```
In [22]: import pandas as pd
    import numpy as np
    import matplotlib.pyplot as plt
    import seaborn as sns
    %matplotlib inline#for encoding
    from sklearn.preprocessing import LabelEncoder
    from sklearn.model_selection import train_test_split
    from sklearn.tree import DecisionTreeClassifier
    from sklearn.metrics import classification_report, confusion_matrix
    from sklearn.tree import plot_tree

UsageError: unrecognized arguments: encoding

In [23]: df = pd.read_csv('Company_Data.csv')
    df
```

Out[23]:		Sales	CompPrice	Income	Advertising	Population	Price	ShelveLoc	Age	Education	Urbar
	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
	395	12.57	138	108	17	203	128	Good	33	14	Yes
	396	6.14	139	23	3	37	120	Medium	55	11	Nc
	397	7.41	162	26	12	368	159	Medium	40	18	Yes
	398	5.94	100	79	7	284	95	Bad	50	12	Yes
	399	9.71	134	37	0	27	120	Good	49	16	Yes

400 rows × 11 columns

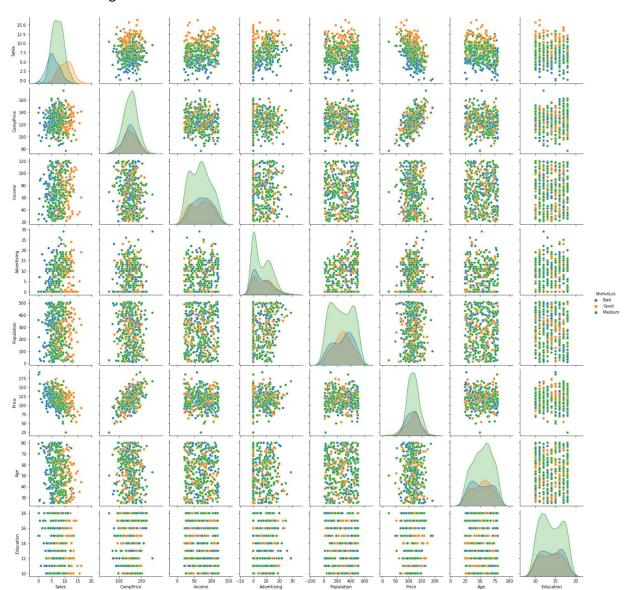
In [24]: df.head()

Out[24]:		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										.

```
In [25]: df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 400 entries, 0 to 399
         Data columns (total 11 columns):
                            Non-Null Count Dtype
              Column
         ---
                                            ----
          0
              Sales
                            400 non-null
                                            float64
          1
                            400 non-null
                                            int64
              CompPrice
          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
                            400 non-null
                                            int64
              Age
          8
              Education
                                            int64
                            400 non-null
          9
              Urban
                            400 non-null
                                            object
          10 US
                            400 non-null
                                            object
         dtypes: float64(1), int64(7), object(3)
         memory usage: 34.5+ KB
In [26]: df.shape
Out[26]: (400, 11)
In [27]: df.isnull().any()
Out[27]: Sales
                         False
         CompPrice
                         False
         Income
                         False
         Advertising
                         False
         Population
                         False
         Price
                         False
         ShelveLoc
                         False
         Age
                         False
         Education
                         False
         Urban
                         False
         US
                         False
         dtype: bool
```

In [28]: # let's plot pair plot to visualise the attributes all at once
sns.pairplot(data=df, hue = 'ShelveLoc')

