## S.CHANDU AP22122040001

```
import math
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn import datasets
```

```
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
from sklearn.metrics import confusion_matrix
from collections import Counter
from sklearn.preprocessing import Normalizer
from sklearn.metrics import accuracy_score
```

iris = datasets.load\_iris() # loading iris data
iris\_df = pd.DataFrame(data= np.c\_[iris['data'], iris['target']], columns= iris['feature\_r
iris\_df.head()

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target
0	5.1	3.5	1.4	0.2	0.0
1	4.9	3.0	1.4	0.2	0.0
2	4.7	3.2	1.3	0.2	0.0
3	4.6	3.1	1.5	0.2	0.0
4	5.0	3.6	1.4	0.2	0.0

Double-click (or enter) to edit

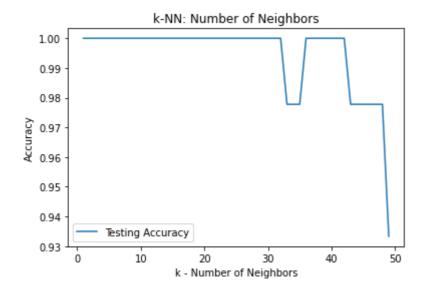
	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)
145	6.7	3.0	5.2	2.3
146	6.3	2.5	5.0	1.9
147	6.5	3.0	5.2	2.0
148	6.2	3.4	5.4	2.3
149	5.9	3.0	5.1	1.8

```
y.head()
          0.0
     0
          0.0
     1
     2
          0.0
     3
          0.0
          0.0
     Name: target, dtype: float64
y.tail()
     145
            2.0
     146
            2.0
     147
            2.0
     148
            2.0
     149
            2.0
     Name: target, dtype: float64
x_train, x_test, y_train, y_test= train_test_split(x, y,shuffle= True, test_size= 0.3, rar
x_train= np.asarray(x_train)
y_train= np.asarray(y_train)
x_test= np.asarray(x_test)
y_test= np.asarray(y_test)
print(f'training set size: {x_train.shape[0]} samples \ntest set size: {x_test.shape[0]} s
     training set size: 105 samples
     test set size: 45 samples
# Normalizing data
scaler= Normalizer().fit(x_train) # the scaler is fitted to the training set
normalized_x_train= scaler.transform(x_train) # the scaler is applied to the training set
normalized_x_{test} = scaler.transform(x_{test}) # the scaler is applied to the test set
# calculating euclidean distance
def euc Dist(x train, x test data):
  dist= []
  for row in range(len(x train)):
    curr_train= x_train[row]
    curr_dist= 0
    for col in range(len(curr_train)):
      curr_dist += (curr_train[col] - x_test_data[col]) **2
    curr_dist= np.sqrt(curr_dist)
    dist.append(curr dist)
  distances= pd.DataFrame(data=dist,columns=['dist'])
  return distances
# finding nearest neighbours
```

def nearest\_neighbors(dist\_data, K):

```
neighbour df= dist data.sort values(by=['dist'], axis=0)
  neighbour df= neighbour df[:K]
  return neighbour_df
# votes function to calculate majority voting
def votes(neighbour_df, y_train):
 count= Counter(y_train[neighbour_df.index])
 y_pred= count.most_common()[0][0] # Majority Voting
 return y_pred
# KNN for detecting K nearest neighbours for the test sample
def KNN(x_train, y_train, x_test, K):
 y_predicted=[]
 for x_test_point in x_test:
      distance_point = euc_Dist(x_train, x_test_point)
     df_nearest_point= nearest_neighbors(distance_point, K)
     y_pred_point = votes(df_nearest_point, y_train)
     y_predicted.append(y_pred_point)
  return y_predicted
K=int(input("Enter k value: "))
y_pred_final= KNN(x_train, y_train, x_test, K)
print(y_pred_final)
     Enter k value: 4
     [1.0, 0.0, 2.0, 1.0, 1.0, 0.0, 1.0, 2.0, 1.0, 2.0, 0.0, 0.0, 0.0, 0.0, 0.0, 1.0, 2.0
print(len(y_pred_final))
     45
accuracy final = accuracy score(y test, y pred final)
print("The test accuracy of K-NN is: ", accuracy_final)
     The test accuracy of K-NN is: 1.0
# confusion matrix
cm = confusion_matrix(y_test, y_pred_final,normalize='true')
print("Confusion matrix:\n",cm)
     Confusion matrix:
      [[1.
                  0.
                              0.
      [0.
                  0.92307692 0.07692308]
      [0.
                  0.15384615 0.84615385]]
# Accuracies for varying k values
accuracies = []
iter = 50
```

```
no_neighbors = np.arange(1, iter)
for i, k in enumerate(no_neighbors):
    knn = KNN(x_train, y_train, x_test, k)
plt.title('k-NN: Number of Neighbors')
plt.plot(no_neighbors, accuracies, label = 'Testing Accuracy')
plt.legend()
plt.xlabel('k - Number of Neighbors')
plt.ylabel('Accuracy')
plt.show()
```



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