<pre>In [1]: import pandas as pd In [2]: import numpy as np In [3]: df = pd.read_csv(r'https://raw.githubusercontent.com/YBI-Foundation/Dataset/main/Big%20Sales%20Data.csv')</pre>
In [4]:
3 FDT36 12.3 Low Fat 0.00000 Baking Goods 34.3874 OUT019 1985 Small Tier 1 Grocery Store 1719.370000 4 FDP12 9.8 Regular 0.045523 Baking Goods 35.0874 OUT017 2007 Medium Tier 2 Supermarket Type1 352.874000 In [5]: df.info() <class 'pandas.core.frame.dataframe'=""> RangeIndex: 14204 entries, 0 to 14203 Data columns (total 12 columns):</class>
Column Non-Null Count Dtype
7 Outlet_Establishment_Year 14204 non-null inte4 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]: df.columns
<pre>Out[6]: Index(['Item_Identifier', 'Item_Weight', 'Item_Fat_Content', 'Item_Visibility',</pre>
Out [7]: Item_Weight Item_MRP Outlet_Establishment_Year Item_Outlet_Sales count 11815.00000 14204.00000 14204.00000 14204.00000 14204.00000 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 0.00000 31.29000 33.29000 25% 8.71000 0.027036 94.01200 1987.00000 922.135101
50% 12.50000 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 [8]: df['Item_Weight'].fillna(df.groupby(['Item_Type'])['Item_Weight'].transform('mean'),inplace=True)
In [9]: df.info() <class 'pandas.core.frame.dataframe'=""> RangeIndex: 14204 entries, 0 to 14203 Data columns (total 12 columns): # Column Non-Null Count Dtype</class>
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
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 0.00000 31.290000 1985.00000 33.290000 25% 9.300000 0.027036 94.012000 1987.00000 922.135101 50% 12.80000 0.054021 142.247000 1999.000000 1768.287680
75% 16.00000 0.094037 185.855600 2004.00000 2988.110400 max 30.00000 0.328391 266.888400 2009.00000 31224.726950 In [11]: import seaborn as sns sns.pairplot(df) Out[11]: seaborn.axisgrid.PairGrid at 0x1c1a0a56530>
25
20000
<pre>In [12]: df[['Item_Identifier']].value_counts() Out[12]:</pre>
FDM52 7 FDM50 7 FDL50 7 FDM10 7 FDR51 7 Length: 1559, dtype: int64 In [13]: df[['Item_Fat_Content']].value_counts()
<pre>Out[13]: Item_Fat_Content Low Fat</pre>
<pre>In [15]: df[['Item_Fat_Content']].value_counts() Out[15]:</pre>
In [17]: df[['Item_Type']].value_counts() Out[17]: 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 [18]: df.replace({'Item_Type':{ 'Fruits and Vegetables':0, 'Snack Foods':0, 'Household':1, 'Frozen Foods': 0, 'Dairy': 0, 'Baking Goods': 0, 'Canned': 0, 'Health and Hygiene': 1, 'Meat': 0, 'Soft Drinks': 0, 'Breads': 0, 'Hard Drinks': 0,
Titem_Type 11518
<pre>In [20]: df[['Outlet_Identifier']].value_counts() Out[20]:</pre>
OUT045
'OUT045':5,'OUT018':6,
2 1550 3 1550 4 1550 5 1548 6 1546 7 1543 8 925 9 880 dtype: int64
<pre>In [23]: df[['Outlet_Size']].value_counts() Out[23]:</pre>
In [25]: df[['Outlet_Size']].value_counts() Out[25]:
<pre>In [26]: df[['Outlet_Location_Type']].value_counts() Out[26]:</pre>
<pre>In [28]: df[['Outlet_Location_Type']].value_counts() Out[28]:</pre>
<pre>In [29]: df[['Outlet_Type']].value_counts() Out[29]: Outlet_Type Supermarket Type1 9294 Grocery Store 1805 Supermarket Type3 1559 Supermarket Type2 1546 dtype: int64 In [30]: df.replace(('Outlet_Type': {'Grocery Store': 0, 'Supermarket Type1': 1, 'Supermarket Type2': 2, 'Supermarket Type3': 3}}, inplace=True)</pre>
In [31]: df[['Outlet_Type']].value_counts() Out[31]: 0 1 9294 0 1805 3 1559 2 1546 dtype: int64
In [32]: df.head() Out[32]: tem_Identifier tem_Weight tem_Fat_Content tem_Visibility tem_Type tem_MRP Outlet_Identifier Outlet_Establishment_Year Outlet_Size Outlet_Location_Type Outlet_Type tem_Outlet_Sales 1 FDT36 12.3 0 0.111448 0 33.4874 2 1999 1 0 0 1 436.608721 2 FDT36 12.3 0 0.111728 0 33.9874 6 2009 1 2 2 564.598400
