LAB4 : Implement Decision tree algorithm for classification

Import Libraries

```
In [15]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

import Dataset

```
In [16]: dataset = pd.read_csv(r"C:\Users\Admin\Downloads\environmental_alerts.csv")
```

EDA Steps

		1					
In [17]:	dataset.head()						
Out[17]:	T	emperature	Humidity	AirQualityIndex	NoiseLevel	AlertLevel	
	0	1.569100	6.3465	-0.1828	-2.4099	0	
	1	-0.278020	8.1881	-3.1338	-2.5276	0	
	2	0.051979	7.0521	-2.0541	-3.1508	0	
	3	-1.755900	11.9459	3.0946	-4.8978	0	
	4	2.428700	9.3821	-3.2477	-1.4543	0	
In [18]:	dataset.shape						
Out[18]:	(100, 5)						
In [19]:	dataset.columns						
Out[19]:	<pre>Index(['Temperature', 'Humidity', 'AirQualityIndex', 'NoiseLevel',</pre>						
In [20]:	data	set.info()					

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99
Data columns (total 5 columns):
    Column
                    Non-Null Count Dtype
--- -----
                   -----
    Temperature 100 non-null Humidity 100 non-null
0
                                   float64
                                   float64
    AirQualityIndex 100 non-null
                                   float64
3
    NoiseLevel 100 non-null
                                   float64
4 AlertLevel
                   100 non-null
                                   int64
dtypes: float64(4), int64(1)
```

memory usage: 4.0 KB

In [21]:	<pre>dataset.describe()</pre>
----------	-------------------------------

Out[21]:		Temperature	Humidity	AirQualityIndex	NoiseLevel	AlertLevel
	count	100.000000	100.000000	100.000000	100.000000	100.000000
	mean	0.938703	2.753227	0.534520	-1.139743	0.410000
	std	2.490773	5.391447	4.039764	1.942759	0.494311
	min	-6.367900	-12.740600	-5.149000	-7.758100	0.000000
	25%	-0.769175	-0.451628	-2.886775	-2.439325	0.000000
	50%	0.929765	2.911450	0.191215	-0.576300	0.000000
	75 %	2.616550	7.107975	2.710300	0.320132	1.000000
	max	5.735300	12.117700	15.557300	1.618100	1.000000

Preprocessing Steps

Step 1 : Seprate input and output variables

```
In [22]: X = dataset.iloc[:, :-1].values
         y = dataset.iloc[:, -1].values
```

Step 4 : splitting Data inti training and testing

```
In [23]: 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)
In [24]: print(X train.shape)
         print(X test.shape)
        (75, 4)
```

(25, 4)

```
In [25]: from sklearn.tree import DecisionTreeClassifier
         clf = DecisionTreeClassifier()
         clf.fit(X train, y train)
Out[25]:
         DecisionTreeClassifier
         DecisionTreeClassifier()
In [26]: print(X test)
                                      -1.4466 ]
        [[ 3.2422
                    6.2265
                              0.12224
                    6.9176
         [ 2.8561
                             -0.79372
                                       0.48403 ]
         [ 0.051979 7.0521
                             -2.0541
                                      -3.1508 ]
         [ 0.26517
                    2.4066
                             -2.8416
                                      -0.59958 ]
         [ 0.5195
                   -3.2633
                             3.0895
                                      -0.9849 ]
                           -3.3522
         [ 0.23874
                    2.0879
                                      -0.66553 ]
         [-1.3414
                   -2.0776
                           2.8093
                                       0.60688 1
                    5.2808
                             -2.2598
         [ 5.7353
                                       0.075416]
         [-0.11996
                    6.8741
                             0.91995 -0.6694 ]
                   -6.572
                             10.5251
                                       -0.16381 ]
         [-3.5895
         [ 3.6575
                    7.2797
                             -2.2692
                                      -1.144
         [ 0.3434
                    0.12415 -0.28733
                                       0.14654 ]
         [-1.9555]
                    0.20692
                              1.2473
                                       -0.3707 1
         [-0.77288 -7.4473
                             6.492
                                       0.36119 ]
                    3.9606
                             -1.983
         [ 5.1731
                                       0.40774 ]
                    2.8452 -3.6436
         [-0.49081
                                      -3.1004
         [-0.69879 -3.3771
                            4.1211
                                       1.5043
         [-4.577
                    3.4515
                              0.66719 -0.94742 ]
         [-6.3679]
                    8.0102
                              0.4247
                                      -3.2207 1
         [ 4.0047
                    0.45937
                              1.3621
                                       1.6181
                    9.6325
         [ 2.229
                             -3.1123
                                      -2.7164 ]
         [ 0.27331
                    4.8773
                             -4.9194
                                      -5.8198 ]
         [-1.7559]
                   11.9459
                              3.0946
                                       -4.8978 ]
         [ 2.4008
                    9.3593
                             -3.3565
                                       -3.3526
         [-1.8391]
                   -9.0883
                              9.2416
                                       -0.10432 ]]
In [28]: dataset['AlertLevel'].value_counts()
Out[28]: AlertLevel
              59
         1
              41
         Name: count, dtype: int64
         Train the Decision Tree Classification model on the
```

