#import library
import pandas as pd

import numpy as np

#import CSV as dataframe

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

df.head() #first five rows

	Item_Identifier	Item_Weight	<pre>Item_Fat_Content</pre>	<pre>Item_Visibility</pre>	<pre>Item_Type</pre>	Ite
0	FDT36	12.3	Low Fat	0.111448	Baking Goods	3
1	FDT36	12.3	Low Fat	0.111904	Baking Goods	3
2	FDT36	12.3	LF	0.111728	Baking Goods	3
4						•

## df.info() #Get info from datset

memory usage: 1.3+ MB

<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	11815 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
dtyp	es: float64(4), int64(1), o	bject(7)	

## df.columns

df.describe()

	Item_Weight	<pre>Item_Fat_Content</pre>	<pre>Item_Visibility</pre>	<pre>Item_Type</pre>	<pre>Item_MRP</pre>
count	14204.000000	14204.000000	14204.000000	14204.000000	14204.000000
mean	12.790642	0.353351	0.065953	0.208814	141.004977
std	4.251186	0.478027	0.051459	0.452384	62.086938
min	4.555000	0.000000	0.000000	0.000000	31.290000
25%	9.300000	0.000000	0.027036	0.000000	94.012000
50%	12.800000	0.000000	0.054021	0.000000	142.247000
75%	16.000000	1.000000	0.094037	0.000000	185.855600
max	30.000000	1.000000	0.328391	2.000000	266.888400

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

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	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

df.describe()

		Item_Weight	Item_Visibility	<pre>Item_MRP</pre>	Outlet_Establishment_Year	Ite	
	count	14204.000000	14204.000000	14204.000000	14204.000000		
					100= 000001		
df.co	lumns #	get column nar	nes				
	<pre>Index(['Item_Identifier', 'Item_Weight', 'Item_Fat_Content', 'Item_Visibility',</pre>						
	750/	40 000000	0.004007	105 055000	0004 000000		
df.de	scribe(	) #Get summary	/ statistics				

Item\_Weight Item\_Visibility Item\_MRP Outlet\_Establishment\_Year count 14204.000000 14204.000000 14204.000000 14204.000000 12.790642 0.065953 141.004977 1997.830681 mean 62.086938 std 4.251186 0.051459 8.371664 min 4.555000 0.000000 31.290000 1985.000000 25% 9.300000 0.027036 94.012000 1987.000000 50% 12.800000 0.054021 142.247000 1999.000000

185.855600

266.888400

0.094037

0.328391

import seaborn as sns
sns.pairplot(df)

75%

max

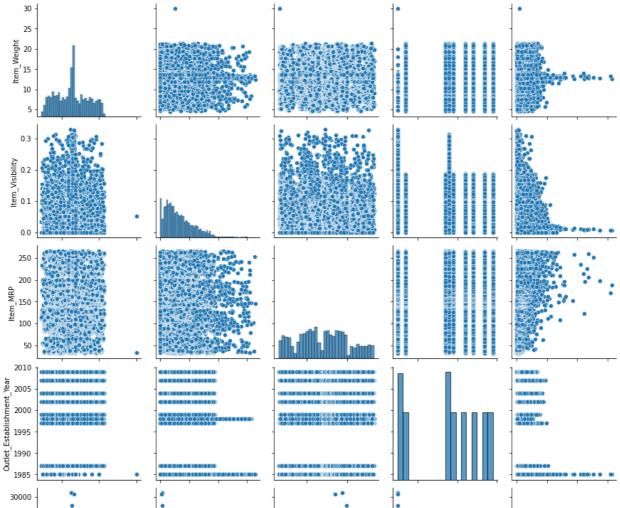
16.000000

30.000000

2004.000000

2009.000000





#get categories and counts of categorical variables
df[['Item\_Identifier']].value\_counts()

Item_Ide	entifi	er		
FDQ08		10	9	
FD024		10	9	
FDQ19		10	9	
FDQ28		10	9	
FDQ31		10	9	
		•		
FDM52		-	7	
FDM50		7		
FDL50		7		
FDM10		-	7	
FDR51		-	7	
Length:	1559,	dtype:	int64	

df[['Item\_Fat\_Content']].value\_counts()

