

In [1]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
sns.set(color_codes=True)
from sklearn.preprocessing import LabelEncoder
```

Importing the libraries

In [2]:

```
dataset = pd.read_csv("Social_Network_Ads.csv")
```

Importing the dataset

In [3]:

```
dataset.head()
```

Out [3]:

User ID	Gender	Age	EstimatedSalary	Purchased	
0	15624510	Male	19	19000	0
1	15810944	Male	35	20000	0
2	15688575	Female	26	43000	0
3	15603246	Female	27	57000	0
4	15804002	Male	19	76000	0

In [4]:

```
dataset.shape
```

Out [4]:

```
(480, 5)
```

In [5]:

```
dataset.columns
```

Out [5]:

```
Index(['User ID', 'Gender', 'Age', 'EstimatedSalary', 'Purchased'], dtype='object')
```

In [6]:

```
dataset.isna().any()
```

Out [6]:

```
User ID      False
Gender      False
Age         False
EstimatedSalary  False
Purchased   False
dtype: bool
```

In [7]:

```
dataset.isna().sum()
```

Out [7]:

```
User ID      0
Gender       0
Age          0
EstimatedSalary  0
Purchased    0
dtype: int64
```

In [8]:

```
from sklearn.preprocessing import LabelEncoder
```

In [9]:

```
Gender=LabelEncoder()
```

In [10]:

```
dataset['Gender']=Gender.fit_transform(dataset['Gender'])
```

In [11]:

```
dataset
```

Out [11]:

User ID	Gender	Age	EstimatedSalary	Purchased	
0	15624510	1	19	19000	0
1	15810944	1	35	20000	0
2	15688575	0	26	43000	0
3	15603246	0	27	57000	0
4	15804002	1	19	76000	0
...	...	...	...	...	...
395	15691863	0	46	41000	1
396	15706701	1	51	23000	1
397	15645296	0	50	20000	1
398	15755018	1	36	33000	0
399	15984041	0	49	36000	1

400 rows × 5 columns

In [12]:

```
feature_cols=['Gender','Age','EstimatedSalary']
X=dataset[feature_cols]
y=dataset.Purchased
```

In [13]:

```
X
```

Out [13]:

Gender	Age	EstimatedSalary
0	1	19
1	1	35
0	0	26
3	0	27
4	1	19
...	...	...
395	0	46
396	1	51
397	0	50
398	1	36
399	0	49

400 rows × 3 columns

In [14]:

```
y
```

Out [14]:

```
0    0
1    0
2    0
3    0
4    0
...
395   1
396   1
397   1
398   0
399   1
Name: Purchased, Length: 480, dtype: int64
```

Splitting the dataset into the Training set and Test set

In [15]:

```
from sklearn.model_selection import train_test_split
```

In [16]:

```
X_train, X_test, y_train, y_test = train_test_split(X,y,test_size=0.2)
```

Feature Scaling

In [17]:

```
from sklearn.preprocessing import StandardScaler
```

In [18]:

```
sc= StandardScaler()
```

In [19]:

```
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
```

Fitting Decision Tree Classification to the Training set

In [20]:

```
from sklearn.tree import DecisionTreeClassifier
```

In [21]:

```
classifier=DecisionTreeClassifier(criterion='gini')
```

In [22]:

```
classifier.fit(X_train,y_train)
```

Out [22]:

```
DecisionTreeClassifier()
```

Predicting the Test set results

In [23]:

```
classifier.predict(X_test)
```

Out [23]:

```
array([0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1,
       1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1,
       1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 1, 0, 1, 1,
       1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 2], dtype=int64)
```

In [24]:

```
X_test
```

Out [24]:

