

Importing the libraries

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
```

Importing the dataset

```
df =
pd.read_csv('https://github.com/YBI-Foundation/Dataset/raw/main/Big
%20Sales%20Data.csv')
```

Get Information of Dataframe

```
df.head()
```

	Item_Identifier	Item_Weight	Item_Fat_Content	Item_Visibility	\
0	FDT36	12.3	Low Fat	0.111448	
1	FDT36	12.3	Low Fat	0.111904	
2	FDT36	12.3	LF	0.111728	
3	FDT36	12.3	Low Fat	0.000000	
4	FDP12	9.8	Regular	0.045523	

	Item_Type	Item_MRP	Outlet_Identifier	Outlet_Establishment_Year	\
0	Baking Goods	33.4874	OUT049		1999
1	Baking Goods	33.9874	OUT017		2007
2	Baking Goods	33.9874	OUT018		2009
3	Baking Goods	34.3874	OUT019		1985
4	Baking Goods	35.0874	OUT017		2007

	Outlet_Size	Outlet_Location_Type	Outlet_Type
Item_Outlet_Sales			
0	Medium	Tier 1	Supermarket Type1
436.608721			
1	Medium	Tier 2	Supermarket Type1
443.127721			
2	Medium	Tier 3	Supermarket Type2
564.598400			
3	Small	Tier 1	Grocery Store
1719.370000			

4 Medium Tier 2 Supermarket Type1
352.874000

`df.info()` *#gives column name, count, not null category, D-type(data type)*

<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

`df.describe()` *#gives the linear relation of each column with another column*

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

	Item_Outlet_Sales
count	14204.000000
mean	2185.836320

```
std          1827.479550
min           33.290000
25%          922.135101
50%          1768.287680
75%          2988.110400
max          31224.726950
```

```
df.isnull().sum()  #(df.isna()).sum() gives same result)
 #gives the sum of all null values columns-wise
```

```
Item_Identifier      0
Item_Weight          2389
Item_Fat_Content      0
Item_Visibility      0
Item_Type            0
Item_MRP             0
Outlet_Identifier    0
Outlet_Establishment_Year  0
Outlet_Size          0
Outlet_Location_Type  0
Outlet_Type          0
Item_Outlet_Sales    0
dtype: int64
```

```
df.nunique()  #gives total no. of unique entries
```

```
Item_Identifier      1559
Item_Weight          416
Item_Fat_Content      5
Item_Visibility      13006
Item_Type            16
Item_MRP             8052
Outlet_Identifier     10
Outlet_Establishment_Year  9
Outlet_Size           3
Outlet_Location_Type  3
Outlet_Type           4
Item_Outlet_Sales    9144
dtype: int64
```

```
df.count()  #gives total no. of entries in columns it
```

```
Item_Identifier      14204
Item_Weight          11815
Item_Fat_Content      14204
Item_Visibility      14204
Item_Type            14204
Item_MRP             14204
Outlet_Identifier     14204
Outlet_Establishment_Year  14204
Outlet_Size          14204
```

```
Outlet_Location_Type      14204
Outlet_Type                14204
Item_Outlet_Sales          14204
dtype: int64
```

```
df.columns #give column names in the dataframe
```

```
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')
```

```
df.shape
```

```
(14204, 12)
```

```
df.dtypes
```

```
Item_Identifier      object
Item_Weight          float64
Item_Fat_Content      object
Item_Visibility      float64
Item_Type            object
Item_MRP             float64
Outlet_Identifier     object
Outlet_Establishment_Year  int64
Outlet_Size          object
Outlet_Location_Type  object
Outlet_Type          object
Item_Outlet_Sales    float64
dtype: object
```

Taking care of missing data

```
df['Item_Weight'].fillna(df.groupby(['Item_Type'])
                        ['Item_Weight'].transform('mean'),inplace=True)
```

```
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	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

```