Item_Fat_Content	
Low Fat	8485
Regular	4824
LF	522
reg	195
low fat	178
dtype: int64	

```
df.replace({'Item_Fat_Content':{'LF':'Low Fat','reg':'Regular','low fat':'Low Fat'}},inpla
df[['Item_Fat_Content']].value_counts()
     Item_Fat_Content
     Low Fat
                         9185
                         5019
     Regular
     dtype: int64
df.replace({'Item_Fat_Content':{'Low Fat':0,'Regular':1}},inplace=True)
df[['Item_Type']].value_counts()
     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
df.replace({'Item_Type':{'Fruits and Vegetables':0,'Snack Foods':0,'Household':1,'Frozen F
                          'Canned':0,'Health and Hygiene':1,'Meat':0,'Soft Drinks':0,'Bread
                          'Others':2, 'Starchy Foods':0, 'Breakfast':0, 'Seafood':0}}, inplace=
df[['Item_Type']].value_counts()
     Item_Type
                  11518
     1
                   2406
                    280
     dtype: int64
df[['Outlet_Identifier']].value_counts()
     Outlet Identifier
     OUT027
                           1559
     OUT013
                           1553
     0UT035
                           1550
     0UT046
                           1550
```

```
0UT049
                           1550
     0UT045
                           1548
     0UT018
                           1546
     0UT017
                           1543
     OUT010
                            925
     0UT019
                            880
     dtype: int64
df.replace({'Outlet_Identifier':{'OUT027':0,'OUT013':1,
                                   'OUT049':2,'OUT046':3,'OUT035':4,
                                  'OUT045':5,'OUT018':6,
                                  'OUT017':7,'OUT010':8,'OUT019':9,
                                  }},inplace=True)
df[['Outlet_Identifier']].value_counts()
     Outlet_Identifier
                           1559
     1
                           1553
     2
                           1550
     3
                           1550
     4
                           1550
     5
                           1548
     6
                           1546
     7
                           1543
                            925
     9
                            880
     dtype: int64
df[['Outlet_Size']].value_counts()
     Outlet_Size
     1
                     7122
     0
                     5529
     2
                     1553
     dtype: int64
df.replace({'Outlet_Size':{'Small':0,'Medium':1,'High':2}},inplace=True)
df[['Outlet_Size']].value_counts()
     Outlet_Size
                     7122
     1
     0
                     5529
     2
                     1553
     dtype: int64
df[['Outlet_Location_Type']].value_counts()
     Outlet_Location_Type
     Tier 3
                              5583
     Tier 2
                              4641
     Tier 1
                              3980
     dtype: int64
```

```
df.replace({'Outlet_Location_Type':{'Tier 1':0,'Tier 2':1,'Tier 3':2}},inplace=True)
df[['Outlet_Location_Type']].value_counts()
     Outlet_Location_Type
                             5583
     1
                             4641
                             3980
     dtype: int64
df.replace({'Outlet_Type':{'Grocery Store':0,'Supermarket Type1':1,'Supermarket Type2':2,'
df[['Outlet_Type']].value_counts()
     Outlet_Type
                    9294
     0
                    1805
     3
                    1559
```

df.head()

dtype: int64

1546

	Item_Identifier	Item_Weight	<pre>Item_Fat_Content</pre>	Item_Visibility	<pre>Item_Type</pre>	Ite
0	FDT36	12.3	0	0.111448	0	3
1	FDT36	12.3	0	0.111904	0	3
2	FDT36	12.3	0	0.111728	0	3
3	FDT36	12.3	0	0.000000	0	3
4	FDP12	9.8	1	0.045523	0	3
4						•

df.head()

```
Item_Identifier Item_Weight Item_Fat_Content Item_Visibility Item_Type
      0
                  FDT36
                                 12.3
                                                     0
                                                                0.111448
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
        Item Weight
                                     14204 non-null float64
      1
      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
                                    14204 non-null int64
      8
          Outlet_Size
                                    14204 non-null int64
      9
          Outlet_Location_Type
      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
df.shape #get shape of datframe
     (14204, 12)
y=df['Item_Outlet_Sales']
y.shape
     (14204,)
у
     0
               436.608721
     1
               443.127721
     2
               564.598400
     3
              1719.370000
               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
```