```
array([[ -0.98142253, -1.66291941,  0.36467588],
       [ -0.98142253, -0.74576609, -1.37463394],
       [ -0.98142253, -0.45857666, -0.53387969],
       [  0.18182912, -0.73561911,  0.39670455],
       [ -0.98142253, -0.73259316, -1.28750995],
       [ -0.98142253,  0.93104959,  0.77647517],
       [ -0.98142253,  0.00463209,  1.26323146],
       [ -0.98142253,  0.18991559,  2.10381571],
       [  0.18182912, -1.19971066,  0.27771889],
       [  0.18182912, -1.01442716, -1.43243527],
       [  0.18182912, -0.68896966, -0.50489403],
       [  0.18182912, -1.16706961, -1.14257363],
       [ -0.98142253, -0.27329316,  0.04583358],
       [  0.18182912, -1.29235241, -0.33989005],
       [  0.18182912, -0.36593491, -0.765765  ],
       [ -0.98142253,  0.00463209,  1.23424579],
       [  0.18182912, -1.19971066,  0.5896642  ],
       [ -0.98142253, -0.82914366, -0.64982335],
       [  0.18182912, -1.18706961, -1.0846073  ],
       [  0.18182912,  0.18991559, -0.12880804],
       [  0.18182912,  0.18991559,  0.24873323],
       [ -0.98142253,  1.48698609,  1.00236948],
       [ -0.98142253,  0.83846784,  1.8429473  ],
       [ -0.98142253, -0.82914366,  0.39366155],
       [  0.18182912, -0.27329316,  0.16177624],
       [ -0.98142253,  0.26593491, -0.30199438],
       [ -0.98142253, -1.11633209, -1.43243527],
       [ -0.98142253,  0.28255734,  0.38670455],
       [ -0.98142253, -1.16706961, -0.99705931],
       [ -0.98142253, -0.73561911,  1.35018845],
       [  0.18182912, -0.27329316, -0.56286536],
       [ -0.98142253, -0.27329316, -0.56286536],
       [ -0.98142253, -0.45857666, -1.20654996],
       [ -0.980406962358,  0.64896016,  0.56757553],
       [ -0.98142253, -1.29235241, -1.0846073  ],
       [ -0.98142253, -1.01442716, -0.4469227  ],
       [ -0.98142253,  0.93104959,  1.7849734  ],
       [  0.18182912, -0.18065141, -0.27306872],
       [  0.18182912,  0.93104959,  0.05996571],
       [ -0.98142253,  0.09727384,  1.8719394  ],
       [  0.18182912, -0.27329316, -0.88167066],
       [  0.18182912,  0.27519099, -0.12880804],
       [ -0.98142253,  1.95610884, -0.79475967],
       [ -0.98142253, -0.18065141,  0.16177624],
       [ -0.98142253, -0.92178541, -0.30199438],
       [ -0.98142253, -0.45857666, -1.26323146],
       [ -0.98142253, -0.68896966,  0.68351818],
       [  0.18182912, -0.92178541,  0.56757553],
       [  0.18182912, -1.19971066,  0.5896642  ],
       [  0.18182912, -0.27329316,  0.10384913],
       [ -0.98142253, -1.57627766, -1.54837792],
       [  0.18182912, -0.36459416, -0.18605173],
       [  0.18182912,  0.69727384,  0.21374756],
       [  0.18182912, -0.82914366,  0.38670455],
       [  0.18182912,  0.93104959, -1.05562164],
       [  0.18182912, -0.64896016, -0.09999174],
       [  0.18182912,  1.95610884, -0.12880137],
       [ -0.98142253,  0.83846784, -0.64982335],
       [  0.18182912,  0.28255734, -0.50489403],
       [ -0.98142253,  0.83846784, -0.56286536],
       [ -0.98142253,  0.83846784,  2.16178703],
       [  0.18182912, -1.16706961, -0.73673659],
       [  0.18182912,  0.18991559,  1.00236948],
       [  0.18182912,  0.74576609,  1.37917411],
       [ -0.98142253,  2.04275059,  1.11830314],
       [  0.18182912,  0.27519099, -0.4469227  ],
       [  0.18182912,  0.18991559, -0.35966571],
       [  0.18182912,  0.00463209, -0.30199438],
       [ -0.98142253, -0.27329316, -0.73677934],
       [  0.18182912, -1.02069134, -0.12880804],
       [  0.18182912,  0.28255734,  0.27771889],
       [  0.18182912, -0.73561911,  0.50960642 ],
       [  0.18182912,  0.18991559, -0.35966571],
       [  0.18182912,  0.18991559,  0.05996571],
       [  0.18182912,  0.00463209,  0.04583358],
       [  0.18182912, -0.92178541, -1.0846073  ],
       [ -0.98142253, -0.27329316,  0.79946064 ],
       [ -0.98142253, -0.64386616,  1.408151978]]]
```

In [25]:

