In [1]:
```
import pandas as pd
import numpy as np
```

In [3]:
```
d=pd.read_csv('https://github.com/YBI-Foundation/Dataset/raw/main/Big%20Sales%20D
```

In [4]:
```
d.head()
```

Out[4]:

	Item_Identifier	Item_Weight	Item_Fat_Content	Item_Visibility	Item_Type	Item_MRP	Outlet_Iden
0	FDT36	12.3	Low Fat	0.111448	Baking Goods	33.4874	OU
1	FDT36	12.3	Low Fat	0.111904	Baking Goods	33.9874	OU
2	FDT36	12.3	LF	0.111728	Baking Goods	33.9874	OU
3	FDT36	12.3	Low Fat	0.000000	Baking Goods	34.3874	OU
4	FDP12	9.8	Regular	0.045523	Baking Goods	35.0874	OU

In [5]:
```
d.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14204 entries, 0 to 14203
Data columns (total 12 columns):
 #   Column                     Non-Null Count  Dtype
---  ------                     --------------  -----
 0   Item_Identifier            14204 non-null  object
 1   Item_Weight                11815 non-null  float64
 2   Item_Fat_Content           14204 non-null  object
 3   Item_Visibility            14204 non-null  float64
 4   Item_Type                  14204 non-null  object
 5   Item_MRP                   14204 non-null  float64
 6   Outlet_Identifier          14204 non-null  object
 7   Outlet_Establishment_Year  14204 non-null  int64
 8   Outlet_Size                14204 non-null  object
 9   Outlet_Location_Type       14204 non-null  object
 10  Outlet_Type                14204 non-null  object
 11  Item_Outlet_Sales          14204 non-null  float64
dtypes: float64(4), int64(1), object(7)
memory usage: 1.3+ MB
```

In [6]:
```
d.columns
```

Out[6]:
```
Index(['Item_Identifier', 'Item_Weight', 'Item_Fat_Content', 'Item_Visibility',
       'Item_Type', 'Item_MRP', 'Outlet_Identifier',
       'Outlet_Establishment_Year', 'Outlet_Size', 'Outlet_Location_Type',
       'Outlet_Type', 'Item_Outlet_Sales'],
      dtype='object')
```

In [7]:
```
d.describe()
```

Out[7]:

	Item_Weight	Item_Visibility	Item_MRP	Outlet_Establishment_Year	Item_Outlet_Sales
count	11815.000000	14204.000000	14204.000000	14204.000000	14204.000000
mean	12.788355	0.065953	141.004977	1997.830681	2185.836320
std	4.654126	0.051459	62.086938	8.371664	1827.479550
min	4.555000	0.000000	31.290000	1985.000000	33.290000
25%	8.710000	0.027036	94.012000	1987.000000	922.135101
50%	12.500000	0.054021	142.247000	1999.000000	1768.287680
75%	16.750000	0.094037	185.855600	2004.000000	2988.110400
max	30.000000	0.328391	266.888400	2009.000000	31224.726950

In [13]:
```
d['Item_Weight'].fillna(d.groupby(['Item_Type'])['Item_Weight'].transform('mean')
```

In [15]:
```
d.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14204 entries, 0 to 14203
Data columns (total 12 columns):
 #   Column                     Non-Null Count  Dtype
---  ------                     --------------  -----
 0   Item_Identifier            14204 non-null  object
 1   Item_Weight                14204 non-null  float64
 2   Item_Fat_Content           14204 non-null  object
 3   Item_Visibility            14204 non-null  float64
 4   Item_Type                  14204 non-null  object
 5   Item_MRP                   14204 non-null  float64
 6   Outlet_Identifier          14204 non-null  object
 7   Outlet_Establishment_Year  14204 non-null  int64
 8   Outlet_Size                14204 non-null  object
 9   Outlet_Location_Type       14204 non-null  object
 10  Outlet_Type                14204 non-null  object
 11  Item_Outlet_Sales          14204 non-null  float64
dtypes: float64(4), int64(1), object(7)
memory usage: 1.3+ MB
```

