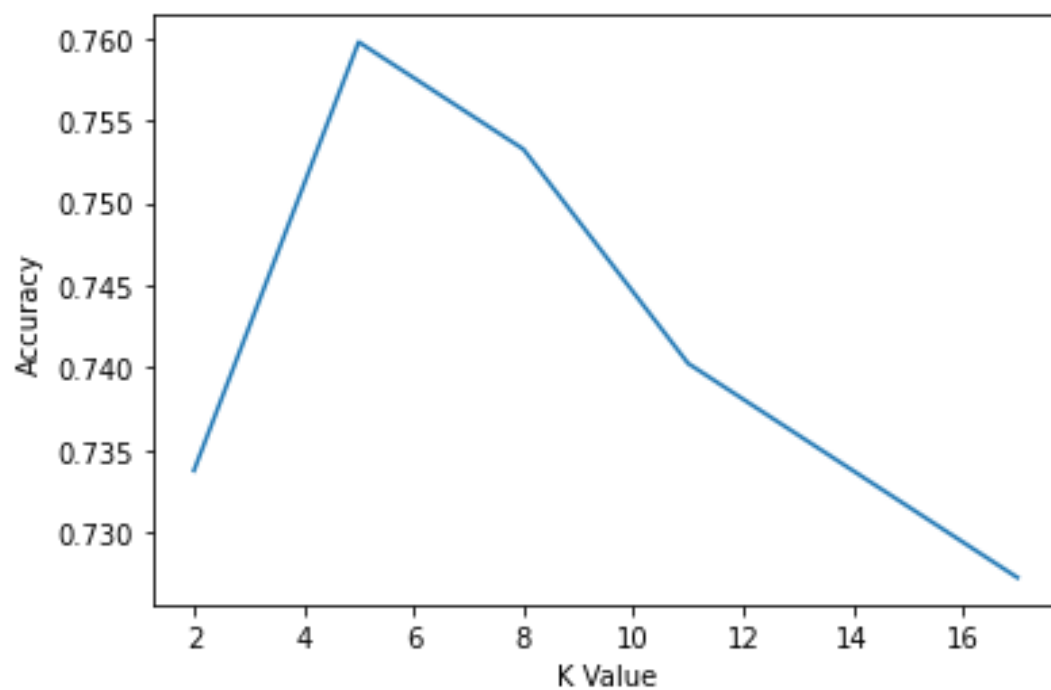
	precision	recall	f1-score	support
0	0.74	0.90	0.81	99
1	0.71	0.44	0.54	55
accuracy			0.73	154
macro avg	0.72	0.67	0.68	154
weighted avg	0.73	0.73	0.72	154

Now K = 17

[[87 12]

[30 25]]

	precision	recall	f1-score	support
0	0.74	0.88	0.81	99
1	0.68	0.45	0.54	55
accuracy	0.73			154
macro avg	0.71	0.67	0.67	154
weighted avg	0.72	0.73	0.71	154



KNN with Diabetes Dataset

```
df = pd.read_csv('/content/Diabetes.csv')
```

```
df.head()
```

	Pregnancies	Glucose	BloodPressure	...	DiabetesPedigreeFunction	Age	Outcome
0	6	148	72 ...		0.627	50	1
1	1	85	66 ...		0.351	31	0
2	8	183	64 ...		0.672	32	1
3	1	89	66 ...		0.167	21	0


```
4      0    137      40 ...      2.288  33      1
```

```
[5 rows x 9 columns]
```

```
from sklearn.model_selection import train_test_split
```

```
arr = df.values
```

```
X = arr[:,0:8]
```

```
Y = arr[:,8]
```

```
x_train, x_test, y_train, y_test = train_test_split(X, Y, test_size = 0.20,  
random_state=1)
```

```
print(x_test)
```

```
[[7.00e+00 1.36e+02 7.40e+01 ... 2.60e+01 6.47e-01 5.10e+01]
```

```
 [1.00e+00 1.51e+02 6.00e+01 ... 2.61e+01 1.79e-01 2.20e+01]
```

```
 [6.00e+00 1.09e+02 6.00e+01 ... 2.50e+01 2.06e-01 2.70e+01]
```

```
 ...
```

```
 [8.00e+00 1.88e+02 7.80e+01 ... 4.79e+01 1.37e-01 4.30e+01]
```

```
 [5.00e+00 1.36e+02 8.20e+01 ... 0.00e+00 6.40e-01 6.90e+01]
```

```
 [4.00e+00 9.00e+01 0.00e+00 ... 2.80e+01 6.10e-01 3.10e+01]]
```

```
Dist = []
```

```
row_id = 0
```

```
index = 0
```

```
euclidean = pd.DataFrame(columns=['Row ID of X Train', 'Row ID of X Test', 'Distance',  
'Outcome'])
```

```
for i in range(len(x_train)):
```

```
    d = []
```

```
        for j in range(len(x_test)):
```

```
            l = [i,j]
```

```
diff = x_train[i] - x_test[j]
dist = np.linalg.norm(diff)
```

```
l.append(dist)
l.append(y_train[i])
```

```
euclidean.loc[index] = l
index += 1
d.append(dist)
```

```
Dist.extend(d)
```

```
euclidean
```

	Row ID of X Train	Row ID of X Test	Distance	Outcome
0	0.0	0.0	28.436070	1.0
1	0.0	1.0	141.348682	1.0
2	0.0	2.0	138.930183	1.0
3	0.0	3.0	156.103828	1.0
4	0.0	4.0	62.659875	1.0
...
94551	613.0	149.0	61.204806	1.0
94552	613.0	150.0	142.338699	1.0
94553	613.0	151.0	94.890878	1.0
94554	613.0	152.0	64.717931	1.0
94555	613.0	153.0	86.965585	1.0

```
[94556 rows x 4 columns]
```

```
matrix = euclidean.values
```

```
sorted_matrix = matrix[matrix[:,2].argsort()]
```

```
data = pd.DataFrame(sorted_matrix,columns=['Row ID of X Train', 'Row ID of X  
Test','Distance','Outcome'])
```

```
data.head()
```

	Row ID of X Train	Row ID of X Test	Distance	Outcome
0	67.0	18.0	3.586227	0.0
1	440.0	29.0	3.834006	1.0
2	359.0	100.0	4.240072	0.0
3	281.0	61.0	4.244632	0.0
4	137.0	137.0	4.275231	1.0

```
y_pred = []
```

```
for i in range(X_test.shape[0]):
```

```
    out = data.loc[data["Row ID of X Test"]==i].iloc[:,5:]["Outcome"].sum()
```

```
    if out > 2:
```

```
        y_pred.append(1)
```

```
    else:
```

```
        y_pred.append(0)
```

```
accuracy_score(y_test,y_pred)
```

```
0.7337662337662337
```