

**CSE 3024** 

## Web Mining

# LAB ASSESSMENT - 6

NAME: Vibhu Kumar Singh

**REG. NO: 19BCE0215** 

TEACHER: Mr. Hiteshwar Kumar Azad

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

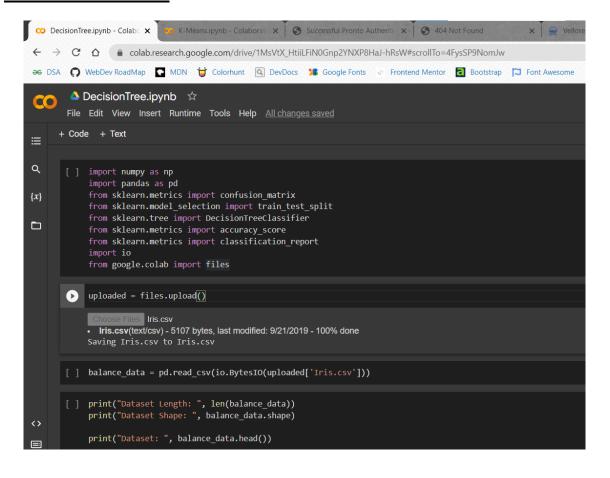
## **HANDWRITTEN CODE:**

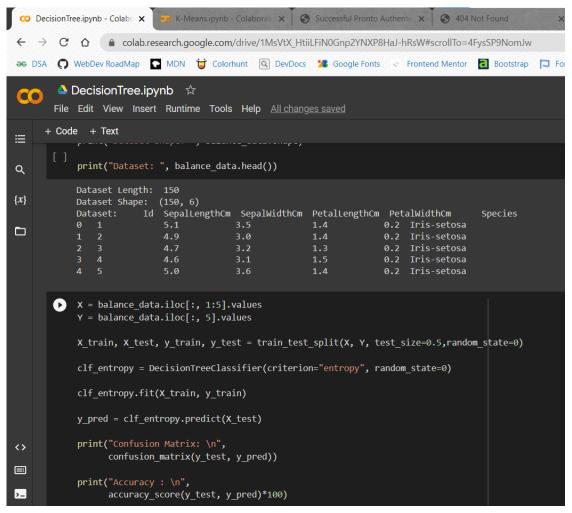
1	VIBHU KUMAR SINGH 19BCE0215	-
1	I suport numpy as no	
	import pandal as pol	
	from sklearn meterics import confusion materix	
	from Sklearn meters inport confusion-mater's from Sklearn model selection import thain test-sp	<i>U</i> <del></del>
	import io	
	uploaded = files.upload()	
_	balance_data = pd. Mead_csv(io. Bytes. ID (uploaded ['In	(([' v2). E
	point (" Dataset Length: ", len (balance_data)	
	point ("Dataset Length: ", len (balance_data)  point ("Dataset Shape: ", balance_data.shape)	
	porent ( " Nataset: ", balance data. head ())	
	X = balance_deta. iloc [: , 1:5]. values Y = balance_deta. iloc [: , 5]. values	*1000
	X train, X-test, y-train, y-test = train_test_split(X, = 0.5, random_etate=0)	y, test_size
	clf_entoupy = DecessionTrueClassifier (seritorion = "entoupy sandom_state:	teogy"
	clf-entropy.fit (x train, y- train)	
	y prid = clf_entropy. predict (x test)	
	pur ("confermon Martonix: \n", confermon martix (y)	est, y procl)
	print ("Accuracy & 14" classification suport (y-test, y-pred) + posint ("Report: \n", (y-test, y-pred))	(00)
	prout ("Prediction & in", y pred)	