```
classifier.score(X_test, y_test)
```

Out [25]:

```
0.875
```

Making the Confusion Matrix

In [26]:

```
y_pred = classifier.predict(X_test)
```

In [27]:

```
from sklearn.metrics import confusion_matrix
```

In [28]:

```
cm = confusion_matrix(y_test, y_pred)
```

In [29]:

```
print(cm)
```

[ [53 9]
 [ 1 17]]

Visualising the Training set results

In [30]:

```
plt.figure(figsize=(12,8))
from sklearn import tree
tree.plot_tree(classifier.fit(X_train, y_train))
```

Out [30]:

```
[Text(360.375, 419.3485714285714, 'X[1] <= 0.422\ngini = 0.476\nnsamples = 320\nvalue = [195, 125]'),
Text(185.07000000000002, 388.2857142857143, 'X[2] <= 0.582\ngini = 0.291\nnsamples = 221\nvalue = [182, 39]'),
Text(165.00000000000001, 357.22285714285715, 'X[1] <= -0.134\ngini = 0.683\nnsamples = 185\nvalue = [177, 8]'),
Text(50.220000000000006, 326.15999999999997, 'X[0] <= 0.134\ngini = 0.222\nnsamples = 63\nvalue = [55, 8]'),
Text(65.100000000000004, 295.09714285714284, 'X[0] <= 0.019\ngini = 0.165\nnsamples = 11\nvalue = [10, 1]'),
Text(94.86, 295.09714285714284, 'X[2] <= 0.408\ngini = 0.375\nnsamples = 32\nvalue = [24, 8]'),
Text(70.98, 264.0342857142857, 'X[0] <= 0.031\ngini = 0.35\nnsamples = 31\nvalue = [24, 7]'),
Text(70.76, 232.97142857142856, 'X[0] <= 0.019\ngini = 0.5\nnsamples = 2\nvalue = [1, 0]'),
Text(14.88, 201.90857142857142, 'X[0] <= 0.019\ngini = 0.165\nnsamples = 11\nvalue = [10, 1]'),
Text(44.64, 201.90857142857142, 'X[2] <= 0.234\ngini = 0.375\nnsamples = 16\nvalue = [12, 4]'),
Text(98.28, 176.84571428571428, 'gini = 0.0\nnsamples = 1\nvalue = [0, 1]'),
Text(89.26, 176.84571428571428, 'X[1] <= 0.194\ngini = 0.32\nnsamples = 5\nvalue = [4, 1]'),
Text(114.88, 139.78285714285715, 'X[2] <= 0.234\ngini = 0.444\nnsamples = 9\nvalue = [8, 1]'),
Text(108.04, 108.95428571428571, 'X[2] <= 0.06\nngini = 0.5\nnsamples = 1\nvalue = [1, 1]'),
Text(89.28, 108.95428571428571, 'gini = 0.0\nnsamples = 1\nvalue = [0, 1]'),
Text(287.04, 108.95428571428571, 'X[2] <= 0.084\ngini = 0.5\nnsamples = 1\nvalue = [0, 1]'),
Text(106.0, 201.9085714285714, 'X[2] <= 0.118\ngini = 0.401\nnsamples = 18\nvalue = [19, 5]'),
Text(148.8, 176.84571428571428, 'X[0] <= 0.019\ngini = 0.5\nnsamples = 2\nvalue = [1, 1]'),
Text(133.92000000000002, 139.78285714285715, 'gini = 0.0\nnsamples = 1\nvalue = [1, 0]'),
Text(103.68, 139.78285714285715, 'X[2] <= 0.234\ngini = 0.375\nnsamples = 16\nvalue = [12, 4]'),
Text(103.44, 77.65714285714284, 'gini = 0.0\nnsamples = 2\nvalue = [2, 0]'),
Text(288.08, 108.95428571428571, 'gini = 0.0\nnsamples = 1\nvalue = [0, 1]'),
Text(252.96, 139.78285714285715, 'X[2] <= 0.234\ngini = 0.444\nnsamples = 9\nvalue = [8, 1]'),
Text(238.08, 108.95428571428571, 'gini = 0.0\nnsamples = 1\nvalue = [0, 1]'),
Text(267.84000000000003, 108.95428571428571, 'X[0] <= 0.019\ngini = 0.375\nnsamples = 8\nvalue = [6, 2]'),
Text(238.08, 77.65714285714284, 'X[1] <= 0.228\ngini = 0.278\nnsamples = 6\nvalue = [5, 1]'),
Text(223.20000000000002, 46.59428571428572, 'X[0] <= 0.0\nnsamples = 3\nvalue = [4, 0]'),
Text(252.96, 46.59428571428572, 'X[2] <= 0.292\ngini = 0.5\nnsamples = 3\nvalue = [1, 1]'),
Text(238.08, 15.531428571428573, 'gini = 0.0\nnsamples = 1\nvalue = [1, 0]'),
Text(267.84000000000003, 15.531428571428573, 'X[2] <= 0.292\ngini = 0.5\nnsamples = 1\nvalue = [0, 1]'),
Text(297.6, 77.65714285714284, 'X[2] <= 0.292\ngini = 0.5\nnsamples = 2\nvalue = [1, 1]'),
Text(282.72, 46.59428571428572, 'gini = 0.0\nnsamples = 1\nvalue = [0, 1]'),
Text(312.48, 46.59428571428572, 'X[1] <= 0.653\ngini = 0.375\nnsamples = 4\nvalue = [1, 3]'),
Text(109.74000000000001, 264.0342857142857, 'gini = 0.0\nnsamples = 3\nvalue = [0, 1]'),
Text(305.04, 357.22285714285715, 'X[1] <= -1.061\ngini = 0.239\nnsamples = 36\nvalue = [5, 31]'),
Text(320.16, 326.15999999999997, 'gini = 0.0\nnsamples = 1\nvalue = [1, 0]'),
Text(319.02, 326.15999999999997, 'X[2] <= 1.384\ngini = 0.292\nnsamples = 35\nvalue = [4, 31]'),
Text(305.04, 295.09714285714284, 'X[2] <= 1.278\ngini = 0.375\nnsamples = 16\nvalue = [4, 12]'),
Text(288.08, 232.97142857142856, 'X[1] <= -0.875\ngini = 0.26\nnsamples = 13\nvalue = [2, 11]'),
Text(238.08, 232.97142857142856, 'X[0] <= 0.019\ngini = 0.5\nnsamples = 2\nvalue = [1, 1]'),
Text(109.74000000000001, 264.0342857142857, 'X[1] <= 0.234\ngini = 0.375\nnsamples = 16\nvalue = [12, 4]'),
Text(252.96, 201.9085714285714, 'gini = 0.0\nnsamples = 1\nvalue = [0, 1]'),
Text(287.04, 232.97142857142856, 'X[2] <= 1.194\ngini = 0.165\nnsamples = 11\nvalue = [1, 10]'),
Text(282.72, 201.9085714285714, 'X[2] <= 0.084\ngini = 0.5\nnsamples = 8\nvalue = [8, 0]'),
Text(297.6, 77.65714285714284, 'X[2] <= 0.292\ngini = 0.5\nnsamples = 2\nvalue = [1, 1]'),
Text(305.04, 357.22285714285715, 'X[1] <= -1.061\ngini = 0.239\nnsamples = 36\nvalue = [5, 31]'),
Text(320.16, 326.15999999999997, 'gini = 0.0\nnsamples = 1\nvalue = [1, 0]'),
Text(319.02, 326.15999999999997, 'X[2] <= 1.384\ngini = 0.292\nnsamples = 35\nvalue = [4, 31]'),
Text(305.04, 295.09714285714284, 'X[2] <= 1.278\ngini = 0.375\nnsamples = 16\nvalue = [4, 12]'),
Text(288.08, 232.97142857142856, 'X[1] <= -0.875\ngini = 0.26\nnsamples = 13\nvalue = [2, 11]'),
Text(238.08, 232.97142857142856, 'X[0] <= 0.019\ngini = 0.5\nnsamples = 2\nvalue = [1, 1]'),
Text(109.74000000000001, 264.0342857142857, 'X[1] <= 0.234\ngini = 0.375\nnsamples = 16\nvalue = [12, 4]'),
Text(252.96, 201.9085714285714, 'gini = 0.0\nnsamples = 1\nvalue = [0, 1]'),
Text(287.04, 232.97142857142856, 'X[2] <= 1.194\ngini = 0.165\nnsamples = 11\nvalue = [1, 10]'),
Text(282.72, 201.9085714285714, 'X[2] <= 0.084\ngini = 0.5\nnsamples = 8\nvalue = [8, 0]'),
Text(297.6, 77.65714285714284, 'X[2] <= 0.292\ngini = 0.5\nnsamples = 2\nvalue = [1, 1]'),
Text(305.04, 357.22285714285715, 'X[1] <= -1.061\ngini = 0.239\nnsamples = 36\nvalue = [5, 31]'),
Text(320.16, 326.15999999999997, 'gini = 0.0\nnsamples = 1\nvalue = [1, 0]'),
Text(319.02, 326.15999999999997, 'X[2] <= 1.384\ngini = 0.292\nnsamples = 35\nvalue = [4, 31]'),
Text(305.04, 295.09714285714284, 'X[2] <= 1.278\ngini = 0.375\nnsamples = 16\nvalue = [4, 12]'),
Text(288.08, 232.97142857142856, 'X[1] <= -0.875\ngini = 0.26\nnsamples = 13\nvalue = [2, 11]'),
Text(238.08, 232.97142857142856, 'X[0] <= 0.019\ngini = 0.5\nnsamples = 2\nvalue = [1, 1]'),
Text(109.74000000000001, 264.0342857142857, 'X[1] <= 0.234\ngini = 0.375\nnsamples = 16\nvalue = [12, 4]'),
Text(252.96, 201.9085714285714, 'gini = 0.0\nnsamples = 1\nvalue = [0, 1]'),
Text(287.04, 232.97142857142856, 'X[2] <= 1.194\ngini = 0.165\nnsamples = 11\nvalue = [1, 10]'),
Text(282.72, 201.9085714285714, 'X[2] <= 0.084\ngini = 0.5\nnsamples = 8\nvalue = [8, 0]'),
Text(297.6, 77.65714285714284, 'X[2] <= 0.292\ngini = 0.5\nnsamples = 2\nvalue = [1, 1]'),
Text(305.04, 357.22285714285715, 'X[1] <= -1.061\ngini = 0.239\nnsamples = 36\nvalue = [5, 31]'),
Text(320.16, 326.15999999999997, 'gini = 0.0\nnsamples = 1\nvalue = [1, 0]'),
Text(319.02, 326.15999999999997, 'X[2] <= 1.384\ngini = 0.292\nnsamples = 35\nvalue = [4, 31]'),
Text(305.04, 295.09714285714284, 'X[2] <= 1.278\ngini = 0.375\nnsamples = 16\nvalue = [4, 12]'),
Text(288.08, 232.97142857142856, 'X[1] <= -0.875\ngini = 0.26\nnsamples = 13\nvalue = [2, 11]'),
Text(238.08, 232.97142857142856, 'X[0] <= 0.019\ngini = 0.5\nnsamples = 2\nvalue = [1, 1]'),
Text(109.74000000000001, 264.0342857142857, 'X[1] <= 0.234\ngini = 0.375\nnsamples = 16\nvalue = [12, 4]'),
Text(252.96, 201.9085714285714, 'gini = 0.0\nnsamples = 1\nvalue = [0, 1]'),
Text(287.04, 232.97142857142856, 'X[2] <= 1.194\ngini = 0.165\nnsamples = 11\nvalue = [1, 10]'),
Text(282.72, 201.9085714285714, 'X[2] <= 0.084\ngini = 0.5\nnsamples = 8\nvalue = [8, 0]'),
Text(297.6, 77.65714285714284, 'X[2] <= 0.292\ngini = 0.5\nnsamples = 2\nvalue = [1, 1]'),
Text(305.04, 357.22285714285715, 'X[1] <= -1.061\ngini = 0.239\nnsamples = 36\nvalue = [5, 31]'),
Text(320.16, 326.15999999999997, 'gini = 0.0\nnsamples = 
```