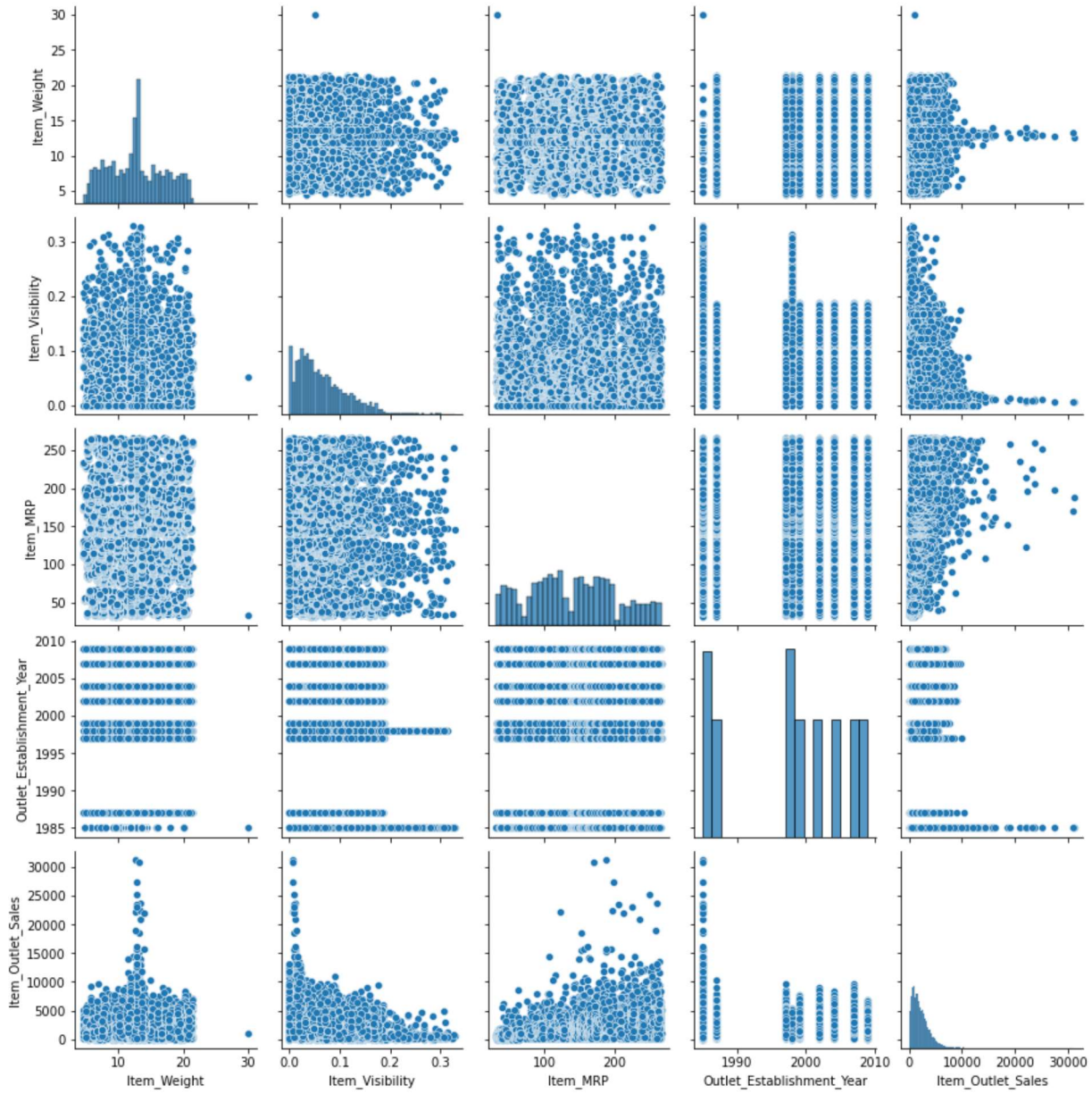
In [16]:
```
d.describe()
```

Out[16]:

	Item_Weight	Item_Visibility	Item_MRP	Outlet_Establishment_Year	Item_Outlet_Sales
count	14204.000000	14204.000000	14204.000000	14204.000000	14204.000000
mean	12.790642	0.065953	141.004977	1997.830681	2185.836320
std	4.251186	0.051459	62.086938	8.371664	1827.479550
min	4.555000	0.000000	31.290000	1985.000000	33.290000
25%	9.300000	0.027036	94.012000	1987.000000	922.135101
50%	12.800000	0.054021	142.247000	1999.000000	1768.287680
75%	16.000000	0.094037	185.855600	2004.000000	2988.110400
max	30.000000	0.328391	266.888400	2009.000000	31224.726950

