```
import pandas as pd
      import matplotlib.pyplot as plt
      import seaborn as sns
      from sklearn.model selection import train test split
      from sklearn import metrics
      from sklearn.tree import DecisionTreeClassifier
      from sklearn.naive bayes import GaussianNB
      from sklearn.svm import SVC
      from sklearn.neighbors import KNeighborsClassifier
      from sklearn.model_selection import KFold, LeaveOneOut, ShuffleSplit, StratifiedKFold
      from sklearn.model_selection import cross_val_score
      from sklearn.metrics import accuracy_score, confusion_matrix
In [2]: iris = pd.read_csv('D:\\24 - Machine_Learning\\download files\\iris.csv')
      iris
Out[2]:
                                Sepal
                                                          Petal
             Sepal length
                                       Petal length
                                                                      Class
                                width
                                                          width
          0
                      5 1
                                  3.5
                                                14
                                                           0.2
                                                                  Iris-setosa
                      4.9
                                                            0.2
          1
                                  3.0
                                                1.4
                                                                  Iris-setosa
          2
                      4.7
                                  3.2
                                                1.3
                                                            0.2
                                                                  Iris-setosa
                      4.6
                                  3.1
                                                1.5
                                                            0.2
                                                                  Iris-setosa
          3
                      5.0
                                  3.6
                                                1.4
                                                            0.2
                                                                  Iris-setosa
        145
                      6.7
                                  3.0
                                               5.2
                                                            2.3
                                                                Iris-virginica
        146
                      6.3
                                  25
                                               5.0
                                                            1.9
                                                                Iris-virginica
        147
                      6.5
                                  3.0
                                               5.2
                                                            2.0
                                                                 Iris-virginica
        148
                      6.2
                                  3.4
                                               5.4
                                                            2.3
                                                                Iris-virginica
                      5.9
                                  3.0
                                               5.1
                                                               Iris-virginica
       149
                                                            1.8
       150 rows × 5 columns
In [3]: iris.dtypes
Out[3]:Sepal length
                     float64
       Sepal width
                     float64
       Petal length
                     float64
       Petal width
                     float64
       Class
                    object
       dtype: object
In [4]: features = iris[['Sepal length', 'Sepal width', 'Petal length', 'Petal width']].values
      classes = iris['Class'].values
In [5]: # k-fold cross validation technique
      kf = KFold(n_splits=5, random_state=1,shuffle=True)
      # stratified kfold cross validation technique
      skf = StratifiedKFold(n splits=5)
      # LeaveOneOut cross validation technique
      loocv = LeaveOneOut()
      # shuffle split cross validation technique
      shvc = ShuffleSplit()
In [6]: # evaluating the data sets with kfold and DecisionTreeClassifier
      dst = DecisionTreeClassifier()
      scores = cross val score(dst.features.classes.scoring='accuracy'.cv=kf)
      print('Accuracy using Decision Tree: %2f%%'%(scores.mean()*100))
      print(scores)
Accuracy using Decision Tree: 94.000000%
[0.96666667 0.96666667 0.96666667 0.93333333 0.86666667]
In [7]: # evaluating the data sets with kfold and Naive Bayes Classifier
      nb = GaussianNB()
      scores = cross_val_score(nb,features,classes,scoring='accuracy',cv=kf)
      print('Accuracy using GaussianNB: %2f%%'%(scores.mean()*100))
      print(scores)
Accuracy using GaussianNB: 95.333333%
[0.96666667 0.96666667 1.
                                 0.9
                                         0.93333333]
In [8]: # evaluating the data sets with kfold and SVM
      scores = cross_val_score(svm,features,classes,scoring='accuracy',cv=kf)
      print('Accuracy using SVC: %2f%%'%(scores.mean()*100))
      print(scores)
Accuracy using SVC: 94.666667%
[0.96666667 0.93333333 0.9
                                  0.96666667 0.96666667]
In [9]: # evaluating the data sets with kfold and KNN Classifier
```

