NAME – ANMOL
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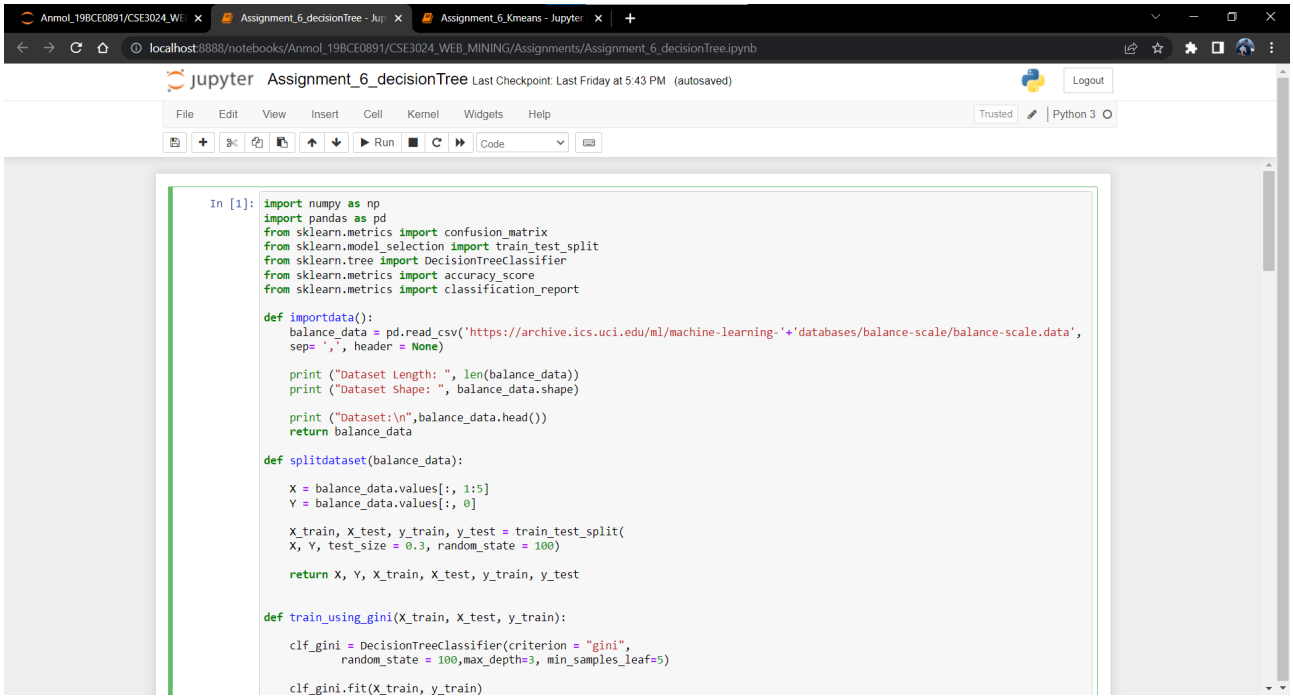
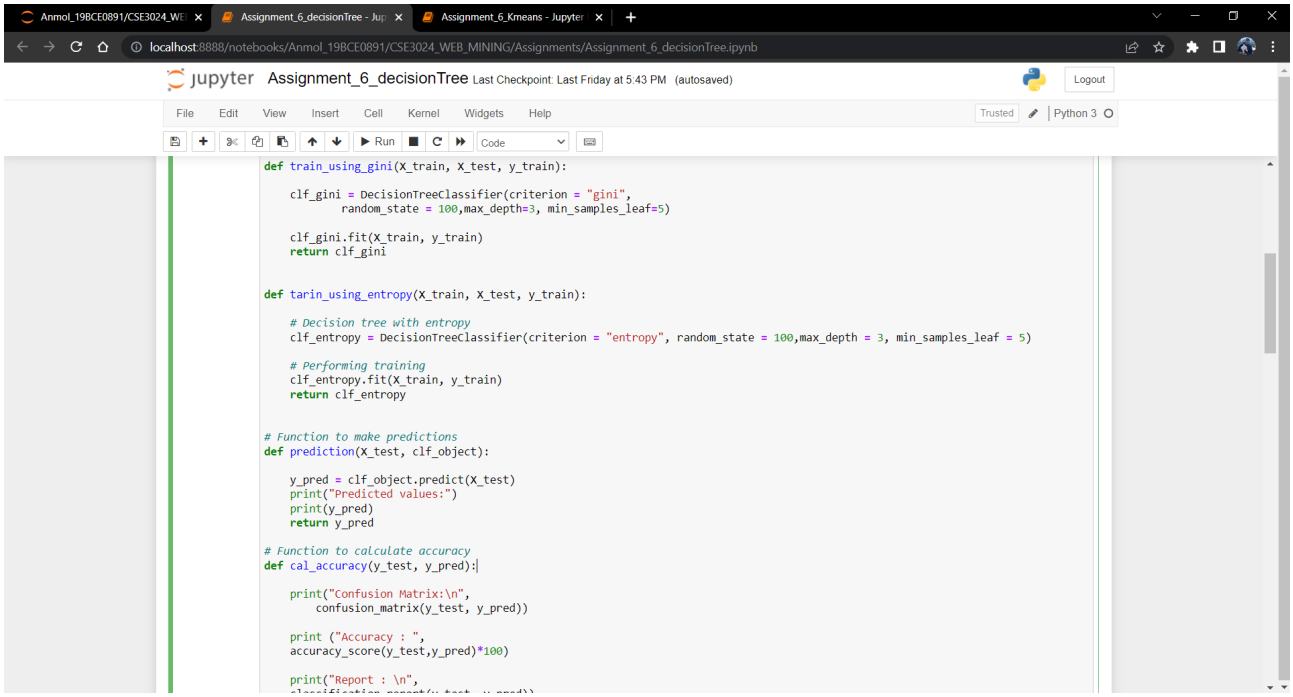
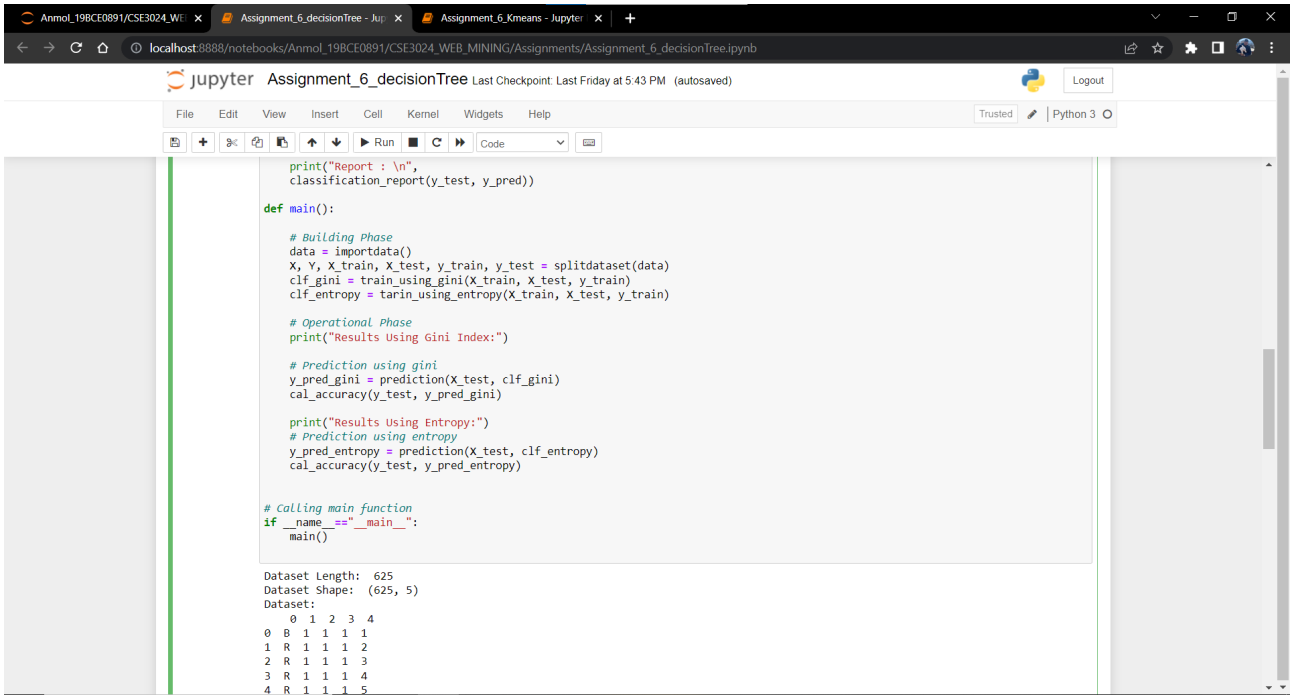