In [17]:
```
import seaborn as sns
sns.pairplot(d)
```

Out[17]: <seaborn.axisgrid.PairGrid at 0x1e51e40eee0>



Categories and counts of categorical variables

In [18]:
```
d[['Item_Identifier']].value_counts()
```

Out[18]:
```
Item_Identifier
FDQ08              10
FDO24              10
FDQ19              10
FDQ28              10
FDQ31              10
                   ..
FDM52               7
FDM50               7
FDL50               7
FDM10               7
FDR51               7
Length: 1559, dtype: int64
```

In [19]:
```
d['Item_Fat_Content'].value_counts()
```

Out[19]:
```
Low Fat    8485
Regular    4824
LF          522
reg         195
low fat     178
Name: Item_Fat_Content, dtype: int64
```

In [22]:
```
d.replace({'Item_Fat_Content':{'LF':'Low Fat','reg':'Regular','low fat':'Low Fat'
```

In [23]:
```
d[['Item_Fat_Content']].value_counts()
```

Out[23]:
```
Item_Fat_Content
Low Fat             9185
Regular             5019
dtype: int64
```

In [24]:
```
d.replace({'Item_Fat_Content':{'Low Fat':0,'Regular':1}},inplace=True)
```

In [25]:
```
d['Item_Type'].value_counts()
```

Out[25]:
```
Fruits and Vegetables    2013
Snack Foods              1989
Household                1548
Frozen Foods             1426
Dairy                    1136
Baking Goods             1086
Canned                   1084
Health and Hygiene        858
Meat                      736
Soft Drinks               726
Breads                    416
Hard Drinks               362
Others                    280
Starchy Foods             269
Breakfast                 186
Seafood                    89
Name: Item_Type, dtype: int64
```

In [28]:
```
d.replace({'Item_Type':{'Fruits and Vegetables':0,'Snack Foods':0,'Household':1,'
```

In [29]:
```
d['Item_Type'].value_counts()
```

Out[29]:
```
0    11518
1     2406
2      280
Name: Item_Type, dtype: int64
```

In [30]:
```
d['Outlet_Identifier'].value_counts()
```

Out[30]:
```
OUT027    1559
OUT013    1553
OUT049    1550
OUT046    1550
OUT035    1550
OUT045    1548
OUT018    1546
OUT017    1543
OUT010     925
OUT019     880
Name: Outlet_Identifier, dtype: int64
```

In [34]:
```
['OUT027':0,'OUT013':1,'OUT049':2,'OUT046':3,'OUT035':4,'OUT045':5,'OUT018':6,'OUT
```

In [36]:
```
d['Outlet_Identifier'].value_counts()
```

Out[36]:
```
0    1559
1    1553
2    1550
3    1550
4    1550
5    1548
6    1546
7    1543
8     925
9     880
Name: Outlet_Identifier, dtype: int64
```

In [38]:
```
d['Outlet_Size'].value_counts()
```

Out[38]:
```
Medium    7122
Small     5529
High      1553
Name: Outlet_Size, dtype: int64
```

In [39]:
```
d.replace({'Outlet_Size':{'Small':0,'Medium':1,'High':2}},inplace=True)
```

In [41]:
```
d['Outlet_Size'].value_counts()
```

Out[41]:
```
1    7122
0    5529
2    1553
Name: Outlet_Size, dtype: int64
```

In [42]:
```
d['Outlet_Location_Type'].value_counts()
```

Out[42]:
```
Tier 3    5583
Tier 2    4641
Tier 1    3980
Name: Outlet_Location_Type, dtype: int64
```

In [43]:
```
d.replace({'Outlet_Location_Type':{'Tier 1':0,'Tier 2':1,'Tier 3':2}},inplace=Tru
```

In [44]:
```
d['Outlet_Location_Type'].value_counts()
```

Out[44]:
```
2    5583
1    4641
0    3980
Name: Outlet_Location_Type, dtype: int64
```

In [45]:
```
d['Outlet_Type'].value_counts()
```

Out[45]:
```
Supermarket Type1    9294
Grocery Store        1805
Supermarket Type3    1559
Supermarket Type2    1546
Name: Outlet_Type, dtype: int64
```

In [99]:
```
d.replace({'Outlet_Type':{'Grocery Store':0,'Supermarket Type1':1,'Supermarket Ty
```