X=df[['Item\_Weight', 'Item\_Fat\_Content', 'Item\_Visibility', 'Item\_Type', 'Item\_MRP', 'Outlet\_Identifier',

3

0

```
'Outlet_Establishment_Year', 'Outlet_Size', 'Outlet_Location_Type', 'Outlet Type']]
```

X=df.drop(['Item\_Identifier','Item\_Outlet\_Sales'],axis=1)

X.shape

(14204, 10)

Χ

	Item_Weight	<pre>Item_Fat_Content</pre>	<pre>Item_Visibility</pre>	<pre>Item_Type</pre>	Item_MRP	Outlet
0	12.300000	0	0.111448	0	33.4874	
1	12.300000	0	0.111904	0	33.9874	
2	12.300000	0	0.111728	0	33.9874	
3	12.300000	0	0.000000	0	34.3874	
4	9.800000	1	0.045523	0	35.0874	
14199	12.800000	0	0.069606	0	261.9252	
14200	12.800000	0	0.070013	0	262.8252	
14201	12.800000	0	0.069561	0	263.0252	
14202	13.659758	0	0.069282	0	263.5252	
14203	12.800000	0	0.069727	0	263.6252	
14204 rd	ows × 10 column	s				
4						•

from sklearn.preprocessing import StandardScaler

```
sc=StandardScaler()
```

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

X\_std=sc.fit\_transform(X\_std)

X std

```
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]])
```

X[['Item\_Weight','Item\_Visibility','Item\_MRP','Outlet\_Establishment\_Year']]=pd.DataFrame(X

Χ

	Item_Weight	<pre>Item_Fat_Content</pre>	Item_Visibility	<pre>Item_Type</pre>	Item_MRP	Outle <sup>-</sup>
0	-0.115417	0	0.884136	0	-1.731787	
1	-0.115417	0	0.893006	0	-1.723734	
2	-0.115417	0	0.889583	0	-1.723734	
3	-0.115417	0	-1.281712	0	-1.717291	
4	-0.703509	1	-0.397031	0	-1.706016	
14199	0.002201	0	0.070990	0	1.947664	
14200	0.002201	0	0.078898	0	1.962160	
14201	0.002201	0	0.070120	0	1.965381	
14202	0.204448	0	0.064694	0	1.973435	
14203	0.002201	0	0.073349	0	1.975046	
14204 rc	ows × 10 column	s				
4						•

from sklearn.model\_selection import train\_test\_split #Get train test split

X\_train,X\_test,y\_train,y\_test=train\_test\_split(X,y,test\_size=0.1,random\_state=2529)

X\_train.shape,X\_test.shape,y\_train.shape,y\_test.shape

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

from sklearn.ensemble import RandomForestRegressor

rfr=RandomForestRegressor(random\_state=2529)

rfr.fit(X\_train,y\_train)

RandomForestRegressor(random\_state=2529)

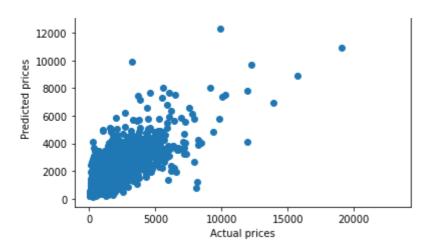
y\_pred=rfr.predict(X\_test) #Get Model Prediction

y\_pred.shape

```
(1421,)
y_pred
     array([1445.29507934, 669.51312572, 1883.54185796, ..., 2228.46101734,
            3251.93307564, 460.5156873 ])
from sklearn.metrics import mean_squared_error,mean_absolute_error,r2_score #Get Model Eva
mean_squared_error(y_test,y_pred)
     1611177.5560500463
mean_absolute_error(y_test,y_pred)
     828.3494726840753
r2_score(y_test,y_pred)
     0.5806344037136959
import matplotlib.pyplot as plt #get Visualization of actual vs predicted result
plt.scatter(y_test,y_pred)
plt.xlabel('Actual prices')
plt.ylabel('Predicted prices')
plt.title('Actual Price vs Predicted price')
plt.show()
```

 $\Box$ 

Actual Price us Prodicted price



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