

9.Problem : Write a program to implement k-Nearest Neighbour algorithm to classify the iris data set. Print both correct and wrong predictions. Java/Python ML library classes can be used for this problem.

In [6]:

```
# Author : Dr.Thyagaraju G S , Context Innovations Lab , DEpt of CSE , SDMIT - Ujire
# Date : July 11 2018
# Python program to demonstrate
# KNN classification algorithm
# on IRIS dataset

from sklearn.datasets import load_iris
from sklearn.neighbors import KNeighborsClassifier
import numpy as np
from sklearn.model_selection import train_test_split
iris_dataset=load_iris()

print("\n IRIS FEATURES \ TARGET NAMES: \n ", iris_dataset.target_names)
for i in range(len(iris_dataset.target_names)):
    print("\n[{0}]:[{1}]".format(i,iris_dataset.target_names[i]))

print("\n IRIS DATA :\n",iris_dataset["data"])

X_train, X_test, y_train, y_test = train_test_split(iris_dataset["data"], iris_dataset["target"], r
andom_state=0)

print("\n Target :\n",iris_dataset["target"])
print("\n X TRAINING DATA SET \n", X_train)
print("\n Y TRAINING DATA SET \n", y_train)
print("\n X TESTING DATA SET \n", X_test)
print("\n Y TESTING DATA SET \n", y_test)

kn = KNeighborsClassifier(n_neighbors=1)
kn.fit(X_train, y_train)

print("Prediction Test / Validation \n ")
for i in range(len(X_test)):
    x = X_test[i]
    x_new = np.array([x])
    prediction = kn.predict(x_new)
    print("\n Actual : {0} {1}, Predicted :{2}{3}".format(y_test[i],iris_dataset["target_names"][y
test[i]],prediction,iris_dataset["target_names"][prediction]))

print("\n TEST SCORE[ACCURACY]: {:.2f}\n".format(kn.score(X_test, y_test)))
```

```
IRIS FEATURES \ TARGET NAMES:
['setosa' 'versicolor' 'virginica']
```

```
[0]:[setosa]
```

```
[1]:[versicolor]
```

```
[2]:[virginica]
```

```
IRIS DATA :
[[ 5.1  3.5  1.4  0.2]
 [ 4.9  3.   1.4  0.2]
 [ 4.7  3.2  1.3  0.2]
 [ 4.6  3.1  1.5  0.2]
 [ 5.   3.6  1.4  0.2]
 [ 5.4  3.9  1.7  0.4]
 [ 4.6  3.4  1.4  0.3]
 [ 5.   3.4  1.5  0.2]
 [ 4.4  2.9  1.4  0.2]
 [ 4.9  3.1  1.5  0.1]
 [ 5.4  3.7  1.5  0.2]
 [ 4.8  3.4  1.6  0.2]
 [ 4.8  3.   1.4  0.1]]
```

