

## 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

```

```
count    Item_Outlet_Sales  
mean      14204.000000  
          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

```

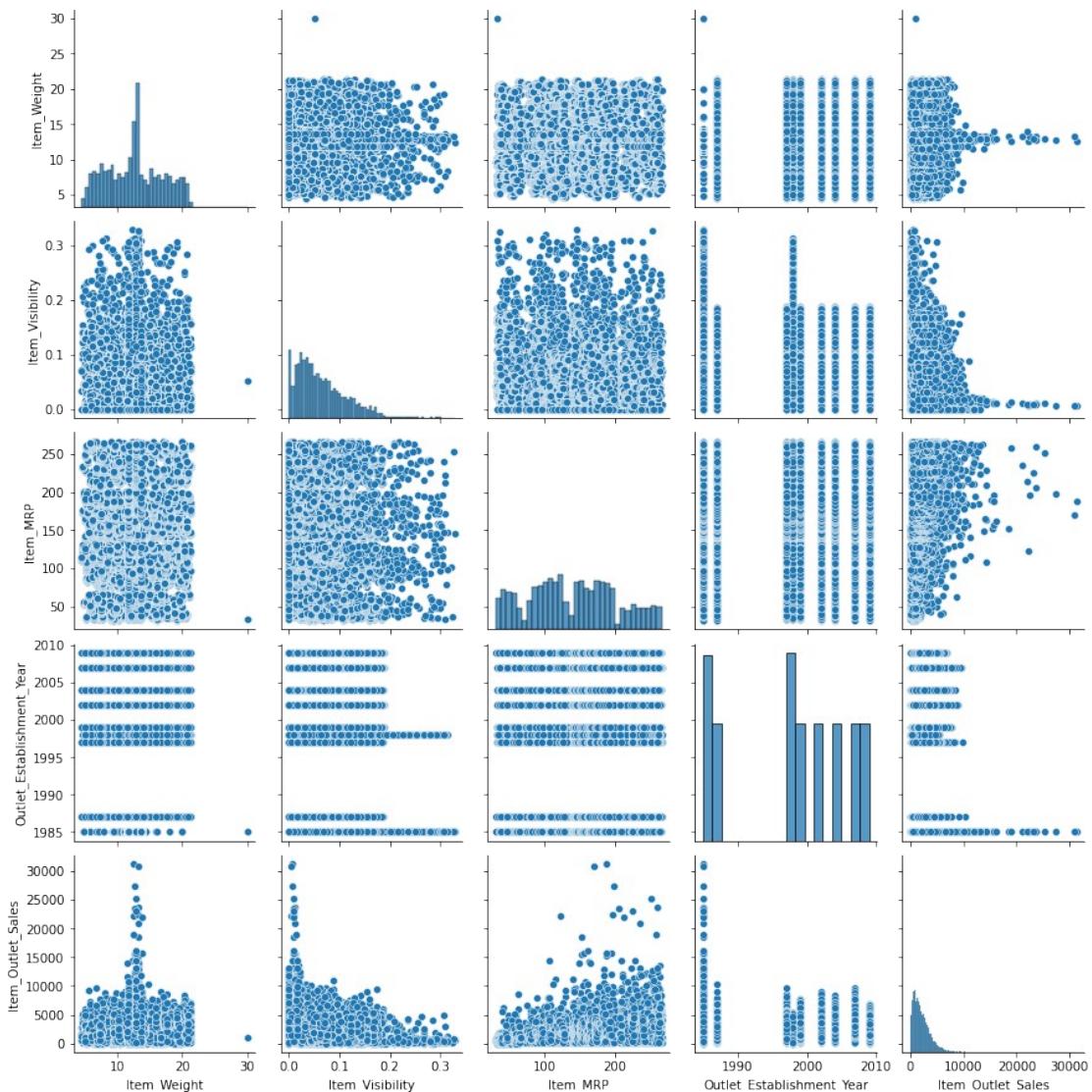
```
4   Item_Type                  14204 non-null  object
5   Item_MRP                   14204 non-null  float64
6   Outlet_Identifier          14204 non-null  object
7   Outlet_Establishment_Year  14204 non-null  int64
8   Outlet_Size                 14204 non-null  object
9   Outlet_Location_Type       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
1997.830681			
std	4.251186	0.051459	62.086938
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			
75%	16.000000	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		

```
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
Item_Type \
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_MRP	Item_Weight	Item_Fat_Content	Item_Visibility	Item_Type
0	33.4874	12.300000	0	0.111448	0
1	33.9874	12.300000	0	0.111904	0
2	33.9874	12.300000	0	0.111728	0
3	34.3874	12.300000	0	0.000000	0
4	35.0874	9.800000	1	0.045523	0
...	...	...	...	...	...
14199	261.9252	12.800000	0	0.069606	0
14200	262.8252	12.800000	0	0.070013	0
14201	263.0252	12.800000	0	0.069561	0
14202	263.5252	13.659758	0	0.069282	0
14203	263.6252	12.800000	0	0.069727	0
...	...	...	...	...	...
0	14199	Outlet_Identifier	Outlet_Establishment_Year	Outlet_Size	\
1	14200	2	1999	1	
2	14201	7	2007	1	
3	14202	6	2009	1	
4	14203	9	1985	0	
...	...	7	2007	1	
14199	...	4	2004	0	
14200	...	7	2007	1	
14201	...	1	1987	2	
14202	...	0	1985	1	
14203	...	2	1999	1	
...	...	...	...	...	...
0	14199	Outlet_Location_Type	Outlet_Type		
1	14200	0	1		
2	14201	1	1		
3	14202	2	2		
4	14203	0	0		
...	...	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_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_std,
columns=[['Item_Weight','Item_Visibility','Item_MRP','Outlet_Establishment_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				
0	Outlet_Identifier	Outlet_Establishment_Year	Outlet_Size	\
1	2	0.139681	1	
2	7	1.095319	1	
3	6	1.334228	1	
4	9	-1.532686	0	
..	..	..	..	..
14199	7	1.095319	1	
14200	4	0.736955	0	
14201	1	-1.293777	2	
14202	0	-1.532686	1	
14203	2	0.139681	1	
0	Outlet_Location_Type	Outlet_Type		
1	0	1		
2	1	1		
3	2	2		
4	0	0		
..	..	..	..	..
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()
```