Out[28]: <seaborn.axisgrid.PairGrid at 0x252999a1ca0>



```
In [29]: #Creating dummy vairables dropping first dummy variable
         df=pd.get_dummies(df,columns=['Urban','US'], drop_first=True)
In [30]: print(df.head())
            Sales CompPrice Income Advertising Population Price ShelveLoc
                                                                                Age
             9.50
         0
                         138
                                  73
                                               11
                                                          276
                                                                  120
                                                                            Bad
                                                                                  42
           11.22
         1
                         111
                                  48
                                               16
                                                          260
                                                                  83
                                                                           Good
                                                                                  65
         2 10.06
                         113
                                  35
                                               10
                                                          269
                                                                  80
                                                                        Medium
                                                                                  59
             7.40
                                                                        Medium
         3
                         117
                                 100
                                                4
                                                          466
                                                                  97
                                                                                  55
            4.15
                         141
                                  64
                                                3
                                                          340
                                                                  128
                                                                           Bad
                                                                                  38
                       Urban_Yes US_Yes
            Education
         0
                   17
                               1
                                       1
         1
                               1
                   10
                                       1
         2
                   12
                               1
                                       1
         3
                   14
                               1
                                       1
         4
                   13
                               1
                                       0
In [31]: | df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 400 entries, 0 to 399
         Data columns (total 11 columns):
          #
              Column
                           Non-Null Count
                                           Dtype
              -----
                           -----
              Sales
                           400 non-null
                                           float64
          0
                           400 non-null
          1
              CompPrice
                                           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 Yes
                           400 non-null
                                           uint8
          10 US_Yes
                           400 non-null
                                           uint8
         dtypes: float64(1), int64(7), object(1), uint8(2)
         memory usage: 29.0+ KB
In [32]: from sklearn.metrics import f1_score
         from sklearn.model_selection import train_test_split
In [33]: df['ShelveLoc']=df['ShelveLoc'].map({'Good':1, 'Medium':2, 'Bad':3})
```

```
In [34]: print(df.head())
                                                    Population
            Sales CompPrice Income Advertising
                                                                 Price ShelveLoc
                                                                                   Age \
             9.50
                                   73
                                                                   120
                                                                                    42
         0
                          138
                                                11
                                                            276
                                                                                3
         1
            11.22
                          111
                                   48
                                                16
                                                            260
                                                                    83
                                                                                1
                                                                                    65
         2
            10.06
                          113
                                   35
                                                10
                                                            269
                                                                    80
                                                                                2
                                                                                    59
                                                                                2
                                                                                    55
         3
             7.40
                          117
                                  100
                                                 4
                                                            466
                                                                    97
         4
             4.15
                          141
                                   64
                                                 3
                                                            340
                                                                   128
                                                                                3
                                                                                    38
             Education Urban_Yes US_Yes
         0
                    17
                                1
                                        1
         1
                    10
                                1
                                        1
         2
                    12
                                1
                                        1
         3
                    14
                                1
                                        1
         4
                    13
                                1
                                        0
```

In [35]: x=df.iloc[:,0:6]
y=df['ShelveLoc']

In [36]: x

Out[36]:

		Sales	CompPrice	Income	Advertising	Population	Price
_	0	9.50	138	73	11	276	120
	1	11.22	111	48	16	260	83
	2	10.06	113	35	10	269	80
	3	7.40	117	100	4	466	97
	4	4.15	141	64	3	340	128
	395	12.57	138	108	17	203	128
	396	6.14	139	23	3	37	120
	397	7.41	162	26	12	368	159
	398	5.94	100	79	7	284	95
	399	9.71	134	37	0	27	120

400 rows × 6 columns

```
In [37]: y
Out[37]: 0
                 3
                 1
         1
         2
                 2
         3
                 2
         4
                 3
                . .
         395
                 1
         396
                 2
         397
                 2
         398
                 3
         399
         Name: ShelveLoc, Length: 400, dtype: int64
In [38]: |df['ShelveLoc'].unique()
Out[38]: array([3, 1, 2], dtype=int64)
In [39]: df.ShelveLoc.value_counts()
Out[39]: 2
               219
                96
                85
         Name: ShelveLoc, dtype: int64
         colnames = list(df.columns)
In [40]:
         colnames
Out[40]: ['Sales',
           'CompPrice',
           'Income',
           'Advertising',
           'Population',
           'Price',
           'ShelveLoc',
           'Age',
           'Education',
           'Urban_Yes',
           'US_Yes']
In [41]: | # Splitting data into training and testing data set
         x_train, x_test,y_train,y_test = train_test_split(x,y, test_size=0.2,random_state
```

