# **Sales Prediction using Random Forest**

# **Import Libraries**

#### In [1]:

import pandas as pd
import numpy as np

### In [2]:

df = pd.read\_csv(r'https://raw.githubusercontent.com/YBI-Foundation/Dataset/main/Big%20Sale

## In [3]:

df.head()

#### Out[3]:

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

#### In [4]:

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

### **Get columns**

#### In [5]:

```
df.columns
```

```
Out[5]:
```

#### In [6]:

```
df.describe()
```

#### Out[6]:

	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
4					<b>•</b>

### In [7]:

df['Item\_Weight'].fillna(df.groupby(['Item\_Type'])['Item\_Weight'].transform('mean'), inplac

#### In [8]:

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14204 entries, 0 to 14203
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	<pre>Item_Identifier</pre>	14204 non-null	object
1	Item_Weight	14204 non-null	float64
2	<pre>Item_Fat_Content</pre>	14204 non-null	object
3	<pre>Item_Visibility</pre>	14204 non-null	float64
4	<pre>Item_Type</pre>	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	<pre>Item_Outlet_Sales</pre>	14204 non-null	float64

dtypes: float64(4), int64(1), object(7)

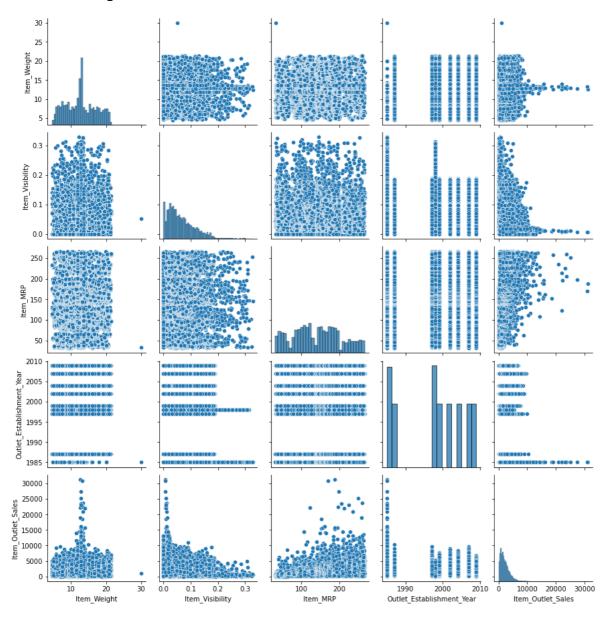
memory usage: 1.3+ MB

# In [9]:

import seaborn as sns
sns.pairplot(df)

# Out[9]:

<seaborn.axisgrid.PairGrid at 0x25601e2af40>



```
In [10]:
df[['Item_Identifier']].value_counts()
Out[10]:
Item_Identifier
FDQ08
                   10
FD024
                   10
FDQ19
                    10
FDQ28
                   10
FDQ31
                   10
FDM52
                    7
                     7
FDM50
                     7
FDL50
                     7
FDM10
FDR51
Length: 1559, dtype: int64
In [11]:
df.replace({'Item_Fat_Content': {'LF': 'Low Fat', 'reg': 'Regular','low fat':'Low Fat'}}, i
In [12]:
df[['Item_Fat_Content']].value_counts()
Out[12]:
Item_Fat_Content
Low Fat
                     9185
                     5019
Regular
dtype: int64
In [13]:
df.replace({'Item_Fat_Content': {'Low Fat' : 0, 'Regular' : 1}}, inplace = True)
```

```
In [14]:
```

```
df[['Item_Type']].value_counts()
Out[14]:
Item_Type
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
dtype: int64
In [15]:
ks':0, 'Others' : 2, 'Starchy Foods': 0, 'Breakfast' : 0, 'Seafood' : 0 }}, inplace = True)
In [16]:
df[['Item_Type']].value_counts()
Out[16]:
Item_Type
              11518
0
               2406
1
2
                280
dtype: int64
In [17]:
df[['Outlet_Identifier']].value_counts()
Out[17]:
Outlet_Identifier
OUT027
                      1559
0UT013
                      1553
0UT035
                      1550
0UT046
                      1550
0UT049
                      1550
0UT045
                      1548
0UT018
                      1546
0UT017
                      1543
OUT010
                       925
0UT019
                        880
dtype: int64
```

```
In [18]:
JT046' : 3, 'OUT035' : 4, 'OUT045' : 5, 'OUT018' : 6, 'OUT017' : 7, 'OUT010' : 8, 'OUT019'
In [19]:
df[['Outlet_Identifier']].value_counts()
Out[19]:
Outlet_Identifier
                      1559
1
                      1553
2
                      1550
3
                      1550
4
                      1550
5
                      1548
6
                      1546
7
                      1543
8
                       925
9
                       880
dtype: int64
In [20]:
df[['Outlet_Size']].value_counts()
Out[20]:
Outlet_Size
Medium
                7122
Small
                5529
High
                1553
dtype: int64
In [21]:
df.replace({'Outlet_Size' : {'Small' : 0, 'Medium' : 1, 'High' : 2}}, inplace = True)
In [22]:
df[['Outlet_Size']].value_counts()
Out[22]:
Outlet_Size
                7122
0
                5529
                1553
dtype: int64
```

```
In [23]:
df[['Outlet_Location_Type']].value_counts()
Out[23]:
Outlet_Location_Type
Tier 3
                         5583
Tier 2
                         4641
Tier 1
                         3980
dtype: int64
In [24]:
df.replace({'Outlet_Location_Type': {'Tier 1': 0, 'Tier 2': 1, 'Tier 3': 2}}, inplace = Tru
In [25]:
df[['Outlet_Location_Type']].value_counts()
Out[25]:
Outlet_Location_Type
                         5583
1
                         4641
                         3980
dtype: int64
In [26]:
df[['Outlet_Type']].value_counts()
Out[26]:
Outlet_Type
Supermarket Type1
                      9294
Grocery Store
                      1805
Supermarket Type3
                      1559
Supermarket Type2
                      1546
dtype: int64
In [27]:
utlet_Type' : {'Grocery Store':0, 'Supermarket Type1' : 1, 'Supermarket Type2' : 2, 'Superma
In [28]:
df[['Outlet_Type']].value_counts()
Out[28]:
Outlet Type
                9294
1
0
                1805
3
                1559
                1546
dtype: int64
```

```
In [29]:
```

```
df.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	<pre>Item_Fat_Content</pre>	14204 non-null	int64
3	<pre>Item_Visibility</pre>	14204 non-null	float64
4	<pre>Item_Type</pre>	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	<pre>Item_Outlet_Sales</pre>	14204 non-null	float64
	C1 1 C4 (4) 1 1 C4 (7)	1 • 1/4\	

dtypes: float64(4), int64(7), object(1)

memory usage: 1.3+ MB

# In [30]:

```
df.head()
```

#### Out[30]:

	Item_Identifier	Item_Weight	Item_Fat_Content	Item_Visibility	Item_Type	Item_MRP	Outlet_I
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	
4							•

# **Get Shape of Dataframe**

```
In [31]:
df.shape
```

Out[31]:

(14204, 12)

#### In [32]:

```
y = df['Item_Outlet_Sales']
```

```
In [33]:
y.shape
Out[33]:
(14204,)
In [34]:
У
Out[34]:
          436.608721
0
1
          443.127721
2
          564.598400
3
         1719.370000
4
          352.874000
         4984.178800
14199
14200
         2885.577200
14201
         2885.577200
14202
         3803.676434
14203
         3644.354765
Name: Item_Outlet_Sales, Length: 14204, dtype: float64
In [35]:
df.columns
Out[35]:
Index(['Item_Identifier', 'Item_Weight', 'Item_Fat_Content', 'Item_Visibilit
у',
       'Item_Type', 'Item_MRP', 'Outlet_Identifier',
       'Outlet_Establishment_Year', 'Outlet_Size', 'Outlet_Location_Type',
       'Outlet_Type', 'Item_Outlet_Sales'],
      dtype='object')
In [36]:
X = [['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']]
In [37]:
X = df.drop(['Item_Identifier', 'Item_Outlet_Sales'], axis = 1)
In [38]:
X.shape
Out[38]:
(14204, 10)
```

```
In [39]:
```

Χ

#### Out[39]:

	Item_Weight	Item_Fat_Content	Item_Visibility	Item_Type	Item_MRP	Outlet_Identifier	С	
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 ı	14204 rows × 10 columns							
4							<b>&gt;</b>	

# **Get X Variables Standardized**

```
In [41]:
```

from sklearn.preprocessing import StandardScaler

```
In [42]:
```

```
sc = StandardScaler()
```

```
In [43]:
```

```
X_std = df[['Item_Weight', 'Item_Visibility', 'Item_MRP', 'Outlet_Establishment_Year']]
```

# In [44]:

```
X_std = sc.fit_transform(X_std)
```

#### Out[47]:

	Item_Weight	Item_Fat_Content	Item_Visibility	Item_Type	Item_MRP	Outlet_Identifier	О
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	
	rows × 10 colu	umns					
4							

# **Get train Test Split**

```
In [48]:
```

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

```
In [49]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X,y, test_size = 0.1, random_state = 25
```

```
In [51]:

X_train.shape, X_test.shape, y_train.shape, y_test.shape

Out[51]:

((12783, 10), (1421, 10), (12783,), (1421,))
```

# **Get Model Train**

```
In [52]:
from sklearn.ensemble import RandomForestRegressor

In [53]:
rfr = RandomForestRegressor (random_state = 2539)

In [54]:
rfr.fit(X_train, y_train)
Out[54]:
```

# **Get Model Prediction**

RandomForestRegressor(random\_state=2539)

```
In [55]:

y_pred = rfr.predict(X_test)

In [56]:

y_pred.shape

Out[56]:
(1421,)

In [57]:

y_pred

Out[57]:
array([1428.48491758, 739.39517005, 1764.06852049, ..., 2131.8834375, 3221.29313926, 448.15959596])
```

# **Get Model Evaluation**

```
In [59]:
```

from sklearn.metrics import mean\_squared\_error, mean\_absolute\_error, r2\_score

#### In [60]:

```
mean_squared_error(y_test, y_pred)
```

#### Out[60]:

1617511.846634074

#### In [61]:

```
mean_absolute_error(y_test, y_pred)
```

#### Out[61]:

830.7489828870267

#### In [62]:

```
r2_score(y_test, y_pred)
```

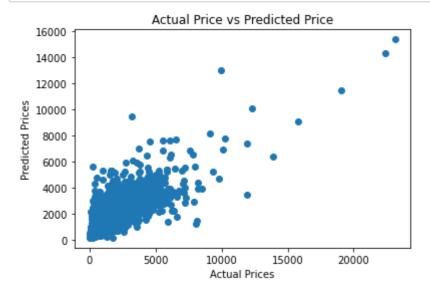
#### Out[62]:

0.5789856819214596

# **Get Visualization of Actual vs Predicted Results**

#### In [64]:

```
import matplotlib.pyplot as plt
plt.scatter(y_test, y_pred)
plt.xlabel("Actual Prices")
plt.ylabel("Predicted Prices")
plt.title("Actual Price vs Predicted Price")
plt.show()
```



In [ ]:			