```
In [3]:
          1 from sklearn.model selection import train test split
          2 from sklearn.neighbors import KNeighborsClassifier
          3 from sklearn import datasets
          4 iris=datasets.load iris()
          5 x train,x test,y train,y test=train test split(iris.data,iris.target,test size=0.4)
          7 print("dataset is split into training and testing...")
          8 print("size of training data and its label", x train.shape, y train.shape)
          9 print("size of training data and its label", x test.shape, y test.shape)
         10 for i in range(len(iris.target names)):
                 print("label",i,"-",str(iris.target names[i]))
         12 classifier=KNeighborsClassifier(n neighbors=3)
         13 classifier.fit(x train,y train)
         14 v pred=classifier.predict(x test)
         15 print("results of classification using k-nn with k=3")
         16 for r in range(0,len(x test)):
                 print("sample:",str(x test[r]),"Actual-label:",str(y test[r]),"predicted-label:",str(y pred[r]))
         17
         18 print("classification accuracy:",classifier.score(x test,y test)*100);
        dataset is split into training and testing...
        size of training data and its label (90, 4) (90,)
        size of training data and its label (60, 4) (60,)
        label 0 - setosa
        label 1 - versicolor
        label 2 - virginica
        results of classification using k-nn with k=3
        sample: [5.8 4. 1.2 0.2] Actual-label: 0 predicted-label: 0
        sample: [6.5 2.8 4.6 1.5] Actual-label: 1 predicted-label: 1
        sample: [6.2 2.9 4.3 1.3] Actual-label: 1 predicted-label: 1
        sample: [4.6 3.6 1. 0.2] Actual-label: 0 predicted-label: 0
        sample: [5.5 3.5 1.3 0.2] Actual-label: 0 predicted-label: 0
        sample: [7.6 3. 6.6 2.1] Actual-label: 2 predicted-label: 2
        sample: [5. 3.2 1.2 0.2] Actual-label: 0 predicted-label: 0
        sample: [5. 3.4 1.6 0.4] Actual-label: 0 predicted-label: 0
        sample: [5.7 2.9 4.2 1.3] Actual-label: 1 predicted-label: 1
        sample: [5.5 2.5 4. 1.3] Actual-label: 1 predicted-label: 1
        sample: [4.6 3.2 1.4 0.2] Actual-label: 0 predicted-label: 0
        sample: [5. 3.6 1.4 0.2] Actual-label: 0 predicted-label: 0
        sample: [4.9 2.4 3.3 1. ] Actual-label: 1 predicted-label: 1
        sample: [6.8 3. 5.5 2.1] Actual-label: 2 predicted-label: 2
        sample: [6. 3.4 4.5 1.6] Actual-label: 1 predicted-label: 1
```

```
sample: [5.1 2.5 3. 1.1] Actual-label: 1 predicted-label: 1
sample: [5.7 2.5 5. 2.] Actual-label: 2 predicted-label: 2
sample: [5.4 3.7 1.5 0.2] Actual-label: 0 predicted-label: 0
sample: [4.4 3. 1.3 0.2] Actual-label: 0 predicted-label: 0
sample: [6. 3. 4.8 1.8] Actual-label: 2 predicted-label: 2
sample: [6.9 3.1 5.4 2.1] Actual-label: 2 predicted-label: 2
sample: [5.7 4.4 1.5 0.4] Actual-label: 0 predicted-label: 0
sample: [4.9 3.1 1.5 0.1] Actual-label: 0 predicted-label: 0
sample: [7.2 3.6 6.1 2.5] Actual-label: 2 predicted-label: 2
sample: [5. 3. 1.6 0.2] Actual-label: 0 predicted-label: 0
sample: [6. 2.7 5.1 1.6] Actual-label: 1 predicted-label: 2
sample: [6.4 2.8 5.6 2.2] Actual-label: 2 predicted-label: 2
sample: [6.3 2.5 4.9 1.5] Actual-label: 1 predicted-label: 2
sample: [5.1 3.8 1.5 0.3] Actual-label: 0 predicted-label: 0
sample: [5.5 2.3 4. 1.3] Actual-label: 1 predicted-label: 1
sample: [7.9 3.8 6.4 2. ] Actual-label: 2 predicted-label: 2
sample: [6.8 3.2 5.9 2.3] Actual-label: 2 predicted-label: 2
sample: [5.5 2.4 3.7 1. ] Actual-label: 1 predicted-label: 1
sample: [6.4 3.2 4.5 1.5] Actual-label: 1 predicted-label: 1
sample: [5.2 3.4 1.4 0.2] Actual-label: 0 predicted-label: 0
sample: [5.8 2.8 5.1 2.4] Actual-label: 2 predicted-label: 2
sample: [5.5 2.6 4.4 1.2] Actual-label: 1 predicted-label: 1
sample: [4.8 3. 1.4 0.3] Actual-label: 0 predicted-label: 0
sample: [4.6 3.4 1.4 0.3] Actual-label: 0 predicted-label: 0
sample: [5.8 2.7 5.1 1.9] Actual-label: 2 predicted-label: 2
sample: [5.7 2.8 4.5 1.3] Actual-label: 1 predicted-label: 1
sample: [6.7 3.3 5.7 2.5] Actual-label: 2 predicted-label: 2
sample: [5. 3.5 1.6 0.6] Actual-label: 0 predicted-label: 0
sample: [5.4 3.9 1.3 0.4] Actual-label: 0 predicted-label: 0
sample: [7.3 2.9 6.3 1.8] Actual-label: 2 predicted-label: 2
sample: [5. 3.4 1.5 0.2] Actual-label: 0 predicted-label: 0
sample: [5.2 2.7 3.9 1.4] Actual-label: 1 predicted-label: 1
sample: [5.1 3.4 1.5 0.2] Actual-label: 0 predicted-label: 0
sample: [4.4 2.9 1.4 0.2] Actual-label: 0 predicted-label: 0
sample: [5.1 3.3 1.7 0.5] Actual-label: 0 predicted-label: 0
sample: [6.2 3.4 5.4 2.3] Actual-label: 2 predicted-label: 2
sample: [5.1 3.8 1.6 0.2] Actual-label: 0 predicted-label: 0
sample: [4.8 3. 1.4 0.1] Actual-label: 0 predicted-label: 0
sample: [5.4 3.9 1.7 0.4] Actual-label: 0 predicted-label: 0
sample: [5.3 3.7 1.5 0.2] Actual-label: 0 predicted-label: 0
sample: [7.7 2.6 6.9 2.3] Actual-label: 2 predicted-label: 2
sample: [5.1 3.8 1.9 0.4] Actual-label: 0 predicted-label: 0
```

sample: [6.7 3.1 4.4 1.4] Actual-label: 1 predicted-label: 1
sample: [4.3 3. 1.1 0.1] Actual-label: 0 predicted-label: 0
sample: [6.3 2.9 5.6 1.8] Actual-label: 2 predicted-label: 2
classification accuracy: 96.6666666666667

localhost:8888/notebooks/Documents/4GW19IS041/Program 8(K-Nearest).ipynb