Train the Decision Tree Classification model on the Training set

```
Out[29]:
                                DecisionTreeClassifier
         DecisionTreeClassifier(ccp_alpha=0.0001, criterion='entropy', rando
         m_state=0,
                                 splitter='random')
In [30]: print(classifier.classes_)
        [0 1]
In [31]: classifier.max features
Out[31]: 4
In [32]: classifier.tree
Out[32]: <sklearn.tree._tree.Tree at 0x17be85cf4b0>
In [33]: classifier.n outputs
Out[33]: 1
         Predicting the Test set results
In [34]: ypred = classifier.predict(X test)
In [35]: print(ypred)
        [0\ 0\ 0\ 1\ 0\ 1\ 1\ 0\ 0\ 1\ 0\ 1\ 1\ 1\ 0\ 0\ 1\ 0\ 0\ 1]
         Compare the predicted and Actual Output
In [36]: print(np.concatenate((ypred.reshape(len(ypred),1),
                               y_test.reshape(len(y_test),1)),
```

1))

```
[[0 0]]
[0 0]
 [0 0]
[1 1]
[0 0]
[1\ 1]
 [1 \ 1]
 [0 0]
 [0 0]
[1\ 1]
 [0 0]
[1\ 1]
 [1\ 1]
[1\ 1]
[0 0]
[1 1]
[1 1]
 [1\ 1]
[1\ 1]
[0 0]
[0 0]
[1\ 1]
 [0 0]
 [0 0]
 [1 1]]
```

Making the Confusion Matrix

```
In [22]: from sklearn.metrics import confusion matrix
         cm = confusion_matrix(y_test, ypred)
         print(cm)
        [[194
                1]
         [ 1 147]]
In [30]: print("Accuracy =",(194+147)/343*100)
        Accuracy = 99.41690962099126
In [24]: from sklearn.metrics import accuracy_score
         print("Test Accuracy =",accuracy_score(y_test,ypred))
        Test Accuracy = 0.9941690962099126
         Build the Classification Report
In [25]: from sklearn.metrics import classification_report
         print(classification_report(y_test,ypred))
```

	precision	recall	f1-score	support
0 1	0.99 0.99	0.99 0.99	0.99 0.99	195 148
accuracy macro avg	0.99	0.99	0.99 0.99	343 343
weighted avg	0.99	0.99	0.99	343

Visualising the Training set results

VISUALIZE TEXT REPRESENTATION

```
In [27]: !pip install -U scikit-learn
```

Requirement already up-to-date: scikit-learn in c: $programdata\anaconda3\lib\site-packages (1.0.2)$

Requirement already satisfied, skipping upgrade: scipy>=1.1.0 in c:\programd ata\anaconda3\lib\site-packages (from scikit-learn) (1.2.1)

Requirement already satisfied, skipping upgrade: numpy>=1.14.6 in c:\program data\anaconda3\lib\site-packages (from scikit-learn) (1.21.6)

Requirement already satisfied, skipping upgrade: threadpoolctl>=2.0.0 in c:\programdata\anaconda3\lib\site-packages (from scikit-learn) (3.1.0)