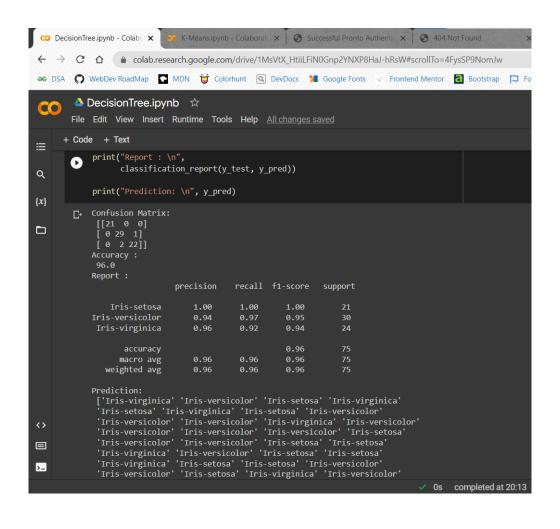
#### **CODE:**

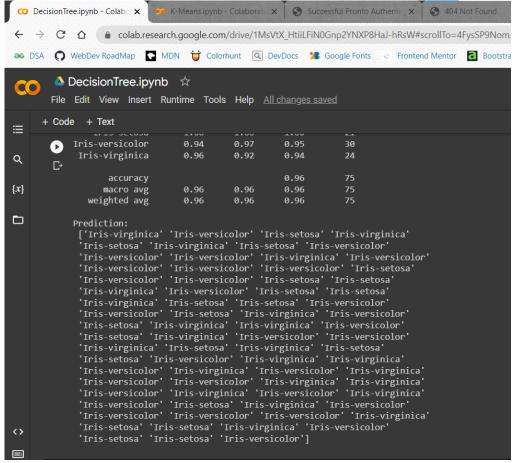
```
#NAME: VIBHU KUMAR SINGH
#ROLL NO: 19BCE0215
#WEB MINING
import numpy as np
import pandas as pd
from sklearn.metrics import confusion_matrix
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score
from sklearn.metrics import classification report
import io
from google.colab import files
uploaded = files.upload()
balance_data = pd.read_csv(io.BytesIO(uploaded['Iris.csv']))
print("Dataset Length: ", len(balance_data))
print("Dataset Shape: ", balance_data.shape)
print("Dataset: ", balance_data.head())
X = balance_data.iloc[:, 1:5].values
Y = balance_data.iloc[:, 5].values
X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size=0.5,random_state
=0)
clf_entropy = DecisionTreeClassifier(criterion="entropy", random_state=0)
clf_entropy.fit(X_train, y_train)
y_pred = clf_entropy.predict(X_test)
print("Confusion Matrix: \n",
      confusion_matrix(y_test, y_pred))
print("Accuracy : \n",
      accuracy_score(y_test, y_pred)*100)
print("Report : \n",
      classification_report(y_test, y_pred))
print("Prediction: \n", y pred)
```

#### **CODE SCREENSHOT:**









#### **OUTPUT:**

Confusion Matrix:

[[21 0 0]

[0291]

[0 2 22]]

Accuracy:

96.0

Report:

precision recall f1-score support

Iris-setosa	1.00	1.00	1.00	21
Iris-versicolor	0.94	0.97	0.95	30
Iris-virginica	0.96	0.92	0.94	24

