Experiment 8 Aim: Implementation of Decision tree using sklearn. Course Outcomes: CO4 and CO5

```
In [1]: import pandas as pd
        from sklearn.tree import DecisionTreeClassifier
        import matplotlib.pyplot as plt
        df = pd.read csv("Exp-8 shows.csv")
In [3]:
In [4]: df
Out[4]:
            Age Experience Rank Nationality Go
         0
             36
                       10
                                       UK NO
             42
                                      USA NO
                       12
             23
         2
                        4
                                        N NO
             52
                                      USA NO
         3
             43
                       21
                              8
                                      USA YES
                       14
                                       UK NO
            44
                              5
         6
             66
                        3
                              7
                                       N YES
             35
                                       UK YES
         7
                       14
                              9
                                       N YES
         8
             52
                       13
                              7
             35
                        5
                                       N YES
                                      USA NO
        10
             24
                        3
                              5
             18
                                       UK YES
        11
        12
            45
                        9
                              9
                                       UK YES
In [5]: d = {'UK': 0, 'USA': 1, 'N': 2}
        df['Nationality'] = df['Nationality'].map(d)
        d = {'YES': 1, 'NO': 0}
        df['Go'] = df['Go'].map(d)
        print(df)
```

Age Experience Rank Nationality Go

```
9
              36
                          10
                                                   0
        0
        1
              42
                          12
                                 4
                                               1
                                                   0
         2
              23
                           4
                                 6
                                               2
                                                  0
         3
              52
                           4
                                 4
                                               1
                                                   0
         4
                          21
                                 8
                                               1
              43
                                                  1
                                                   a
         5
              44
                          14
                           3
                                 7
                                               2
         6
              66
                                                  1
         7
              35
                          14
                                                  1
                                 9
         8
              52
                          13
                                 7
                                                  1
         9
              35
                           5
                                 9
                                               2
                                                  1
        10
              24
                           3
                                 5
                                               1
                                                  0
        11
             18
                           3
                                 7
                                                  1
        12
             45
                                               0
                                                  1
In [6]: features = ['Age', 'Experience', 'Rank', 'Nationality']
        X = df[features]
         v = df['Go']
        dtree = DecisionTreeClassifier()
         dtree = dtree.fit(X, y)
In [8]: print(dtree.predict([[40, 10, 7, 1]]))
         print("[1] means 'GO'")
         print("[0] means 'NO'")
         [0]
         [1] means 'GO'
        [0] means 'NO'
        C:\Users\admin\anaconda3\Lib\site-packages\sklearn\base.py:439: UserWarning: X does not have valid feature names, but DecisionTr
        eeClassifier was fitted with feature names
          warnings.warn(
```

Part-2 Build a decision tree model for drug dataset About the dataset Imagine that you are a medical researcher compiling data for a study. You have collected data about a set of patients, all of whom suffered from the same illness. During their course of treatment, each patient responded to one of 5 medications, Drug A, Drug B, Drug c, Drug x and y.

Part of your job is to build a model to find out which drug might be appropriate for a future patient with the same illness. The features of this dataset are Age, Sex, Blood Pressure, and the Cholesterol of the patients, and the target is the drug that each patient responded to.

It is a sample of multiclass classifier, and you can use the training part of the dataset to build a decision tree, and then use it to predict the class of a unknown patient, or to prescribe a drug to a new patient.

```
In [18]: df = pd.read_csv('Exp8_DT.csv')
    df.head()
```

Out[18]:		Age	Sex	ВР	Cholesterol	Na_to_K	Drug
	0	23	F	HIGH	HIGH	25.355	drugY
	1	47	М	LOW	HIGH	13.093	drugC
	2	47	М	LOW	HIGH	10.114	drugC
	3	28	F	NORMAL	HIGH	7.798	drugX
	4	61	F	LOW	HIGH	18.043	drugY

```
In [21]: from sklearn.model_selection import train_test_split
    from sklearn.preprocessing import LabelEncoder
    input = df.iloc[:, :-1]
    Target = df.iloc[:, -1]
    le_sex = LabelEncoder()
    le_BP = LabelEncoder()
    le_cholesterol = LabelEncoder()
    input['Sex'] = le_sex.fit_transform(input['Sex'])
    input['BP'] = le_BP.fit_transform(input['BP'])
    input['Cholesterol'] = le_cholesterol.fit_transform(input['Cholesterol'])
    input
```