[4.3 3. 1.1 0.1]
[5.8 4. 1.2 0.2]
[5.7 4.4 1.5 0.4]
[5.4 3.9 1.3 0.4]
[5.1 3.5 1.4 0.3]
[5.7 3.8 1.7 0.3]
[5.1 3.8 1.5 0.3]
[5.4 3.4 1.7 0.2]
[5.1 3.7 1.5 0.4]
[4.6 3.6 1. 0.2]
[5.1 3.3 1.7 0.5]
[4.8 3.4 1.9 0.2]
[5. 3. 1.6 0.2]
[5. 3.4 1.6 0.4]
[5.2 3.5 1.5 0.2]
[5.2 3.4 1.4 0.2]
[4.7 3.2 1.6 0.2]
[4.8 3.1 1.6 0.2]
[5.4 3.4 1.5 0.4]
[5.2 4.1 1.5 0.1]
[5.5 4.2 1.4 0.2]
[4.9 3.1 1.5 0.1]
[5. 3.2 1.2 0.2]
[5.5 3.5 1.3 0.2]
[4.9 3.1 1.5 0.1]
[4.4 3. 1.3 0.2]
[5.1 3.4 1.5 0.2]
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[4.4 3.2 1.3 0.2]
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[7. 3.2 4.7 1.4]
[6.4 3.2 4.5 1.5]
[6.9 3.1 4.9 1.5]
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[6.5 2.8 4.6 1.5]
[5.7 2.8 4.5 1.3]
[6.3 3.3 4.7 1.6]
[4.9 2.4 3.3 1.]
[6.6 2.9 4.6 1.3]
[5.2 2.7 3.9 1.4]
[5. 2. 3.5 1.]
[5.9 3. 4.2 1.5]
[6. 2.2 4. 1.]
[6.1 2.9 4.7 1.4]
[5.6 2.9 3.6 1.3]
[6.7 3.1 4.4 1.4]
[5.6 3. 4.5 1.5]
[5.8 2.7 4.1 1.]
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[5.6 2.5 3.9 1.1]
[5.9 3.2 4.8 1.8]
[6.1 2.8 4. 1.3]
[6.3 2.5 4.9 1.5]
[6.1 2.8 4.7 1.2]
[6.4 2.9 4.3 1.3]
[6.6 3. 4.4 1.4]
[6.8 2.8 4.8 1.4]
[6.7 3. 5. 1.7]
[6. 2.9 4.5 1.5]
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[5.5 2.4 3.8 1.1]
[5.5 2.4 3.7 1.]
[5.8 2.7 3.9 1.2]
[6. 2.7 5.1 1.6]
[5.4 3. 4.5 1.5]
[6. 3.4 4.5 1.6]
[6.7 3.1 4.7 1.5]
[6.3 2.3 4.4 1.3]
[5.6 3. 4.1 1.3]
[5.5 2.5 4. 1.3]

[6.5 3. 5.5 1.8]
[6.7 3.3 5.7 2.5]
[6. 2.2 5. 1.5]
[6.7 2.5 5.8 1.8]
[5.6 2.5 3.9 1.1]
[7.7 3. 6.1 2.3]
[6.3 3.3 4.7 1.6]
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[6.3 2.7 4.9 1.8]
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[4.9 2.5 4.5 1.7]
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[7. 3.2 4.7 1.4]
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[5.8 2.7 5.1 1.9]
[5.6 2.7 4.2 1.3]
[5.6 2.9 3.6 1.3]
[5.5 2.5 4. 1.3]
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[7.2 3.2 6. 1.8]
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[6.4 2.7 5.3 1.9]
[5.7 3. 4.2 1.2]
[5.4 3.4 1.7 0.2]
[5.7 4.4 1.5 0.4]
[6.9 3.1 4.9 1.5]
[4.6 3.1 1.5 0.2]
[5.9 3. 5.1 1.8]
[5.1 2.5 3. 1.1]
[4.6 3.4 1.4 0.3]
[6.2 2.2 4.5 1.5]
[7.2 3.6 6.1 2.5]
[5.7 2.9 4.2 1.3]
[4.8 3. 1.4 0.1]
[7.1 3. 5.9 2.1]
[6.9 3.2 5.7 2.3]
[6.5 3. 5.8 2.2]
[6.4 2.8 5.6 2.1]
[5.1 3.8 1.6 0.2]
[4.8 3.4 1.6 0.2]
[6.5 3.2 5.1 2.]
[6.7 3.3 5.7 2.1]
[4.5 2.3 1.3 0.3]
[6.2 3.4 5.4 2.3]
[4.9 3. 1.4 0.2]
[5.7 2.5 5. 2.]
[6.9 3.1 5.4 2.1]
[4.4 3.2 1.3 0.2]
[5. 3.6 1.4 0.2]
[7.2 3. 5.8 1.6]
[5.1 3.5 1.4 0.3]
[4.4 3. 1.3 0.2]
[5.4 3.9 1.7 0.4]
[5.5 2.3 4. 1.3]
[6.8 3.2 5.9 2.3]
[7.6 3. 6.6 2.1]
[5.1 3.5 1.4 0.2]
[4.9 3.1 1.5 0.1]
[5.2 3.4 1.4 0.2]
[5.7 2.8 4.5 1.3]
[6.6 3. 4.4 1.4]
[5. 3.2 1.2 0.2]
[5.1 3.3 1.7 0.5]
[6.4 2.9 4.3 1.3]
[5.4 3.4 1.5 0.4]
[7.7 2.6 6.9 2.3]
[4.9 2.4 3.3 1.]
[7.9 3.8 6.4 2.]
[6.7 3.1 4.4 1.4]
[5.2 4.1 1.5 0.1]
[6. 3. 4.8 1.8]
[5.8 4. 1.2 0.2]
[7.7 2.8 6.7 2.]
[5.1 3.8 1.5 0.3]

```
[ 4.7  3.2  1.6  0.2]
[ 7.4  2.8  6.1  1.9]
[ 5.   3.3  1.4  0.2]
[ 6.3  3.4  5.6  2.4]
[ 5.7  2.8  4.1  1.3]
[ 5.8  2.7  3.9  1.2]
[ 5.7  2.6  3.5  1. ]
[ 6.4  3.2  5.3  2.3]
[ 6.7  3.   5.2  2.3]
[ 6.3  2.5  4.9  1.5]
[ 6.7  3.   5.   1.7]
[ 5.   3.   1.6  0.2]
[ 5.5  2.4  3.7  1. ]
[ 6.7  3.1  5.6  2.4]
[ 5.8  2.7  5.1  1.9]
[ 5.1  3.4  1.5  0.2]
[ 6.6  2.9  4.6  1.3]
[ 5.6  3.   4.1  1.3]
[ 5.9  3.2  4.8  1.8]
[ 6.3  2.3  4.4  1.3]
[ 5.5  3.5  1.3  0.2]
[ 5.1  3.7  1.5  0.4]
[ 4.9  3.1  1.5  0.1]
[ 6.3  2.9  5.6  1.8]
[ 5.8  2.7  4.1  1. ]
[ 7.7  3.8  6.7  2.2]
[ 4.6  3.2  1.4  0.2]]
```

Y TRAINING DATA SET

```
[1 1 2 0 2 0 0 1 2 2 2 2 1 2 1 1 2 2 2 2 1 2 1 0 2 1 1 1 1 2 0 0 2 1 0 0 1
0 2 1 0 1 2 1 0 2 2 2 2 0 0 2 2 0 2 0 2 2 0 0 2 0 0 0 1 2 2 0 0 0 1 1 0 0
1 0 2 1 2 1 0 2 0 2 0 0 2 0 2 1 1 1 2 2 1 1 0 1 2 2 0 1 1 1 1 0 0 0 2 1 2
0]
```

X TESTING DATA SET

```
[[ 5.8  2.8  5.1  2.4]
[ 6.   2.2  4.   1. ]
[ 5.5  4.2  1.4  0.2]
[ 7.3  2.9  6.3  1.8]
[ 5.   3.4  1.5  0.2]
[ 6.3  3.3  6.   2.5]
[ 5.   3.5  1.3  0.3]
[ 6.7  3.1  4.7  1.5]
[ 6.8  2.8  4.8  1.4]
[ 6.1  2.8  4.   1.3]
[ 6.1  2.6  5.6  1.4]
[ 6.4  3.2  4.5  1.5]
[ 6.1  2.8  4.7  1.2]
[ 6.5  2.8  4.6  1.5]
[ 6.1  2.9  4.7  1.4]
[ 4.9  3.1  1.5  0.1]
[ 6.   2.9  4.5  1.5]
[ 5.5  2.6  4.4  1.2]
[ 4.8  3.   1.4  0.3]
[ 5.4  3.9  1.3  0.4]
[ 5.6  2.8  4.9  2. ]
[ 5.6  3.   4.5  1.5]
[ 4.8  3.4  1.9  0.2]
[ 4.4  2.9  1.4  0.2]
[ 6.2  2.8  4.8  1.8]
[ 4.6  3.6  1.   0.2]
[ 5.1  3.8  1.9  0.4]
[ 6.2  2.9  4.3  1.3]
[ 5.   2.3  3.3  1. ]
[ 5.   3.4  1.6  0.4]
[ 6.4  3.1  5.5  1.8]
[ 5.4  3.   4.5  1.5]
[ 5.2  3.5  1.5  0.2]
[ 6.1  3.   4.9  1.8]
[ 6.4  2.8  5.6  2.2]
[ 5.2  2.7  3.9  1.4]
[ 5.7  3.8  1.7  0.3]
[ 6.   2.7  5.1  1.6]]
```

Y TESTING DATA SET

```
[2 1 0 2 0 2 0 1 1 1 2 1 1 1 1 0 1 1 0 0 2 1 0 0 2 0 0 1 1 0 2 1 0 2 2 1 0
1]
```

Prediction Test / Validation

Actual : 2 virginica, Predicted : [2] ['virginica']
Actual : 1 versicolor, Predicted : [1] ['versicolor']
Actual : 0 setosa, Predicted : [0] ['setosa']
Actual : 2 virginica, Predicted : [2] ['virginica']
Actual : 0 setosa, Predicted : [0] ['setosa']
Actual : 2 virginica, Predicted : [2] ['virginica']
Actual : 0 setosa, Predicted : [0] ['setosa']
Actual : 1 versicolor, Predicted : [1] ['versicolor']
Actual : 1 versicolor, Predicted : [1] ['versicolor']
Actual : 1 versicolor, Predicted : [1] ['versicolor']
Actual : 2 virginica, Predicted : [2] ['virginica']
Actual : 1 versicolor, Predicted : [1] ['versicolor']
Actual : 1 versicolor, Predicted : [1] ['versicolor']
Actual : 1 versicolor, Predicted : [1] ['versicolor']
Actual : 1 versicolor, Predicted : [1] ['versicolor']
Actual : 0 setosa, Predicted : [0] ['setosa']
Actual : 1 versicolor, Predicted : [1] ['versicolor']
Actual : 1 versicolor, Predicted : [1] ['versicolor']
Actual : 0 setosa, Predicted : [0] ['setosa']
Actual : 0 setosa, Predicted : [0] ['setosa']
Actual : 2 virginica, Predicted : [2] ['virginica']
Actual : 1 versicolor, Predicted : [1] ['versicolor']
Actual : 0 setosa, Predicted : [0] ['setosa']
Actual : 0 setosa, Predicted : [0] ['setosa']
Actual : 2 virginica, Predicted : [2] ['virginica']
Actual : 0 setosa, Predicted : [0] ['setosa']
Actual : 0 setosa, Predicted : [0] ['setosa']
Actual : 1 versicolor, Predicted : [1] ['versicolor']
Actual : 1 versicolor, Predicted : [1] ['versicolor']
Actual : 0 setosa, Predicted : [0] ['setosa']
Actual : 2 virginica, Predicted : [2] ['virginica']
Actual : 1 versicolor, Predicted : [1] ['versicolor']
Actual : 0 setosa, Predicted : [0] ['setosa']
Actual : 2 virginica, Predicted : [2] ['virginica']
Actual : 2 virginica, Predicted : [2] ['virginica']
Actual : 1 versicolor, Predicted : [1] ['versicolor']
Actual : 0 setosa, Predicted : [0] ['setosa']

Actual : 1 versicolor, Predicted :[2]['virginica']

TEST SCORE[ACCURACY]: 0.97