```
df.describe()
```

	Item_Weight	Item_Visibility	Item_MRP
Outlet_Establishment_Year \			
count	14204.000000	14204.000000	14204.000000
mean	12.790642	0.065953	141.004977
std	4.251186	0.051459	62.086938
min	4.555000	0.000000	31.290000
25%	9.300000	0.027036	94.012000
50%	12.800000	0.054021	142.247000
75%	16.000000	0.094037	185.855600
max	30.000000	0.328391	266.888400

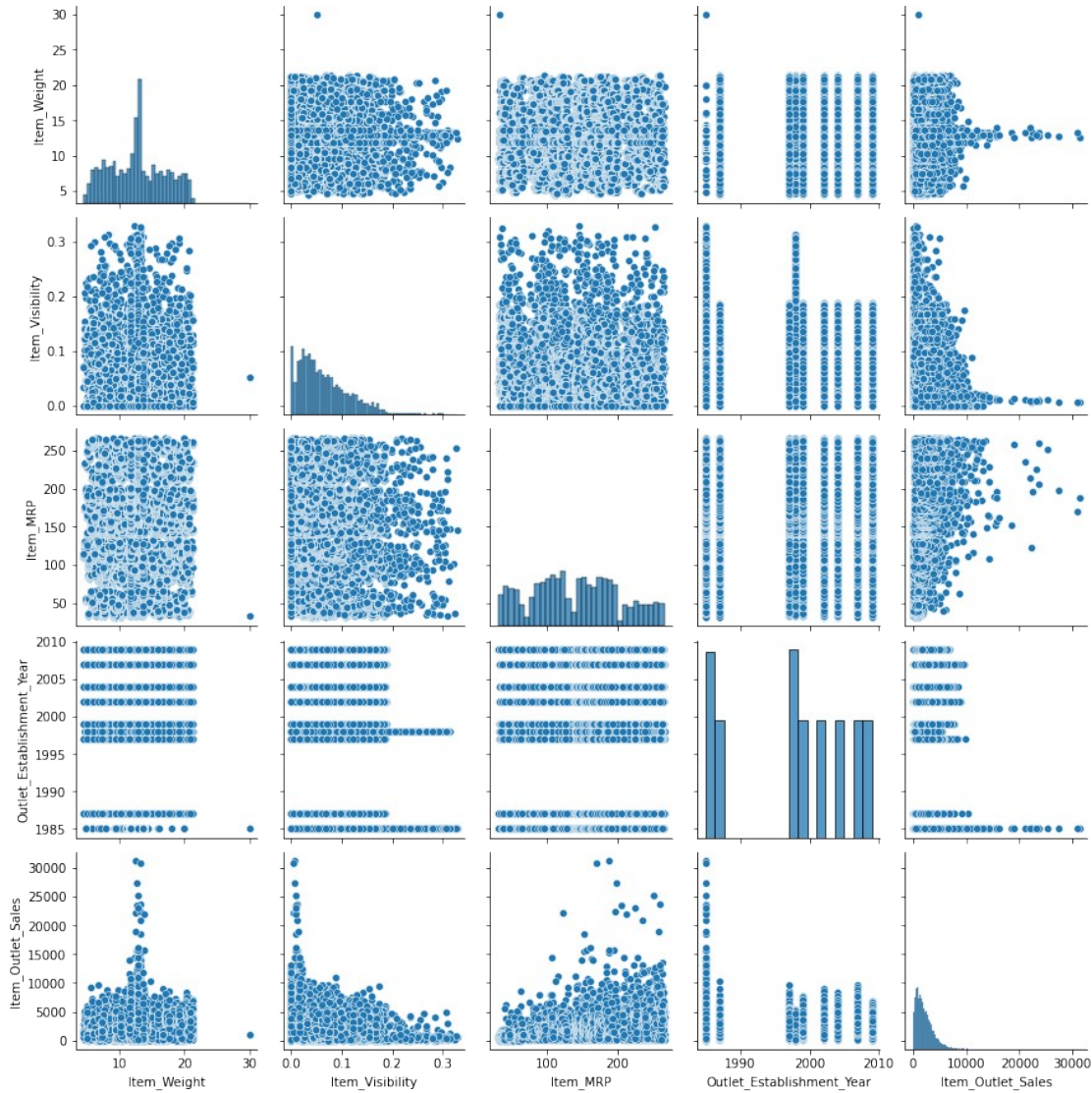
	Item_Outlet_Sales
count	14204.000000
mean	2185.836320
std	1827.479550
min	33.290000
25%	922.135101
50%	1768.287680
75%	2988.110400
max	31224.726950

