

```
In [33]: import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, confusion_matrix
from sklearn.preprocessing import LabelEncoder
from sklearn.ensemble import RandomForestClassifier
from sklearn.ensemble import GradientBoostingClassifier
import warnings
warnings.filterwarnings('ignore')
```

```
In [3]: cm_data = pd.read_csv('Company_Data.csv')
cm_data
```

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
...
395	12.57	138	108	17	203	128	Good	33	14	Yes
396	6.14	139	23	3	37	120	Medium	55	11	No
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 [7]: cm_data.loc[cm_data['Sales'] >=8, 'sales'] = 'High'
cm_data.loc[cm_data['Sales'] <8, 'sales'] = 'Low'
```

```
In [8]: cm_data.drop('Sales', axis = 1, inplace = True)
```

```
In [10]: cm_data.shape
```

Out[10]: (400, 11)

```
In [11]: cm_data.isna().sum()
```

```
Out[11]: CompPrice      0
         Income        0
         Advertising    0
         Population     0
         Price          0
         ShelfLoc       0
         Age            0
         Education      0
         Urban          0
         US             0
         sales          0
         dtype: int64
```

```
In [13]: cm_data.dtypes
```

```
Out[13]: CompPrice      int64
         Income        int64
         Advertising    int64
         Population     int64
         Price          int64
         ShelfLoc       object
         Age            int64
         Education      int64
         Urban          object
         US             object
         sales          object
         dtype: object
```

```
In [14]: le = LabelEncoder()
         cm_data['ShelveLoc'] = le.fit_transform(cm_data['ShelveLoc'])
         cm_data['ShelveLoc'].unique()
```

```
Out[14]: array([0, 1, 2])
```

```
In [15]: cm_data['Urban'] = le.fit_transform(cm_data['Urban'])
         cm_data['Urban'].unique()
```

```
Out[15]: array([1, 0])
```

```
In [16]: cm_data['US'] = le.fit_transform(cm_data['US'])
         cm_data['US'].unique()
```

```
Out[16]: array([1, 0])
```

In [17]: `cm_data.head()`

Out[17]:

	CompPrice	Income	Advertising	Population	Price	ShelveLoc	Age	Education	Urban	US	sa
0	138	73	11	276	120	0	42	17	1	1	H
1	111	48	16	260	83	1	65	10	1	1	H
2	113	35	10	269	80	2	59	12	1	1	H
3	117	100	4	466	97	2	55	14	1	1	L
4	141	64	3	340	128	0	38	13	1	0	L

In [18]: `x = cm_data.drop('sales',axis=1)`
`y = cm_data[['sales']]`

In [19]: `x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.20,random_state=`

In [20]: `x_train.shape,y_train.shape`

Out[20]: ((320, 10), (320, 1))

In [21]: `x_test.shape,y_test.shape`

Out[21]: ((80, 10), (80, 1))

In [24]: `rf = RandomForestClassifier()`

In [25]: `rf.fit(x_train,y_train)`

Out[25]: `RandomForestClassifier()`

In [26]: `y_pred_train = rf.predict(x_train)`

In [27]: `print('Accuracy score :',accuracy_score(y_train,y_pred_train))`

Accuracy score : 1.0

In [28]: `print('Confusion Metric :\n',confusion_matrix(y_train,y_pred_train))`

Confusion Metric :
 [[133 0]
 [0 187]]

In [29]: `y_pred_test = rf.predict(x_test)`

```
In [30]: print('Accuracy score :', accuracy_score(y_test,y_pred_test))
```

Accuracy score : 0.75

```
In [32]: print('Confusion Metric :\n',confusion_matrix(y_test,y_pred_test))
```

Confusion Metric :

[[16 15]

[5 44]]

```
In [35]: gb = GradientBoostingClassifier()  
gb.fit(x_train,y_train)
```

```
Out[35]: GradientBoostingClassifier()
```

```
In [36]: y_pred = gb.predict(x_test)
```

```
In [37]: print('Accuracy score :', accuracy_score(y_test,y_pred))
```

Accuracy score : 0.775

```
In [38]: print('Confusion Metric :\n',confusion_matrix(y_test,y_pred))
```

Confusion Metric :

[[20 11]

[7 42]]

```
In [ ]:
```