In [100]:
```
d['Outlet_Type'].value_counts()
```

Out[100]:
```
1    9294
3    3105
0    1805
Name: Outlet_Type, dtype: int64
```

In [101]:
```
d.head()
```

Out[101]:

	Item_Identifier	Item_Weight	Item_Fat_Content	Item_Visibility	Item_Type	Item_MRP	Outlet_Iden
0	FDT36	12.3	0	0.111448	0	33.4874	
1	FDT36	12.3	0	0.111904	0	33.9874	
2	FDT36	12.3	0	0.111728	0	33.9874	
3	FDT36	12.3	0	0.000000	0	34.3874	
4	FDP12	9.8	1	0.045523	0	35.0874	

In [102]:
```
d.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14204 entries, 0 to 14203
Data columns (total 12 columns):
 #   Column                     Non-Null Count  Dtype
---  ------                     --------------  -----
 0   Item_Identifier            14204 non-null  object
 1   Item_Weight                14204 non-null  float64
 2   Item_Fat_Content           14204 non-null  int64
 3   Item_Visibility            14204 non-null  float64
 4   Item_Type                  14204 non-null  int64
 5   Item_MRP                   14204 non-null  float64
 6   Outlet_Identifier          14204 non-null  int64
 7   Outlet_Establishment_Year  14204 non-null  int64
 8   Outlet_Size                14204 non-null  int64
 9   Outlet_Location_Type       14204 non-null  int64
 10  Outlet_Type                14204 non-null  int64
 11  Item_Outlet_Sales          14204 non-null  float64
dtypes: float64(4), int64(7), object(1)
memory usage: 1.3+ MB
```

In [103]:
```
d.shape
```

Out[103]:
```
(14204, 12)
```

In [104]:
```
y=d['Item_Outlet_Sales']
```

In [105]:
```
y.shape
```

Out[105]:
```
(14204,)
```

In [106]:
```
y
```

Out[106]:
```
0         436.608721
1         443.127721
2         564.598400
3        1719.370000
4         352.874000
              ...
14199    4984.178800
14200    2885.577200
14201    2885.577200
14202    3803.676434
14203    3644.354765
Name: Item_Outlet_Sales, Length: 14204, dtype: float64
```

In [107]:
```
Type','Item_MRP','Outlet_Identifier','Outlet_Establishment_Year','Outlet_Size','O
```

In [108]:
```
x=d.drop(['Item_Identifier','Item_Outlet_Sales'],axis=1)
```

In [109]:
```
x.shape
```

Out[109]:
```
(14204, 10)
```

In [110]:
```
x
```

Out[110]:

	Item_Weight	Item_Fat_Content	Item_Visibility	Item_Type	Item_MRP	Outlet_Identifier	Outle
0	12.300000	0	0.111448	0	33.4874	2	
1	12.300000	0	0.111904	0	33.9874	7	
2	12.300000	0	0.111728	0	33.9874	6	
3	12.300000	0	0.000000	0	34.3874	9	
4	9.800000	1	0.045523	0	35.0874	7	
...	...	...	...	...	...	...	
14199	12.800000	0	0.069606	0	261.9252	4	
14200	12.800000	0	0.070013	0	262.8252	7	
14201	12.800000	0	0.069561	0	263.0252	1	
14202	13.659758	0	0.069282	0	263.5252	0	
14203	12.800000	0	0.069727	0	263.6252	2	

14204 rows × 10 columns

Get X Variables Standardized

In [111]:
```
from sklearn.preprocessing import StandardScaler
```

In [112]:
```
s=StandardScaler()
```

In [113]:
```
x_std=d[['Item_Weight','Item_Visibility','Item_MRP','Outlet_Establishment_Year']]
```

In [114]:
```
x_std=s.fit_transform(x_std)
```

In [115]:
```
x_std
```

Out[115]:
```
array([[-0.11541705,  0.88413635, -1.73178716,  0.13968068],
       [-0.11541705,  0.89300616, -1.72373366,  1.09531886],
       [-0.11541705,  0.88958331, -1.72373366,  1.3342284 ],
       ...,
       [ 0.00220132,  0.07011952,  1.96538148, -1.29377659],
       [ 0.20444792,  0.06469366,  1.97343499, -1.53268614],
       [ 0.00220132,  0.07334891,  1.97504569,  0.13968068]])
```