Building Decision Tree Classifier using Entropy Criteria

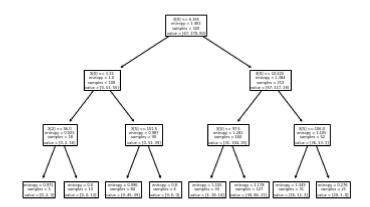
```
In [43]: from sklearn.tree import DecisionTreeClassifier
```

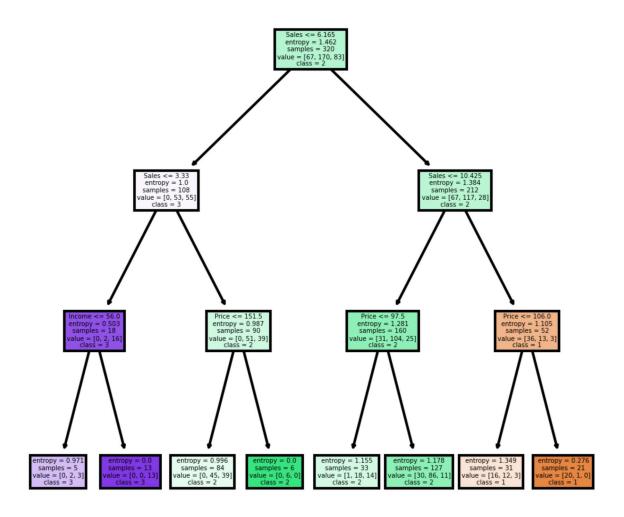
```
In [44]: model = DecisionTreeClassifier(criterion = 'entropy',max_depth=3)
model.fit(x_train,y_train)

Out[44]: DecisionTreeClassifier(criterion='entropy', max_depth=3)

In [45]: from sklearn import tree

In [47]: #PLot the decision tree
tree.plot_tree(model);
```





```
In [49]: #Predicting on test data
         preds = model.predict(x_test) # predicting on test data set
         pd.Series(preds).value_counts() # getting the count of each category
Out[49]: 2
              63
              13
         dtype: int64
In [50]: preds
Out[50]: array([2, 2, 2, 2, 2, 2, 1, 2, 1, 2, 2, 2, 2, 1, 2, 2, 2, 2, 2, 2, 1,
                2, 2, 2, 2, 2, 2, 1, 1, 1, 2, 1, 2, 2, 1, 2, 2, 2, 2, 2, 2, 3, 2,
                2, 2, 2, 2, 2, 3, 2, 2, 2, 1, 1, 2, 2, 3, 2, 2, 1, 2, 2, 2,
                2, 2, 2, 1, 2, 2, 2, 2, 2, 3, 2, 2, 2], dtype=int64)
In [51]: pd.crosstab(y_test,preds) # getting the 2 way table to understand the correct and
Out[51]:
              col_0 1 2 3
          ShelveLoc
                 1 8 10 0
                 2 5 41 3
                 3 0 12 1
In [52]: # Accuracy
         np.mean(preds==y_test)
Out[52]: 0.625
```

Building Decision Tree Classifier (CART) using Gini Criteria

```
In [53]: from sklearn.tree import DecisionTreeClassifier
    model_gini = DecisionTreeClassifier(criterion='gini', max_depth=3)

In [54]: model_gini.fit(x_train, y_train)

Out[54]: DecisionTreeClassifier(max_depth=3)

In [55]: #Prediction and computing the accuracy
    pred=model.predict(x_test)
    np.mean(preds==y_test)

Out[55]: 0.625
```

Decision Tree Regression Example