```

import seaborn as sns
sns.pairplot(df)

```

```
<seaborn.axisgrid.PairGrid at 0x7f853f00bcd0>
```



et Categories and Counts of Categorical Variables

```
df[['Item_Identifier']].value_counts()
```

Item_Identifier	
FDQ08	10
FDQ24	10
FDQ19	10
FDQ28	10
FDQ31	10
...	...
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'}} inplace=True)

df['Item_Fat_Content'].value_counts()

Low Fat      9185
Regular      5019
Name: Item_Fat_Content, 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 Foods':0,'Diary':0,'Baking
Goods':0,'Canned':0, 'Health and Hygiene':1,
                        'Heat':0, 'Soft Drinks':0,'Breads':0,'Hard

```

```
Drinks':0, 'Others':2, 'Starchy Foods':0, 'Breakfast':0,
'Seafood':0}}, inplace=True)
```

```
df[['Item_Type']].value_counts()
```

```
Item_Type
0          9646
1          2406
Dairy      1136
Meat        736
2           280
dtype: int64
```

```
df[['Outlet_Identifier']].value_counts()
```

```
Outlet_Identifier
OUT027          1559
OUT013          1553
OUT035          1550
OUT046          1550
OUT049          1550
OUT045          1548
OUT018          1546
OUT017          1543
OUT010           925
OUT019           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_Size']].value_counts()
```

```
Outlet_Size
Medium      7122
Small       5529
High        1553
dtype: int64
```

```
df.replace({'Outlet_Size': {'Small':0, 'Medium':1, 'High': 2}},
inplace=True)
```

```
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_Type']].value_counts()
```

```
Outlet_Type
Supermarket Type1    9294
Grocery Store        1805
Supermarket Type3    1559
Supermarket Type2    1546
dtype: int64
```

```
df.replace({'Outlet_Type':{'Grocery Store':0, 'Supermarket Type1':1, 'Supermarket Type2':2, 'Supermarket Type3':3}}, inplace=True)
```

```
df.replace({'Item_Type':{'Dairy':3, 'Meat':4}}, inplace=True)
```

```
df.head()
```

	Item_Identifier	Item_Weight	Item_Fat_Content	Item_Visibility
0	FDT36	12.3	0	0.111448
0				
1	FDT36	12.3	0	0.111904
0				
2	FDT36	12.3	0	0.111728
0				
3	FDT36	12.3	0	0.000000
0				
4	FDP12	9.8	1	0.045523
0				

	Item_MRP	Outlet_Identifier	Outlet_Establishment_Year	Outlet_Size
0	33.4874	2	1999	1
1	33.9874	7	2007	1
2	33.9874	6	2009	1
3	34.3874	9	1985	0
4	35.0874	7	2007	1

	Outlet_Location_Type	Outlet_Type	Item_Outlet_Sales
0	0	1	436.608721
1	1	1	443.127721
2	2	2	564.598400
3	0	0	1719.370000
4	1	1	352.874000

```

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   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

#Define y
y=df['Item_Outlet_Sales']

y.shape

(14204,)

y
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

df.columns

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')

```

```

X=df.drop(['Item_Identifier','Item_Outlet_Sales'],axis=1)

```

X

	Item_Weight	Item_Fat_Content	Item_Visibility	Item_Type
Item_MRP \				
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				

	Outlet_Identifier	Outlet_Establishment_Year	Outlet_Size \
0	2	1999	1
1	7	2007	1
2	6	2009	1
3	9	1985	0
4	7	2007	1
...
14199	4	2004	0
14200	7	2007	1
14201	1	1987	2
14202	0	1985	1
14203	2	1999	1

	Outlet_Location_Type	Outlet_Type
0	0	1
1	1	1
2	2	2
3	0	0
4	1	1

```

...
14199          1          1
14200          1          1
14201          2          1
14202          2          3
14203          0          1