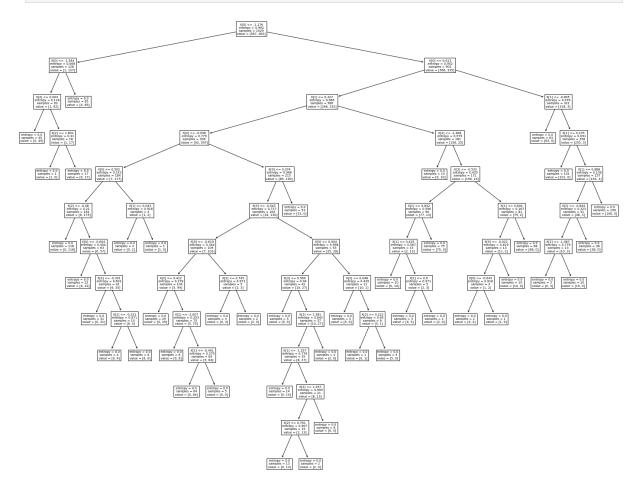
Requirement already satisfied, skipping upgrade: joblib>=0.11 in c:\programd ata\anaconda3\lib\site-packages (from scikit-learn) (1.3.2)

```
|--- feature_0 <= -1.18
   |---| feature 0 <= -1.54
       |--- feature 2 <= 0.00
          |--- class: 1
       |--- feature_2 > 0.00
           |--- feature 2 <= 1.80
             |--- class: 0
           |--- feature_2 > 1.80
          | |--- class: 1
   |---| feature 0 > -1.54
   | |--- class: 1
|---| feature 0 > -1.18
   |---| feature 0 <= 0.61
       |--- feature 1 <= 0.33
           |--- feature_2 <= -0.06
               |--- feature 0 <= 0.50
                  |--- feature 2 <= -0.48
                     |--- class: 1
                   |---| feature 2 > -0.48
                    |--- feature_0 <= -0.60
                     | |--- class: 1
                      |--- feature 0 > -0.60
                          |--- feature 2 <= -0.30
                         | |--- class: 1
                          |---| feature 2 > -0.30
                          | |--- feature 1 <= -0.22
                             | |--- class: 1
                              |---| feature 1 > -0.22
                             | |--- class: 0
                         |--- feature_0 > 0.50
                  |--- feature 1 <= 0.04
                   | |--- class: 1
                   |---| feature 1 > 0.04
                  | |--- class: 0
            --- feature 2 > -0.06
               |--- feature_0 <= 0.07
                   |--- feature_0 <= -0.54
                      |--- feature 0 <= -0.62
                          |--- feature 2 <= 0.41
                             |--- class: 1
                          |---| feature 2 > 0.41
                              |--- feature_1 <= -2.01
                             | |--- class: 1
                              |--- feature_1 > -2.01
                                |--- feature 1 <= -0.44
                                  | |--- class: 1
                                  |--- feature 1 > -0.44
                             |--- feature_0 > -0.62
                          |---| feature 2 <= 0.75
                              |--- class: 1
                           |--- feature 2 > 0.75
                              |--- class: 0
                          |--- feature_0 > -0.54
                      |--- feature 0 <= 0.00
                          |--- feature_3 <= 0.57
```

```
|--- class: 0
                    |--- feature_3 > 0.57
                        |--- feature 2 <= 1.58
                           |--- feature 1 <= -1.26
                           | |--- class: 1
                           |---| feature 1 > -1.26
                           | |--- feature 2 <= 1.30
                              | |--- truncated branch of depth 2
                           | |--- feature 2 > 1.30
                          |  |  |--- class: 0
                        |---| feature 2 > 1.58
                    | | |--- class: 0
               |--- feature 0 > 0.00
               | |--- feature_2 <= 0.05
                    | |--- class: 0
                    |---| feature 2 > 0.05
                      |--- feature 2 <= 0.22
                        | |--- class: 1
                      |--- feature 2 > 0.22
                  | | |--- class: 0
          |--- feature 0 > 0.07
         |--- class: 0
  |---| feature 1 > 0.33
      |--- feature 2 <= -1.47
          |--- class: 1
      |---| feature 2 > -1.47
          |--- feature 3 <= -0.53
             |--- feature 1 <= 0.83
             | --- feature 1 <= 0.43
               | |--- class: 1
             | --- feature 1 > 0.43
               | |--- feature_1 <= 0.60
             |--- feature_1 > 0.60
             |--- feature_1 > 0.83
             | |--- class: 0
          |---| feature 3 > -0.53
             |--- feature 1 <= 0.64
             | |--- feature 0 <= -0.02
             | | |--- feature 0 <= -0.64
               |  |  |--- class: 1
               | |--- feature 0 > -0.64
               | | |--- class: 0
             | |--- feature 0 > -0.02
               | |--- class: 0
             |---| feature 1 > 0.64
         | |--- class: 0
--- feature_0 > 0.61
  |---| feature 1 <= -0.80
      |--- class: 0
  |--- feature 1 > -0.80
      |--- feature 1 <= 0.28
      | |--- class: 0
      |---| feature 1 > 0.28
      | |--- feature 1 <= 0.89
```

Visualize the tree

```
In [29]: from sklearn import tree
fig = plt.figure(figsize=(25,20))
tree.plot_tree(classifier)
plt.show()
```



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