3 FDT36 12.3 0 0.000000 0 34.3874 9 1985 0 0 0 1719.370000 4 FDP12 9.8 1 0.045523 0 35.0874 7 2007 1 1 1 352.874000 In [33]: df.info() <class 'pandas.core.frame.dataframe'=""> RangeIndex: 14204 entries, 0 to 14203</class>
Data columns (total 12 columns): # Column Non-Null Count Dtype
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 [34]: y=df['Item_Outlet_Sales']
In [35]: y Out[35]: 0
14199
'Item_Type', 'Item_MRP', 'Outlet_Identifier', 'Outlet_Establishment_Year', 'Outlet_Size', 'Outlet_Location_Type', 'Outlet_Type' In [37]: X Out[37]: Item_Weight Item_Fat_Content Item_Visibility Item_Type Item_MRP Outlet_Identifier Outlet_Establishment_Year Outlet_Size Outlet_Location_Type Outlet_Type
0 12.300000 0 0.111448 0 33.4874 2 1999 1 0 1 1 12.300000 0 0.111904 0 33.9874 7 2007 1 1 1 2 12.300000 0 0.111728 0 33.9874 6 2009 1 2 2 3 12.300000 0 0.000000 0 34.3874 9 1985 0 0 0 4 9.800000 1 0.045523 0 35.0874 7 2007 1 1 1
14199 12.800000 0 0.069606 0 261.9252 4 2004 0 1 1 14200 12.800000 0 0.070013 0 262.8252 7 2007 1 1 1 14201 12.800000 0 0.069561 0 263.0252 1 1987 2 2 1 14202 13.659758 0 0.069282 0 263.6252 0 1985 1 2 3 14203 12.800000 0 0.069727 0 263.6252 2 1999 1 0 1
14204 rows × 10 columns In [38]: from sklearn.preprocessing import StandardScaler In [39]: sc = StandardScaler() In [40]: X_std = df[['Item_Weight', 'Item_Visibility', 'Item_MRP', 'Outlet_Establishment_Year']]
<pre>In [41]: X_std = sc.fit_transform(X_std) In [42]: X_std Out[42]: array([[-0.11541705,</pre>
[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 [46]: X[['Item_Weight', 'Item_Visibility', 'Item_MRP', 'Outlet_Establishment_Year']] = pd.DataFrame(X_std, columns=[['Item_Weight', 'Item_Visibility', 'Item_MRP', 'Outlet_Establishment_Year']]) C:\Users\balav\AppData\Local\Temp\ipykernel_24632\3276057480.py:1: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy X[['Item_Weight', 'Item_Visibility', 'Item_MRP', 'Outlet_Establishment_Year']] = pd.DataFrame(X_std, columns=[['Item_Weight', 'Item_Visibility', 'Item_MRP', 'Outlet_Establishment_Year']]) In [47]: X Out[47]: tem_Weight tem_Fat_Content tem_Visibility tem_Type tem_MRP Outlet_Identifier Outlet_Establishment_Year Outlet_Size Outlet_Location_Type Outlet_Type 0
1 -0.115417 0 0.893006 0 -1.723734 7 1.095319 1 1 1 2 -0.115417 0 0.889583 0 -1.723734 6 1.334228 1 2 2 3 -0.115417 0 -1.281712 0 -1.717291 9 -1.532686 0 0 0 4 -0.703509 1 -0.397031 0 -1.706016 7 1.095319 1 1 1
14199 0.002201 0 0.070990 0 1.947664 4 0.736955 0 1 1 14200 0.002201 0 0.078898 0 1.962160 7 1.095319 1 1 1 14201 0.002201 0 0.070120 0 1.965381 1 1.293777 2 2 1 14202 0.204448 0 0.064694 0 1.973435 0 -1.532686 1 2 3 14203 0.002201 0 0.073349 0 1.975046 2 0.139681 1 0 1
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, train_size= 0.1, random_state = 22529) In [50]: from sklearn.ensemble import RandomForestRegressor
In [51]: rfr = RandomForestRegressor(random_state=22529) In [52]: rfr.fit(X_train, y_train) Out[52]: RandomForestRegressor RandomForestRegressor(random_state=22529)
In [53]: y_pred = rfr.predict(X_test) In [54]: y_pred Out[54]: array([842.17961098, 989.97031772, 2975.19528623,, 818.18826824, 2954.7788205, 1676.82029206]) In [55]: from sklearn.metrics import mean_squared_error, mean_absolute_error, r2_score
In [56]: mean_squared_error(y_test, y_pred) Out[56]: 1779270.7008865657 In [57]: mean_absolute_error(y_test, y_pred) Out[57]: 832.1364705201643
In [58]: r2_score(y_test, y_pred) Out[58]: 0.47509041569762833 In [59]: import matplotlib.pyplot as plt plt.scatter(y_test, y_pred) plt.xlabel("actual Price") plt.ylabel("Predicted Price") plt.title("Actual Price vs Predicted Price")
plt.title("Actual Price Vs Predicted Price") plt.show() Actual Price Vs Predicted Price 12000 - 10000 -
Predicted Price 6000 -
2000 -
0 5000 10000 15000 20000 30000 Actual Price