3. Use sklearn. datasets import load_iris use k-neighbour classifier to classify the three flowers to setosa, vesicolor and Virginica.

import basic libraries

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

import the dataset and take independent and dependent variables separately

```
In [2]: from sklearn.datasets import load_iris
    iris = load_iris()

X = iris.data[:,:4]  #independent variables
Y = iris.target  #dependent variable
```

split the dataset into train set and test set

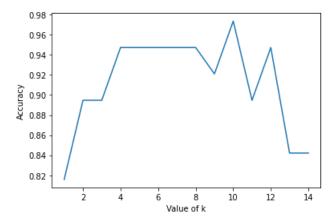
```
In [3]: from sklearn.model_selection import train_test_split
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.25, random_state=0)
```

scale the features

```
In [4]:
    from sklearn.preprocessing import StandardScaler
    scaler = StandardScaler()
    X_train = scaler.fit_transform(X_train)
    X_test = scaler.fit_transform(X_test)
    #scaler.fit(X_train)
    #X_train = scaler.transform(X_train)
    #X_test = scaler.transform(X_test)
```

train the model and fit the training set and compare with test set

```
In [5]: from sklearn.neighbors import KNeighborsClassifier
        from sklearn import metrics
        range_k = range(1,15)
                                      #for checking no. of neighbors(k) from 1 to 14
        scores_list = []
                                      #take an empty list for storing the accuracy for each classifer of k=1 to 14
        for k in range_k:
            classifier = KNeighborsClassifier(n_neighbors=k)
                                                                          #parameter : n_neighbors = no of neighbors
            classifier.fit(X_train,Y_train)
                                                                          #fit the training set
            Y_pred = classifier.predict(X_test)
                                                                          #predict the values for test set
            scores_list.append(metrics.accuracy_score(Y_test,Y_pred))
                                                                         #append the accuracy value in the list
            print("for neighbor = ",k,"accuracy : ",metrics.accuracy_score(Y_test,Y_pred))
        #plot the k value vs accuracy
        plt.plot(range k,scores list)
        plt.xlabel("Value of k")
        plt.ylabel("Accuracy")
        plt.show()
        for neighbor = 1 accuracy : 0.8157894736842105
        for neighbor = 2 accuracy : 0.8947368421052632
        for neighbor = 3 accuracy : 0.8947368421052632
        for neighbor = 4 accuracy :
                                     0.9473684210526315
        for neighbor = 5 accuracy : 0.9473684210526315
        for neighbor = 6 accuracy : 0.9473684210526315
        for neighbor = 7 accuracy : 0.9473684210526315
        for neighbor = 8 accuracy : 0.9473684210526315
        for neighbor = 9 accuracy: 0.9210526315789473
        for neighbor = 10 accuracy : 0.9736842105263158
        for neighbor = 11 \ \text{accuracy} : 0.8947368421052632
        for neighbor = 12 accuracy : 0.9473684210526315
        for neighbor = 13 accuracy : 0.8421052631578947
        for neighbor = 14 accuracy : 0.8421052631578947
```



As for neighbor = 10, the accuracy is maximum so we make our final model with no. of neighbors = 10

final model

```
In [6]: classifier = KNeighborsClassifier(n_neighbors=10)
    classifier.fit(X_train,Y_train)
    Y_pred = classifier.predict(X_test)
    classes = {0:'setosa', 1:'versicolor', 2:'virginicia'}
```

find classification report and confusion matrix

```
In [7]: result = metrics.confusion_matrix(Y_test,Y_pred)
         print("Confusion Matrix:\n",result)
         result1 = metrics.classification_report(Y_test,Y_pred)
         print("Classification Report:\n",result1)
        Confusion Matrix:
         [[13 0 0]
[ 0 15 1]
         [0 0 9]]
        Classification Report:
                        precision
                                     recall f1-score
                                                         support
                    0
                            1.00
                                      1.00
                                                 1.00
                                                             13
                    1
                            1.00
                                      0.94
                                                 0.97
                                                             16
                    2
                            0.90
                                      1.00
                                                 0.95
                                                              9
            accuracy
                                                 0.97
                                                             38
           macro avg
                            0.97
                                      0.98
                                                 0.97
                                                             38
        weighted avg
                            0.98
                                      0.97
                                                 0.97
                                                             38
```

predict the result

```
In [8]: x_new = [[1,1,1,1],[4,3,1.3,0.2]] #predict the results for two randomly taken values
y_predict = classifier.predict(x_new)
print(classes[y_predict[0]])
print(classes[y_predict[1]])

virginicia
virginicia
In []:
```