accuracy		0.9	6 7	5
macro avg	0.96	0.96	0.96	75
weighted avg	0.96	0.96	0.96	75

#### Prediction:

['Iris-virginica' 'Iris-versicolor' 'Iris-setosa' 'Iris-virginica' 'Iris-setosa' 'Iris-virginica' 'Iris-setosa' 'Iris-versicolor' 'Iris-versicolor' 'Iris-versicolor' 'Iris-virginica' 'Iris-versicolor' 'Iris-versicolor' 'Iris-versicolor' 'Iris-versicolor' 'Iris-setosa' 'Iris-versicolor' 'Iris-versicolor' 'Iris-setosa' 'Iris-setosa' 'Iris-virginica' 'Iris-versicolor' 'Iris-setosa' 'Iris-setosa' 'Iris-virginica' 'Iris-setosa' 'Iris-setosa' 'Iris-versicolor' 'Iris-versicolor' 'Iris-setosa' 'Iris-virginica' 'Iris-versicolor' 'Iris-setosa' 'Iris-virginica' 'Iris-virginica' 'Iris-versicolor' 'Iris-setosa' 'Iris-virginica' 'Iris-versicolor' 'Iris-versicolor' 'Iris-virginica' 'Iris-setosa' 'Iris-virginica' 'Iris-setosa' 'Iris-setosa' 'Iris-versicolor' 'Iris-virginica' 'Iris-virginica' 'Iris-versicolor' 'Iris-virginica' 'Iris-versicolor' 'Iris-virginica' 'Iris-versicolor' 'Iris-versicolor' 'Iris-virginica' 'Iris-virginica' 'Iris-versicolor' 'Iris-virginica' 'Iris-versicolor' 'Iris-virginica' 'Iris-versicolor' 'Iris-setosa' 'Iris-virginica' 'Iris-versicolor' 'Iris-versicolor' 'Iris-versicolor' 'Iris-versicolor' 'Iris-virginica' 'Iris-setosa' 'Iris-setosa' 'Iris-virginica' 'Iris-versicolor' 'Iris-setosa' 'Iris-setosa' 'Iris-versicolor']

#### **OUTPUT SCREENSHOT:**

```
Confusion Matrix:
    [[21 0 0]
Ð
     [0291]
     [0 2 22]]
    Accuracy :
    96.0
    Report :
                     precision
                                  recall f1-score
                                                     support
        Iris-setosa
                         1.00
                                   1.00
                                             1.00
                                                          21
    Iris-versicolor
                         0.94
                                   0.97
                                             0.95
                                                         30
    Iris-virginica
                         0.96
                                   0.92
                                             0.94
                                                         24
                                             0.96
                                                         75
           accuracy
                         0.96
                                   0.96
                                             0.96
                                                         75
         macro avg
                                             0.96
                                                         75
       weighted avg
                         0.96
                                   0.96
    Prediction:
     ['Iris-virginica' 'Iris-versicolor' 'Iris-setosa' 'Iris-virginica'
     'Iris-setosa' 'Iris-virginica' 'Iris-setosa' 'Iris-versicolor'
     'Iris-versicolor' 'Iris-versicolor' 'Iris-virginica' 'Iris-versicolor'
     'Iris-versicolor' 'Iris-versicolor' 'Iris-versicolor' 'Iris-setosa'
     'Iris-versicolor' 'Iris-versicolor' 'Iris-setosa' 'Iris-setosa'
     'Iris-virginica' 'Iris-versicolor' 'Iris-setosa' 'Iris-setosa'
     'Iris-virginica' 'Iris-setosa' 'Iris-setosa' 'Iris-versicolor'
     'Iris-versicolor' 'Iris-setosa' 'Iris-virginica' 'Iris-versicolor'
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     'Iris-virginica' 'Iris-setosa' 'Iris-virginica' 'Iris-setosa'
     'Iris-setosa' 'Iris-versicolor' 'Iris-virginica' 'Iris-virginica'
     'Iris-versicolor' 'Iris-virginica' 'Iris-versicolor' 'Iris-virginica'
     'Iris-versicolor' 'Iris-versicolor' 'Iris-virginica' 'Iris-virginica'
     'Iris-versicolor' 'Iris-virginica' 'Iris-versicolor' 'Iris-virginica'
     'Iris-versicolor' 'Iris-setosa' 'Iris-virginica' 'Iris-versicolor'
     'Iris-versicolor' 'Iris-versicolor' 'Iris-versicolor' 'Iris-virginica'
     'Iris-setosa' 'Iris-setosa' 'Iris-virginica' 'Iris-versicolor'
     'Iris-setosa' 'Iris-setosa' 'Iris-versicolor']
```

(P.T.O.)

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

## Ans 2.

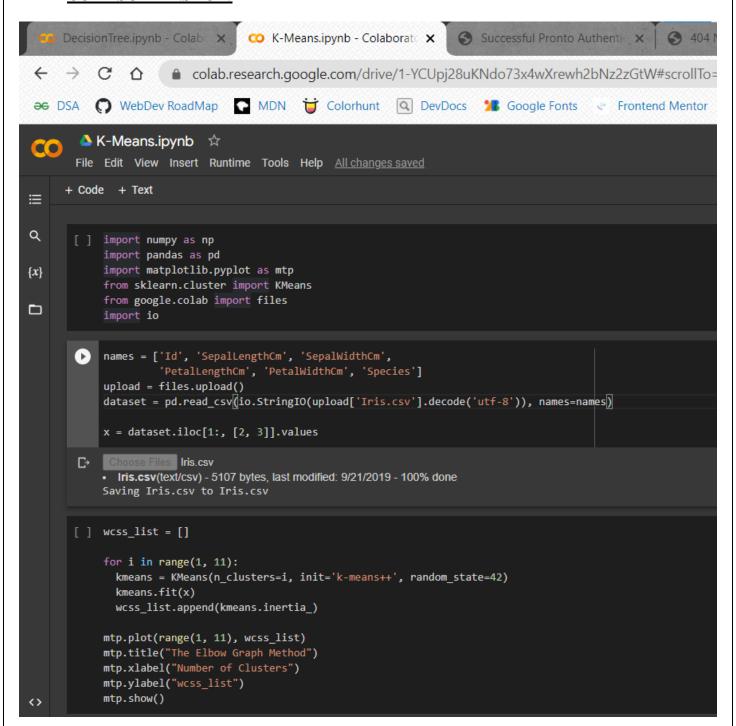
## **HANDWRITTEN CODE:**

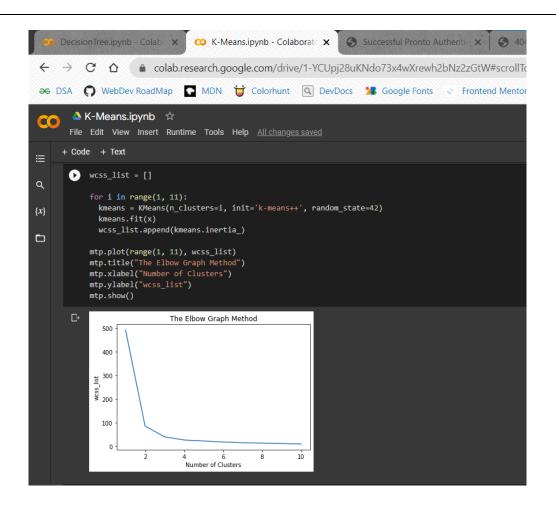
1	VIBHU KUMAR SINGH	19BCF0215
1	·	
	import numpy as up	
	: 'V '	
	impost io	
	names = ['Id', 'Sepallength Cm', 'Species'	'SepalWidth Cm', 'Petal Length Cm
	'Peterwiathan' Species'	7
	upload = files. upland ()	
	dataset = pd - read_cay (io. String)	to (uplad ['iris. cev'].
	dataset = pd. red_csv (io. String)	('utf-8')), names = names
	X = dataset = pd. mad iloc [1:,	[2,3]]. volue
	was_list=[]	
_	for : En range (1,11):	9 0 1 11 1
	Kmeans = KMeans (n-cluster	
_	Itandon _ e	Mt = 42))
_	knueans.fit(X)	2 12 \
_	Wess-list append Censeans	inclus -)
_		4-51
_	mtp. plat ( Grange (1, 11), cucss-	A softer 1 11 )
-	mtp title ("The Elfow Graph	Menod )
-	my. x-lebel ("Number of Clus	ras )
-	nup. ylabel ("wcs_list")	
_	mip. Show ()	
-		
	:	
	mtp. scatter ( Kme ano. cluster - cer	sers_[:, 0], umean cluster
	centus_[:,1], 1=300,	c= "yellow", latel = "Centraid"
	mp. title ('clusters')	

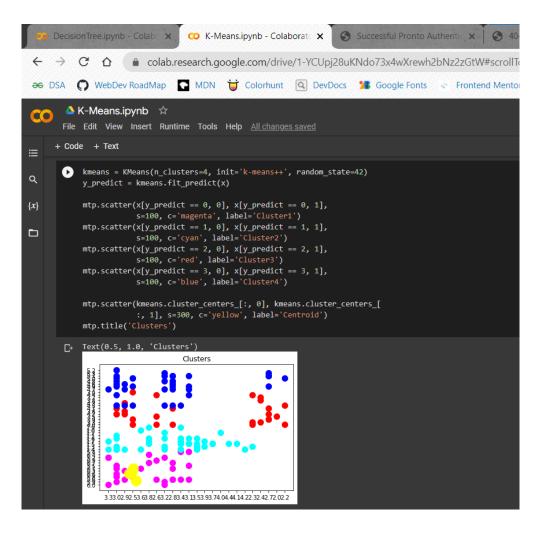
#### **CODE:**

```
#NAME: VIBHU KUMAR SINGH
#ROLL NO: 19BCE0215
#WEB MINING
import numpy as np
import pandas as pd
import matplotlib.pyplot as mtp
from sklearn.cluster import KMeans
from google.colab import files
import io
names = ['Id', 'SepalLengthCm', 'SepalWidthCm',
         'PetalLengthCm', 'PetalWidthCm', 'Species']
upload = files.upload()
dataset = pd.read_csv(io.StringIO(upload['Iris.csv'].decode('utf-8')), names=names)
x = dataset.iloc[1:, [2, 3]].values
wcss_list = []
for i in range(1, 11):
 kmeans = KMeans(n_clusters=i, init='k-means++', random_state=42)
 kmeans.fit(x)
 wcss_list.append(kmeans.inertia_)
mtp.plot(range(1, 11), wcss_list)
mtp.title("The Elbow Graph Method")
mtp.xlabel("Number of Clusters")
mtp.ylabel("wcss list")
mtp.show()
kmeans = KMeans(n_clusters=4, init='k-means++', random_state=42)
y_predict = kmeans.fit_predict(x)
mtp.scatter(x[y_predict == 0, 0], x[y_predict == 0, 1],
            s=100, c='magenta', label='Cluster1')
mtp.scatter(x[y_predict == 1, 0], x[y_predict == 1, 1],
            s=100, c='cyan', label='Cluster2')
mtp.scatter(x[y_predict == 2, 0], x[y_predict == 2, 1],
            s=100, c='red', label='Cluster3')
mtp.scatter(x[y_predict == 3, 0], x[y_predict == 3, 1],
            s=100, c='blue', label='Cluster4')
mtp.scatter(kmeans.cluster_centers_[:, 0], kmeans.cluster_centers_[
            :, 1], s=300, c='yellow', label='Centroid')
mtp.title('Clusters')
```

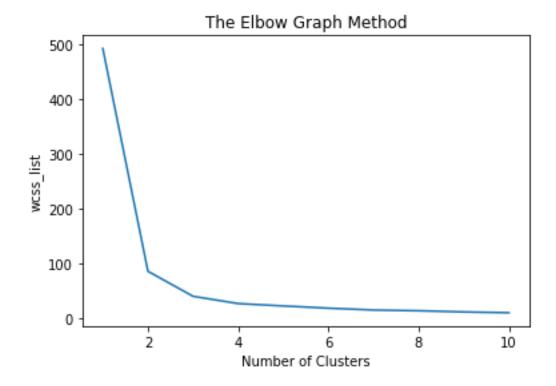
#### **CODE SCREENSHOT:**







### **OUTPUT SCREENSHOT:**



(The Elbow Graph Method)