In [116]:
```
ishment_Year']] = pd.DataFrame(x_std, columns=['Item_Weight','Item_Visibility','It
```

In [117]:
```
x
```

Out[117]:

	Item_Weight	Item_Fat_Content	Item_Visibility	Item_Type	Item_MRP	Outlet_Identifier	Outle
0	-0.115417	0	0.884136	0	-1.731787	2	
1	-0.115417	0	0.893006	0	-1.723734	7	
2	-0.115417	0	0.889583	0	-1.723734	6	
3	-0.115417	0	-1.281712	0	-1.717291	9	
4	-0.703509	1	-0.397031	0	-1.706016	7	
...	...	...	...	...	...	...	
14199	0.002201	0	0.070990	0	1.947664	4	
14200	0.002201	0	0.078898	0	1.962160	7	
14201	0.002201	0	0.070120	0	1.965381	1	
14202	0.204448	0	0.064694	0	1.973435	0	
14203	0.002201	0	0.073349	0	1.975046	2	

14204 rows × 10 columns

Train Test Split

In [118]:
```
from sklearn.model_selection import  train_test_split
```

In [119]:
```
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.1,random_state=252
```

In [120]:
```
x_train.shape,x_test.shape,y_train.shape,y_test.shape
```

Out[120]:
```
((12783, 10), (1421, 10), (12783,), (1421,))
```

Model Train

In [121]:
```
from sklearn.ensemble import RandomForestRegressor
l=RandomForestRegressor(random_state=2529)
```

In [122]:
```
l.fit(x_train,y_train)
```

Out[122]:
```
RandomForestRegressor
RandomForestRegressor(random_state=2529)
```

Model Prediction

In [123]:
```
y_pred=l.predict(x_test)
```

In [124]:
```
y_pred.shape
```

Out[124]:
```
(1421,)
```

In [125]:
```
y_pred
```

Out[125]:
```
array([1459.81352734,  726.65859122, 1917.87747221, ..., 2196.2426537 ,
       3286.85491915,  456.48837472])
```

Model Evaluation

In [126]:
```
from sklearn.metrics import mean_squared_error,mean_absolute_error,r2_score
```

In [127]:
```
mean_squared_error(y_test,y_pred)
```

Out[127]:
```
1625874.2334575457
```

In [128]:
```
mean_absolute_error(y_test,y_pred)
```

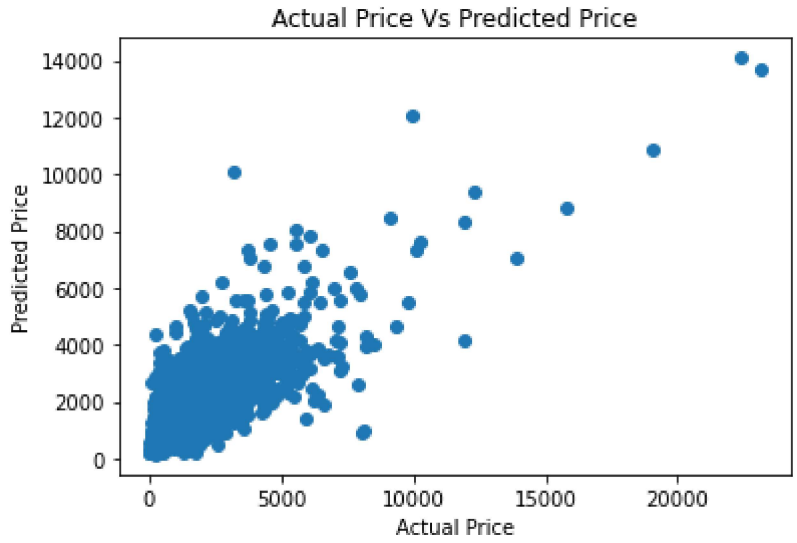
Out[128]:
```
829.99262104014
```

In [129]:
```
r2_score(y_test,y_pred)
```

Out[129]:
```
0.5768090767897451
```

Visualisation of Actual and Predicted Results

In [130]:
```
import matplotlib.pyplot as p
p.scatter(y_test,y_pred)
p.xlabel('Actual Price')
p.ylabel('Predicted Price')
p.title("Actual Price Vs Predicted Price")
p.show()
```



In [ ]: