#### 1.Import datatypes

In [1]: #Task 2:A cloth manufacturing company is interested to know about the segment or

In [2]: import pandas as pd
 from sklearn.preprocessing import LabelEncoder
 from sklearn.metrics import accuracy\_score,confusion\_matrix
 import matplotlib.pyplot as plt #import Libraries

# 2.Import libraries

In [3]: data=pd.read\_csv('Company\_Data.csv') #import dataset
data.head() #Viewing top 5 rows of dataframe

#### Out[3]:

	Sales	CompPrice	Income	Advertising	Population	Price	ShelveLoc	Age	Education	Urban
0	9.50	138	73	11	276	120	Bad	42	17	Yes
1	11.22	111	48	16	260	83	Good	65	10	Yes
2	10.06	113	35	10	269	80	Medium	59	12	Yes
3	7.40	117	100	4	466	97	Medium	55	14	Yes
4	4.15	141	64	3	340	128	Bad	38	13	Yes
4										•

### 3 Data Undestanding

In [4]: data.info() #infomation of all datapoint

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 400 entries, 0 to 399
Data columns (total 11 columns):

#	Column	Non-Null Coun	t Dtype
0	Sales	400 non-null	float64
1	CompPrice	400 non-null	int64
2	Income	400 non-null	int64
3	Advertising	400 non-null	int64
4	Population	400 non-null	int64
5	Price	400 non-null	int64
6	ShelveLoc	400 non-null	object
7	Age	400 non-null	int64
8	Education	400 non-null	int64
9	Urban	400 non-null	object
10	US	400 non-null	object
dtvp	es: float64(1	), int64(7), o	biect(3)

In [5]: data.dtypes #datatypes of each columns

memory usage: 34.5+ KB

Out[5]: Sales float64 CompPrice int64 int64 Advertising int64 int64 Population Price int64 ShelveLoc object Age int64 Education int64 Urban object US object dtype: object

```
In [6]: data.Sales.max(),data.Sales.min() #Find maximum and minimum value in sales colum
 Out[6]: (16.27, 0.0)
 In [7]: data['Sales']=pd.cut(data['Sales'],bins=[1,5,12,17],labels=["low","medium","high"
 In [8]:
         le=LabelEncoder()
         for i in data.columns:
             if data[i].dtypes=="object" or "category":
                 data[i]=le.fit_transform(data[i]) #apply Label encoder in non numeric cd
 In [9]: data.head()
 Out[9]:
            Sales CompPrice
                            Income Advertising Population Price
                                                            ShelveLoc Age
                                                                          Education Urban
          0
               2
                        49
                               51
                                          11
                                                  141
                                                         54
                                                                   0
                                                                       17
                                                                                7
          1
               2
                        22
                               27
                                          16
                                                  129
                                                         18
                                                                   1
                                                                       40
                                                                                0
                                                                                       1
          2
               2
                                                                   2
                                                                                2
                                                                                       1
                        24
                                14
                                          10
                                                  138
                                                                       34
                                                         15
          3
               2
                                           4
                                                  249
                                                         31
                                                                   2
                                                                       30
                                                                                4
                                                                                       1
                        28
                               77
                1
                        52
                               42
                                           3
                                                  178
                                                                   0
                                                                       13
                                                                                3
                                                         62
In [10]: data.dtypes #datatype of each columns
Out[10]: Sales
                        int32
         CompPrice
                        int64
                        int64
         Income
         Advertising
                        int64
         Population
                        int64
         Price
                        int64
                        int32
         ShelveLoc
                        int64
         Age
         Education
                        int64
         Urban
                        int32
         US
                        int32
         dtype: object
In [11]: list(data.columns) #list of all columns name
Out[11]: ['Sales',
           CompPrice',
           'Income',
           'Advertising',
          'Population',
          'Price',
          'ShelveLoc',
          'Age',
          'Education',
          'Urban',
          'US']
"Education": "education", "Urban": "urban", "US": "us"}, inplace=Tr
```

```
Out[13]:
             sales compPrice income advertising population price shelveLoc age education urban
                                                                                           us
          0
                2
                         49
                                 51
                                                                     0
                                                                         17
                                                                                   7
                                           11
                                                    141
                                                           54
          1
                2
                         22
                                 27
                                           16
                                                    129
                                                           18
                                                                     1
                                                                         40
                                                                                   0
                                                                                         1
          2
                2
                         24
                                 14
                                           10
                                                    138
                                                                     2
                                                                         34
                                                                                   2
                                                           15
          3
                2
                         28
                                 77
                                            4
                                                    249
                                                           31
                                                                     2
                                                                         30
                                                                                   4
                                                                                         1
          4
                1
                         52
                                 42
                                            3
                                                    178
                                                           62
                                                                     0
                                                                         13
                                                                                   3
                                                                                         1
                                                                                             (
         4. Data Preparing
In [14]: data['sales'].value_counts() #Show values of each labels
Out[14]: 2
               296
                72
         1
                27
         3
                 5
         Name: sales, dtype: int64
In [15]: x=data.drop(['sales'],axis=1)
         y=data['sales']
                          #Devide data in input and output columns
In [16]: |#Dividing data for training and testing
         from sklearn.model_selection import train_test_split
         x_train,x_test,y_train,y_test=train_test_split(x,y,stratify=y,random_state=12,shu
         5.Build The Model(Gini criteria)
In [17]: #Building Decision Tree Classifier using gini Criteria
In [18]: | from sklearn.tree import DecisionTreeClassifier
         model=DecisionTreeClassifier(criterion='gini', max_depth=7)
         model.fit(x_train,y_train) #Build Model
Out[18]: DecisionTreeClassifier(max depth=7)
In [19]: #predict on train data
         y_train_pred=model.predict(x_train)
In [20]: |#Find out accuracy
         accuracy_score(y_train,y_train_pred)
Out[20]: 0.9633333333333333
In [21]: #Find out confision matrix
         confusion_matrix(y_train,y_train_pred)
Out[21]: array([[ 19,
                         0,
                              1,
                                    0],
                   0,
                            2,
                        52,
                                   0],
                 1,
                         5, 216,
                                   0],
                 0,
                         2,
                                   2]], dtype=int64)
                              0,
In [22]: #predict on test data
         y_test_pred=model.predict(x_test)
```

In [13]: |data.head()

```
In [23]: #Find out accuracy
         accuracy_score(y_test_pred,y_test)
Out[23]: 0.81
In [24]: #confusion_mtrix
         confusion_matrix(y_test,y_test_pred)
Out[24]: array([[ 2, 1, 4,
                             0],
                [ 0, 14, 4,
                              0],
                [ 4, 4, 65,
                              1],
                [ 0,
                     0,
                         1, 0]], dtype=int64)
In [25]: acc = [['Train_accuracy', '96%'], ['Test_accuracy', '81%']]
         acc=pd.DataFrame(acc)
Out[25]:
          0 Train_accuracy 96%
            Test_accuracy 81%
         6 Model Building (entropy criteria)
In [26]: #Build Decision Tree Classifier using Entropy Criteria
In [27]: | model1=DecisionTreeClassifier(criterion='entropy', max_depth=7)
In [28]: model1.fit(x_train,y_train) #fit the model
Out[28]: DecisionTreeClassifier(criterion='entropy', max_depth=7)
In [29]: #predict on train data
         y_pred_train=model.predict(x_train)
In [30]: #Find out accuracy
         accuracy_score(y_train,y_pred_train)
Out[30]: 0.96333333333333334
In [31]: #predict on test data
         y_pred_test=model.predict(x_test)
In [32]: #Find out accuracy
         accuracy_score(y_test,y_pred_test)
Out[32]: 0.81
In [33]: acc = [['Train_accuracy', '96%'], ['Test_accuracy', '81%']]
         acc=pd.DataFrame(acc)
         acc
Out[33]:
                           1
          0 Train_accuracy 96%
             Test_accuracy 81%
```

#### 7.Plot the tree

```
In [34]: #plot the tree
         from sklearn import tree
In [35]: fn=['compPrice', 'income', 'advertising', 'population', 'price', 'shelveLoc', 'age', 'ed
         cn=["0","1","2","3"] #LabeLs
In [36]: plt.figure(figsize=(30,10)) #Size of Tree
         tree.plot_tree(model,max_depth=3,filled=True,feature_names=fn,class_names=cn) #
         plt.show()
         8. Model Building (Random Forest(Entropy Criteria))
In [37]: ##Building Random Forest Classifier using Entropy Criteria.
In [38]: | from sklearn.ensemble import RandomForestClassifier
         rf=RandomForestClassifier(n_estimators=15,criterion='entropy',max_depth=7)
         rf.fit(x_train,y_train)
                                   #fit the model
Out[38]: RandomForestClassifier(criterion='entropy', max_depth=7, n_estimators=15)
In [39]: #Predict on train Data
         y_pred1=rf.predict(x_train)
In [40]: # Find OAccuracy
         accuracy_score(y_pred1,y_train)
Out[40]: 0.95333333333333334
In [41]: #Predict on test Data
         y_pred2=rf.predict(x_test)
In [42]: #Accuracy
         accuracy_score(y_pred2,y_test)
Out[42]: 0.78
        acc = [['Train_accuracy', '95%'], ['Test_accuracy', '78%']]
In [43]:
         acc=pd.DataFrame(acc)
         acc
Out[43]:
          0 Train accuracy 95%
             Test_accuracy 78%
```

## 9.Model Building (Random Forest(Gini Criteria))

```
In [44]: ##Building Random Forest Classifier using Gini Criteria.
In [45]: from sklearn.ensemble import RandomForestClassifier
         rf1=RandomForestClassifier(n_estimators=15,criterion='gini',max_depth=10)
         rf1.fit(x_train,y_train)
                                    #fit the model
Out[45]: RandomForestClassifier(max_depth=10, n_estimators=15)
In [46]: #Predict on train Data
         y_pred1=rf1.predict(x_train)
In [47]: # Find OAccuracy
         accuracy_score(y_pred1,y_train)
Out[47]: 0.99
In [48]: #Predict on test Data
         y_pred2=rf1.predict(x_test)
In [49]: #Accuracy
         accuracy_score(y_pred2,y_test)
Out[49]: 0.76
In [50]: acc = [['Train_accuracy', '99%'], ['Test_accuracy', '76%']]
         acc=pd.DataFrame(acc)
Out[50]:
          0 Train_accuracy 99%
            Test_accuracy 76%
```