```

```
[14204 rows x 10 columns]
```

```
#Get X Variables Standardized
```

```
from sklearn.preprocessing import StandardScaler
```

```
sc=StandardScaler()
```

```
X_std=
df[['Item_Weight','Item_Visibility','Item_MRP','Outlet_Establishment_Y
ear']]
```

```
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_Ye
ar']] = pd.DataFrame(X_std,
columns=[['Item_Weight','Item_Visibility','Item_MRP','Outlet_Establish
ment_Year']])
```

```
X
```

```

      Item_Weight  Item_Fat_Content  Item_Visibility  Item_Type
Item_MRP \
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				

	Outlet_Identifier	Outlet_Establishment_Year	Outlet_Size	\
0	2	0.139681	1	
1	7	1.095319	1	
2	6	1.334228	1	
3	9	-1.532686	0	
4	7	1.095319	1	
...	
14199	4	0.736955	0	
14200	7	1.095319	1	
14201	1	-1.293777	2	
14202	0	-1.532686	1	
14203	2	0.139681	1	

	Outlet_Location_Type	Outlet_Type
0	0	1
1	1	1
2	2	2
3	0	0
4	1	1
...
14199	1	1
14200	1	1
14201	2	1
14202	2	3
14203	0	1

[14204 rows x 10 columns]

Splitting the dataset into the Training set and Test set

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size =
0.3, random_state = 1)
```

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

```
((9942, 10), (4262, 10), (9942,), (4262,))
```

```
df['Item_Type'].value_counts()
```

```
0    9646
```

```
1    2406
```

```
3    1136
```

```
4     736
```

```
2     280
```

```
Name: Item_Type, dtype: int64
```

```
X_train
```

	Item_Weight	Item_Fat_Content	Item_Visibility	Item_Type
Item_MRP \				
12210	-1.484495	0	-0.620225	0
0.099906				
12156	1.284242	0	-0.228613	0
0.032382				
1797	-0.105812	1	-0.230054	0 -
1.329063				
10348	-1.787950	1	-0.420379	4 -
0.597667				
2505	-0.105812	1	-0.186842	0
0.702572				
...
...				
905	1.166623	1	0.037543	0
0.746061				
5192	-0.824656	1	-0.324191	0
1.657564				
12172	0.872577	1	1.834909	0
0.046836				
235	1.225432	0	-1.115763	0 -
0.962284				
13349	0.519722	0	-0.328706	0 -
1.217500				

	Outlet_Identifier	Outlet_Establishment_Year	Outlet_Size	\
12210	6	1.334228	1	
12156	1	-1.293777	2	
1797	0	-1.532686	1	
10348	2	0.139681	1	
2505	9	-1.532686	0	
...	
905	7	1.095319	1	
5192	4	0.736955	0	
12172	6	1.334228	1	
235	4	0.736955	0	
13349	1	-1.293777	2	

	Outlet_Location_Type	Outlet_Type
12210	2	2

12156	2	1
1797	2	3
10348	0	1
2505	0	0
...
905	1	1
5192	1	1
12172	2	2
235	1	1
13349	2	1

[9942 rows x 10 columns]

Model

```
from sklearn.ensemble import RandomForestRegressor
rfg=RandomForestRegressor()
rfg.fit(X_train,y_train)

RandomForestRegressor()
```

Model Prediction

```
y_pred=rfg.predict(X_test)

y_pred.shape

(4262,)

y_pred

array([2946.01017703, 2029.51391296, 1134.53111942, ...,
       2332.65014151,
       2471.72123057, 1382.6964982 ])
```

Model Evaluation

```
from sklearn.metrics import mean_squared_error, mean_absolute_error,
mean_absolute_percentage_error, r2_score

mean_squared_error(y_test,y_pred)

1636851.7710929627

mean_absolute_percentage_error(y_test,y_pred)

0.7260326369580009

r2_score(y_test,y_pred)

0.48070477430390246
```

```
#Get Visualization of Actual Vs Predicted Results
```

```
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()
```