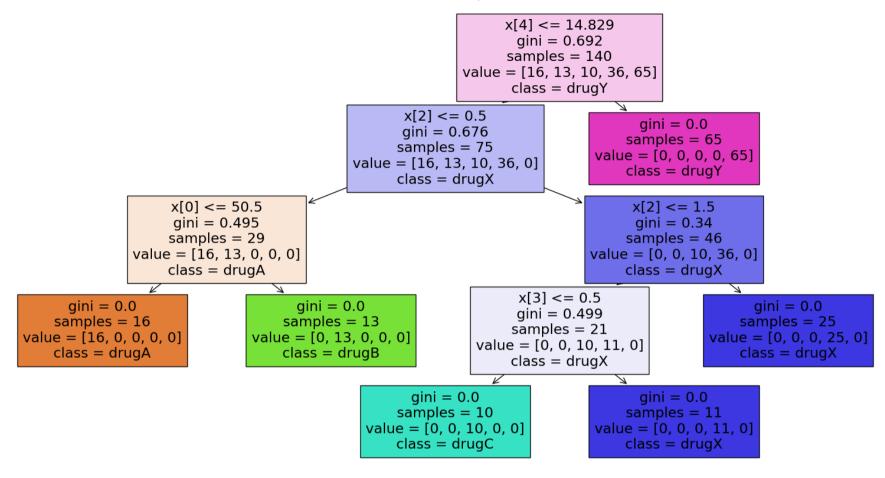
Out[21]: Age Sex BP Cholesterol Na to K

out[ZI].		Age	эех	DP	Cholesteroi	-
	0	23	0	0	0	25.355
	1	47	1	1	0	13.093
	2	47	1	1	0	10.114
	3	28	0	2	0	7.798
	4	61	0	1	0	18.043
	195	56	0	1	0	11.567
	196	16	1	1	0	12.006
	197	52	1	2	0	9.894
	198	23	1	2	1	14.020
	199	40	0	1	1	11.349
	200 rd	ows ×	5 co	lumr	าร	
[22]:	clf clf. accu	= Deci fit(X_ racy =	isio _tra = cl	nTre in, f.sc	y_train, y eclassifie y_train) ore(X_test racy:", ro	r() , y_test)
	Mode	l Accı	uracy	y: 1	00.0	
[24]:	from		arn.	tree	p import pl np.unique(

plt.show()

plt.figure(figsize=(20,10))

plot_tree(clf, class_names=unique_classes, filled=True)



```
In [25]: y_pred = clf.predict(X_test)
In [26]: from sklearn.metrics import classification_report, confusion_matrix, precision_score, recall_score, accuracy_score
In [28]: print(classification_report(y_test, y_pred))
```

		precision	recall	f1-score	support
	drugA	1.00	1.00	1.00	7
	drugB	1.00	1.00	1.00	3
	drugC	1.00	1.00	1.00	6
	drugX	1.00	1.00	1.00	18
	drugY	1.00	1.00	1.00	26
	accuracy			1.00	60
	macro avg	1.00	1.00	1.00	60
	weighted avg	1.00	1.00	1.00	60
In [29]:	print(confusi	on_matrix(y_	test, y_p	ored))	
	[[7000	0]			
	[0300	0]			
	[0060	-			
	[0 0 0 18				
	[0000	26]]			
In [31]:	print(precisi	on_score(y_t	est, y_pr	red, averag	e='micro'))
	1.0				
		, , ,			122
In [33]:	<pre>print(recall_</pre>	score(y_test	, y_pred,	average='	macro'))
	1.0				
In [35]:	print(accurac	y_score(y_te	st, y_pre	ed))	
	1.0				
	1.0				

Learning Outcomes I Got to know about decision tree supervised learning algorithm for classification. Got to know about how to predict using decision tree Got to know how to draw a decision tree

Result/ Conclusion We have successfully trained the model and predicted the results.