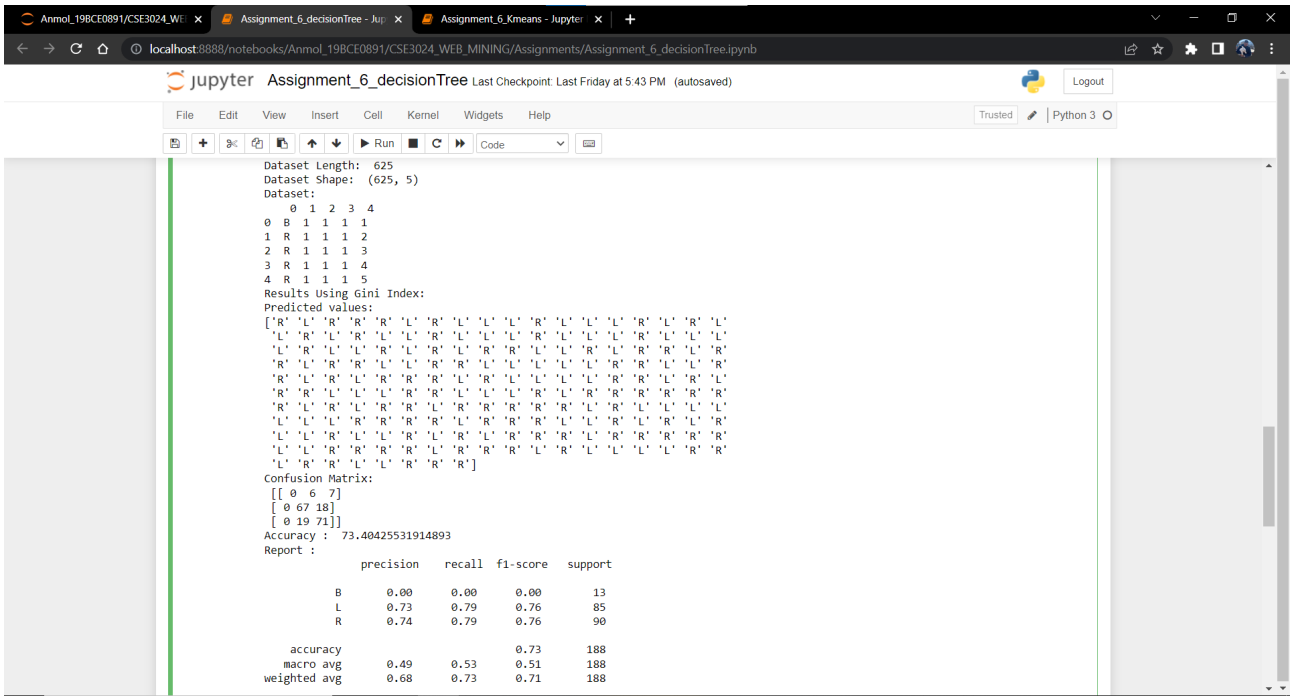
DIGITAL ASSIGNMENT – 6

1. Create a Python programme to implement the decision tree and prints the accuracy percentage, pression, recall and the predicted values.






Anmol_19BCE0891/CSE3024_WE
Assignment_6_decisionTree - Jup
Assignment_6_Kmeans - Jupyter
localhost:8888/notebooks/Anmol_19BCE0891/CSE3024_WEB_MINING/Assignments/Assignment_6_decisionTree.ipynb
jupyter Assignment_6_decisionTree Last Checkpoint: Last Friday at 5:43 PM (autosaved)
Logout
File Edit View Insert Cell Kernel Widgets Help
Trusted Python 3
Run Code
print("Report : \n",
classification_report(y_test, y_pred))
def main():
# Building Phase
data = importdata()
X, Y, X_train, X_test, y_train, y_test = splitdataset(data)
clf_gini = train_using_gini(X_train, X_test, y_train)
clf_entropy = tarin_using_entropy(X_train, X_test, y_train)
# Operational Phase
print("Results Using Gini Index:")
# Prediction using gini
y_pred_gini = prediction(X_test, clf_gini)
cal_accuracy(y_test, y_pred_gini)
print("Results Using Entropy:")
# Prediction using entropy
y_pred_entropy = prediction(X_test, clf_entropy)
cal_accuracy(y_test, y_pred_entropy)
# Calling main function
if __name__=="__main__":
main()
Dataset Length: 625
Dataset Shape: (625, 5)
Dataset:
0 1 2 3 4
0 B 1 1 1 1
1 R 1 1 1 2
2 R 1 1 1 3
3 R 1 1 1 4
4 R 1 1 1 5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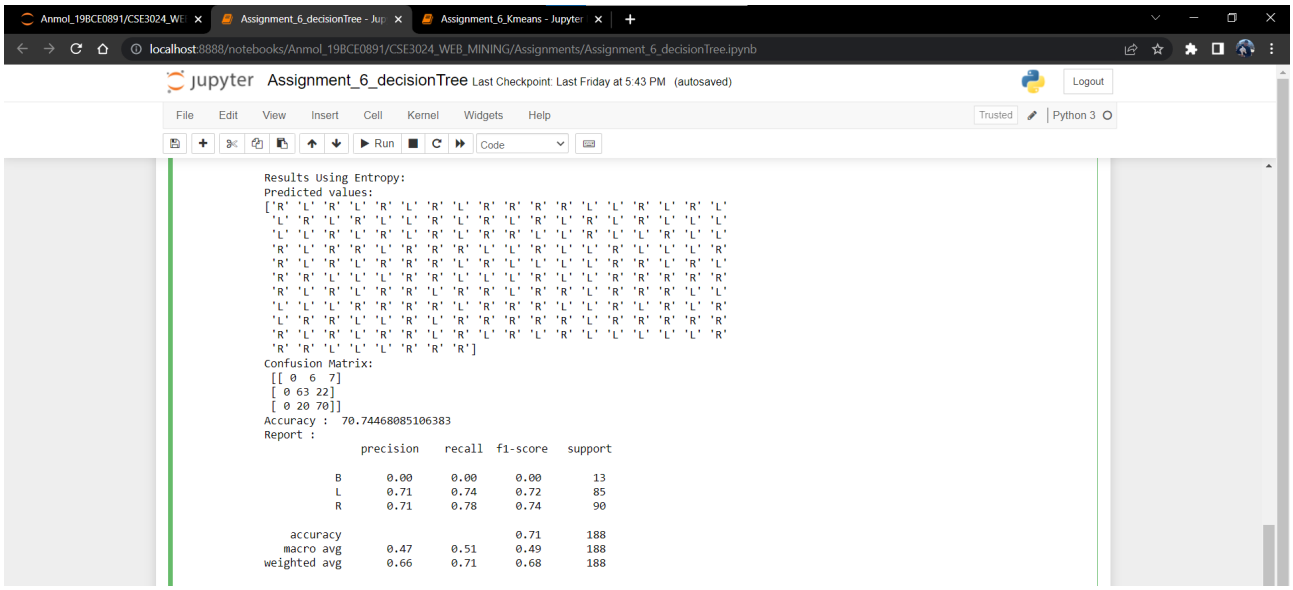
Trusted Python 3
Run Code
Dataset Length: 625
Dataset Shape: (625, 5)
Dataset:
0 1 2 3 4
0 B 1 1 1 1
1 R 1 1 1 2
2 R 1 1 1 3
3 R 1 1 1 4
4 R 1 1 1 5
Results Using Gini Index:
Predicted values:
['R' 'L' 'R' 'R' 'R' 'L' 'R' 'L' 'L' 'L' 'R' 'L' 'L' 'L' 'R' 'L' 'R' 'L'
'L' 'R' 'L' 'R' 'L' 'L' 'R' 'L' 'L' 'L' 'L' 'R' 'L' 'L' 'L' 'L' 'L' 'L'
'L' 'R' 'L' 'L' 'R' 'L' 'R' 'L' 'L' 'R' 'R' 'L' 'L' 'R' 'L' 'R' 'R' 'L'
'R' 'L' 'R' 'R' 'L' 'L' 'R' 'R' 'L' 'L' 'L' 'L' 'L' 'R' 'R' 'L' 'L' 'R'
'R' 'L' 'R' 'L' 'R' 'R' 'R' 'L' 'R' 'L' 'L' 'L' 'L' 'L' 'R' 'R' 'L' 'R' 'L'
'R' 'R' 'L' 'L' 'L' 'R' 'R' 'L' 'L' 'L' 'R' 'L' 'R' 'R' 'R' 'R' 'R' 'R'
'R' 'L' 'R' 'L' 'R' 'R' 'L' 'R' 'R' 'R' 'R' 'R' 'L' 'R' 'L' 'L' 'L' 'L'
'L' 'L' 'L' 'L' 'R' 'R' 'R' 'R' 'L' 'R' 'R' 'R' 'L' 'L' 'R' 'L' 'R' 'L' 'R'
'L' 'L' 'R' 'L' 'L' 'R' 'L' 'R' 'L' 'R' 'R' 'R' 'L' 'R' 'R' 'R' 'R' 'R'
'L' 'L' 'R' 'R' 'R' 'R' 'L' 'R' 'R' 'R' 'L' 'R' 'L' 'L' 'L' 'L' 'R' 'R'
'L' 'R' 'R' 'L' 'L' 'R' 'R' 'R']
Confusion Matrix:
[[0 6 7]
[0 67 18]
[0 19 71]]
Accuracy : 73.40425531914893
Report :
precision recall f1-score support
B 0.00 0.00 0.00 13
L 0.73 0.79 0.76 85
R 0.74 0.79 0.76 90
accuracy 0.73 188
macro avg 0.49 0.53 0.51 188
weighted avg 0.68 0.73 0.71 188



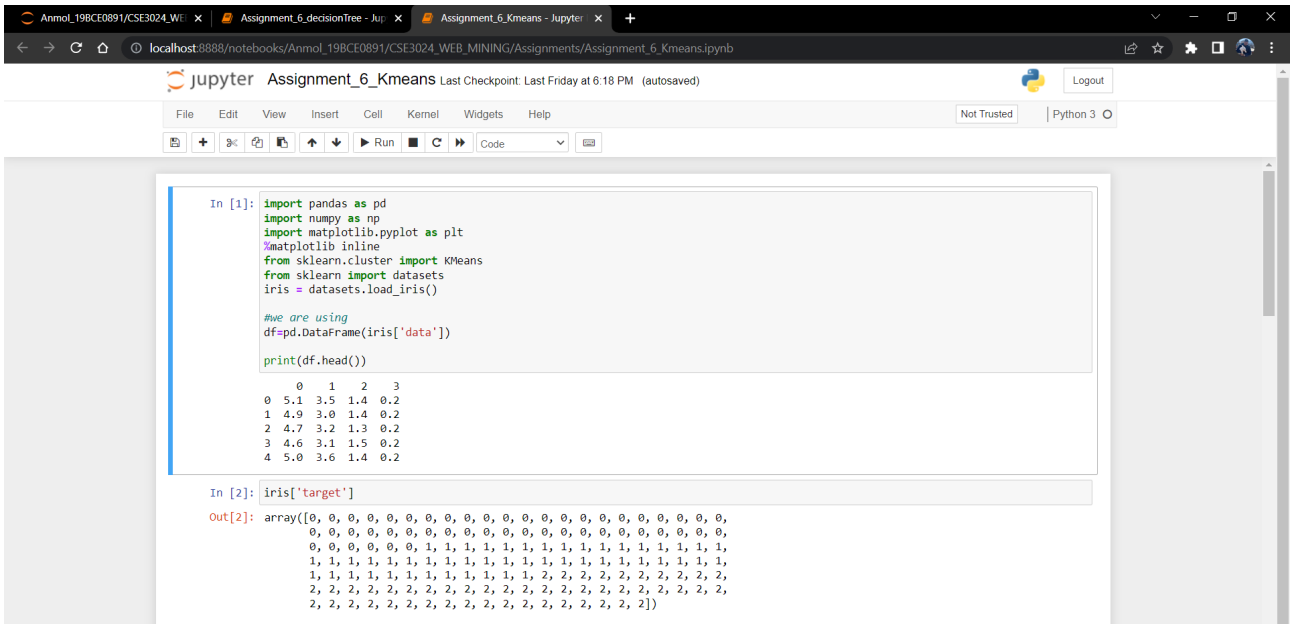
CODE

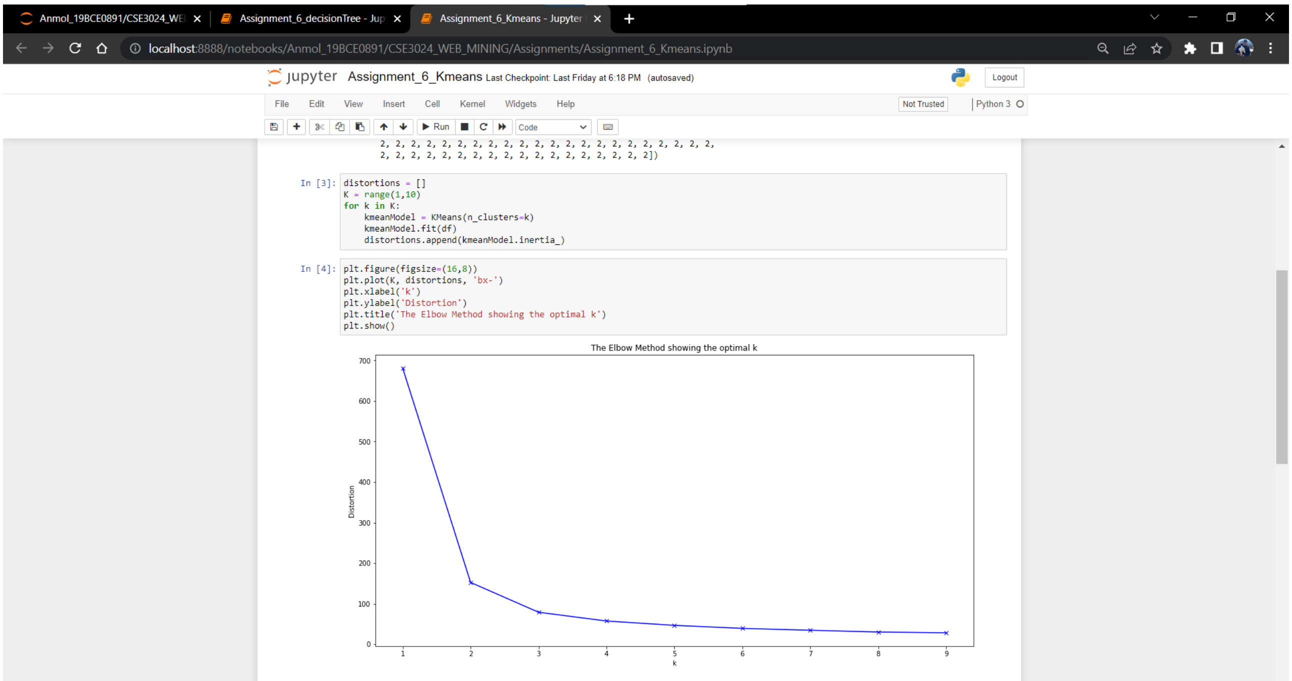
```
import numpy as np
import pandas as pd
from sklearn.metrics import confusion_matrix
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score
from sklearn.metrics import classification_report

def importdata():
    balance_data = pd.read_csv('https://archive.ics.uci.edu/ml/machine-learning-'+'databases/
balance-scale/balance-scale.data',
    sep= ',', header = None)

    print ("Dataset Length: ", len(balance_data))
    print ("Dataset Shape: ", balance_data.shape)

    print ("Dataset:\n",balance_data.head())
    return balance_data

def splitdataset(balance_data):

    X = balance_data.values[:, 1:5]
    Y = balance_data.values[:, 0]

    X_train, X_test, y_train, y_test = train_test_split(
    X, Y, test_size = 0.3, random_state = 100)

    return X, Y, X_train, X_test, y_train, y_test


def train_using_gini(X_train, X_test, y_train):
```

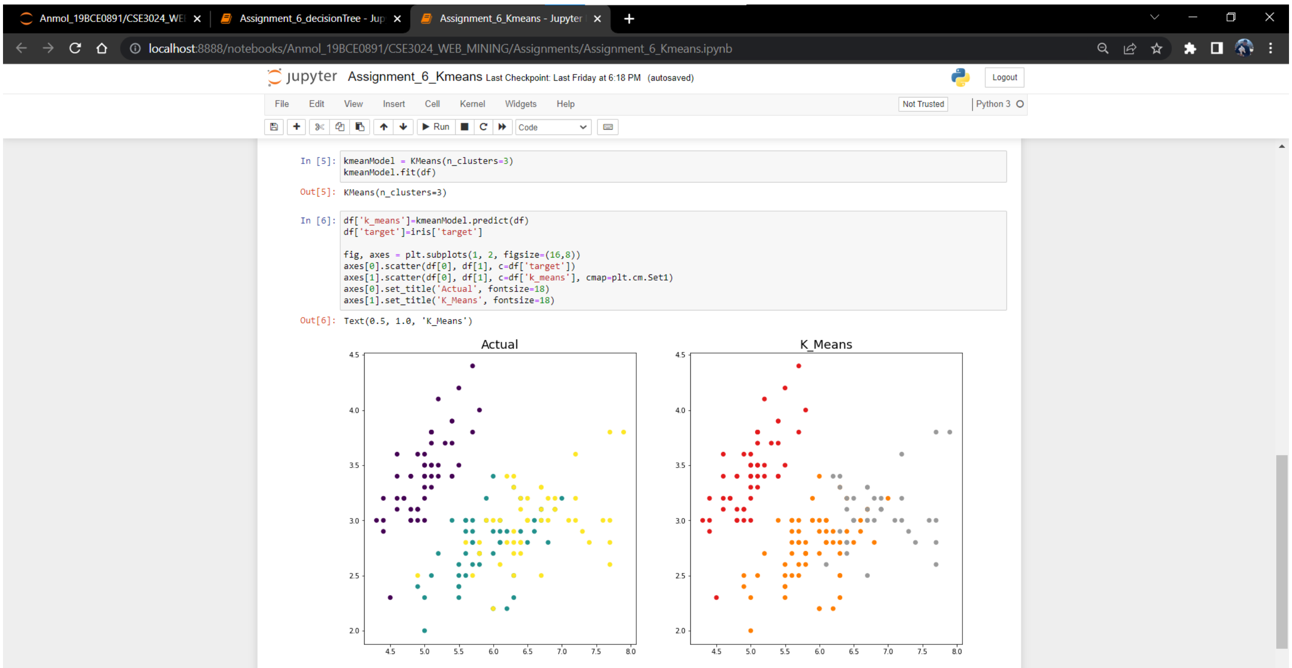
```
    clf_gini = DecisionTreeClassifier(criterion = "gini",
            random_state = 100,max_depth=3, min_samples_leaf=5)

    clf_gini.fit(X_train, y_train)
    return clf_gini


def tarin_using_entropy(X_train, X_test, y_train):

    # Decision tree with entropy
    clf_entropy = DecisionTreeClassifier(criterion = "entropy", random_state = 100,max_depth = 3,
min_samples_leaf = 5)

    # Performing training
    clf_entropy.fit(X_train, y_train)
    return clf_entropy


# Function to make predictions
def prediction(X_test, clf_object):

    y_pred = clf_object.predict(X_test)
    print("Predicted values:")
    print(y_pred)
    return y_pred

# Function to calculate accuracy
def cal_accuracy(y_test, y_pred):

    print("Confusion Matrix:\n",
        confusion_matrix(y_test, y_pred))

    print ("Accuracy : ",
    accuracy_score(y_test,y_pred)*100)

    print("Report : \n",
    classification_report(y_test, y_pred))

def main():

    # Building Phase
    data = importdata()
    X, Y, X_train, X_test, y_train, y_test = splitdataset(data)
    clf_gini = train_using_gini(X_train, X_test, y_train)
    clf_entropy = tarin_using_entropy(X_train, X_test, y_train)

    # Operational Phase
    print("Results Using Gini Index:")
```

```
    # Prediction using gini
    y_pred_gini = prediction(X_test, clf_gini)
    cal_accuracy(y_test, y_pred_gini)

    print("Results Using Entropy:")
    # Prediction using entropy
    y_pred_entropy = prediction(X_test, clf_entropy)
    cal_accuracy(y_test, y_pred_entropy)


# Calling main function
if __name__=="__main__":
    main()
```

2. Create a Python programme that uses the K-means clustering algorithm and displays all clusters in different colours.







CODE

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
from sklearn.cluster import KMeans
from sklearn import datasets
iris = datasets.load_iris()

#we are using
```

```
df=pd.DataFrame(iris['data'])

print(df.head())


iris['target']


distortions = []
K = range(1,10)
for k in K:
    kmeanModel = KMeans(n_clusters=k)
    kmeanModel.fit(df)
    distortions.append(kmeanModel.inertia_)


plt.figure(figsize=(16,8))
plt.plot(K, distortions, 'bx-')
plt.xlabel('k')
plt.ylabel('Distortion')
plt.title('The Elbow Method showing the optimal k')
plt.show()


kmeanModel = KMeans(n_clusters=3)
kmeanModel.fit(df)


df['k_means']=kmeanModel.predict(df)
df['target']=iris['target']

fig, axes = plt.subplots(1, 2, figsize=(16,8))
axes[0].scatter(df[0], df[1], c=df['target'])
axes[1].scatter(df[0], df[1], c=df['k_means'], cmap=plt.cm.Set1)
axes[0].set_title('Actual', fontsize=18)
axes[1].set_title('K_Means', fontsize=18)
```