In [1]: import numpy as np

```
scores = cross_val_score(knn,features,classes,scoring='accuracy',cv=kf)
    print('Accuracy using KNN: %2f%%'%(scores.mean()*100))
    print(scores)
Accuracy using KNN: 95.333333%
     0.96666667 0.9
                    0.93333333 0.96666667]
In [10]: # evaluating the data sets with Stratified KFold and DecisionTree
     scores = cross val score(dst,features,classes,scoring='accuracy',cv=skf)
     print('Accuracy using Decision Tree: %2f%%'%(scores.mean()*100))
     print(scores)
Accuracy using Decision Tree: 96.666667%
[0.96666667 0.96666667 0.9
                        1.
In [11]: # evaluating the data sets Stratified KFold and Naive Bayes Classifier
     scores = cross_val_score(nb,features,classes,scoring='accuracy',cv=skf)
     print('Accuracy using GaussianNB: %2f%%'%(scores.mean()*100))
     print(scores)
Accuracy using GaussianNB: 95.333333%
[0.93333333 0.96666667 0.93333333 0.93333333 1.
In [12]: # evaluating the data sets Stratified KFold and SVM
     scores = cross_val_score(svm,features,classes,scoring='accuracy',cv=skf)
     print('Accuracy using SVM: %2f%%'%(scores.mean()*100))
     print(scores)
Accuracy using SVM: 96.666667%
[0.96666667 0.96666667 0.96666667 0.93333333 1.
In [13]: # evaluating the data sets Stratified KFold and KNN
     scores = cross val score(knn,features,classes,scoring='accuracy',cv=skf)
     print('Accuracy using KNN: %2f%%'%(scores.mean()*100))
     print(scores)
Accuracy using KNN: 97.333333%
              0.93333333 0.96666667 1.
[0.96666667 1.
In [14]: # evaluating the data sets with LeaveOneOut and DecisionTree
     scores = cross_val_score(dst,features,classes,scoring='accuracy',cv=loocv)
     print('Accuracy using GaussianNB: %2f%%'%(scores.mean()*100))
     print(scores)
Accuracy using GaussianNB: 94.000000%
1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 0. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 0.
1. 1. 1. 1. 1. 1. 1. 1. 1. 0. 1. 1. 0. 1. 1. 1. 0. 1. 1. 1. 1. 1. 1. 1. 1.
1. 1. 1. 1. 1. 1.]
In [15]: # evaluating the data sets LeaveOneOut and Naive Bayes Classifier
     scores = cross_val_score(nb,features,classes,scoring='accuracy',cv=loocv)
     print('Accuracy using GaussianNB: %2f%%''%(scores.mean()*100))
     print(scores)
Accuracy using GaussianNB: 95.333333%
1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 0. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 0.
1. 1. 1. 1. 1. 1.]
In [16]: # evaluating the data sets LeaveOneOut and SVM
     scores = cross_val_score(svm,features,classes,scoring='accuracy',cv=loocv)
     print('Accuracy using SVM: %2f%%'%(scores.mean()*100))
     print(scores)
Accuracy using SVM: 96.666667%
1. 1. 1. 1. 1. 1.]
In [17]: # evaluating the data sets LeaveOneOut and KNN
     scores = cross_val_score(knn,features,classes,scoring='accuracy',cv=loocv)
     print('Accuracy using KNN: %2f%%'%(scores.mean()*100))
     print(scores)
```

knn = KNeighborsClassifier()

```
Accuracy using KNN: 96.666667%
1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 0. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 0.
1. 1. 1. 1. 1. 1.]
In [18]: # evaluating the data sets with ShuffleSplit and DecisionTree
      scores = cross val score(dst,features,classes,scoring='accuracy',cv=shvc)
      print('Accuracy using GaussianNB: %2f%%'%(scores.mean()*100))
      print(scores)
Accuracy using GaussianNB: 94.666667%
     1.
           0.93333333 0.86666667 1.
                                      0.86666667
0.86666667 1.
               1.
                      0.93333333]
In [19]: # evaluating the data sets with ShuffleSplit and Naive Bayes Classifier
      scores = cross_val_score(nb,features,classes,scoring='accuracy',cv=shvc)
      print('Accuracy using GaussianNB: %2f%%'%(scores.mean()*100))
      print(scores)
Accuracy using GaussianNB: 95.333333%
      0.93333333 1. 0.8
                            0.866666671.
0.93333333 1.
                      1.
              1.
                           ]
In [20]: # evaluating the data sets with ShuffleSplit and SVM
      scores = cross_val_score(svm,features,classes,scoring='accuracy',cv=shvc)
      print('Accuracy using GaussianNB: %2f%%'%(scores.mean()*100))
      print(scores)
Accuracy using GaussianNB: 98.000000%
           0.93333333 0.93333333 1.
     1.
                                     1.
[1.
      1.
                 0.93333333]
In [21]: # evaluating the data sets with ShuffleSplit and KNN
     scores = cross_val_score(knn,features,classes,scoring='accuracy',cv=shvc)
      print('Accuracy using KNN: %2f%%'%(scores_mean()*100))
      print(scores)
Accuracy using KNN: 96.000000%
      0.86666667 1.
                                 0.86666667
[1.
                     1.
```

0.93333333]

1.

0.93333333 1.