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

```
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.neighbors import KNeighborsClassifier
from sklearn import metrics
import matplotlib.pyplot as plt
iris = load_iris()
print(iris.target_names)
print(iris.data.shape)
x = iris.data[:, :4]
y = iris.target
X_train,X_test,Y_train,Y_test = train_test_split(x,y,test_size=0.3)
scaler = StandardScaler()
scaler.fit(X train)
X_train = scaler.transform(X_train)
X_test = scaler.transform(X_test)
print(X_train.shape)
print(X_test.shape)
range_k = range(1,15)
scores = {}
scores_list = []
for k in range_k:
   classifier = KNeighborsClassifier(n_neighbors=k)
    {\tt classifier.fit(X\_train,Y\_train)}
    Y_pred = classifier.predict(X_test)
    scores[k] = metrics.accuracy_score(Y_test,Y_pred)
    scores_list.append(metrics.accuracy_score(Y_test,Y_pred))
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)
plt.plot(range_k,scores_list)
plt.xlabel("Value of k")
plt.ylabel("Accuracy")
classifier = KNeighborsClassifier(n_neighbors=8)
classifier.fit(X_train,Y_train)
classes = {0:'setosa', 1:'versicolor', 2:'virginicia'}
x_{new} = [[1,1,1,1],[4,3,1.3,0.2]]
y_predict = classifier.predict(x_new)
print(classes[y_predict[0]])
print(classes[y_predict[1]])
['setosa' 'versicolor' 'virginica']
(150, 4)
(105, 4)
(45, 4)
Confusion Matrix:
[[14 0 0]
[ 0 13 1]
  0 3 14]]
Classification Report:
                            recall f1-score support
               precision
                  1.00
           0
                             1.00
                                       1.00
                                                   14
           1
                   0.81
                             0.93
                                       0.87
                                                   14
           2
                  0.93
                             0.82
                                       0.87
                                                   17
   accuracy
                                       0.91
                                                   45
   macro avg
                   0.92
                             0.92
                                       0.91
                                                   45
                   0.92
                             0.91
weighted avg
                                       0.91
virginicia
virginicia
  0.95
  0.94
 0.93
  0.92
```

0.91

4

10

Value of k

12

14