

Bigmart Sales Data

Load the CSV files into pandas DataFrames

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns


test_file_path = '/content/Test.csv'
train_file_path = '/content/Train.csv'

test_df = pd.read_csv(test_file_path)
train_df = pd.read_csv(train_file_path)

merged_df = pd.concat([test_df, train_df], ignore_index=True)

merged_df.to_csv('/content/merged_data.csv', index=False)

#print(merged_df)
merged_df.head(5)
```



	Item_Identifier	Item_Weight	Item_Fat_Content	Item_Visibility	Item_Type	Item_MRP	Outlet_Identifier	Outlet_Establishment_Year	Outlets
0	FDW58	20.750	Low Fat	0.007565	Snack Foods	107.8622	OUT049	1999	1
1	FDW14	8.300	reg	0.038428	Dairy	87.3198	OUT017	2007	1
2	NCN55	14.600	Low Fat	0.099575	Others	241.7538	OUT010	1998	1
3	FDQ58	7.315	Low Fat	0.015388	Snack Foods	155.0340	OUT017	2007	1
4	FDY38	NaN	Regular	0.118599	Dairy	234.2300	OUT027	1985	1

Next steps:

[Generate code with merged_df](#)


 [View recommended plots](#)

[New interactive sheet](#)

Analyze BigMart Sales Data

Descriptive Statistics:

```
print(merged_df.describe())
```



	Item_Weight	Item_Visibility	Item_MRP	Outlet_Establishment_Year	\
count	11765.000000	14204.000000	14204.000000	14204.000000	
mean	12.792854	0.065953	141.004977	1997.830681	
std	4.652502	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%	12.600000	0.054021	142.247000	1999.000000	
75%	16.750000	0.094037	185.855600	2004.000000	
max	21.350000	0.328391	266.888400	2009.000000	

	Item_Outlet_Sales
count	8523.000000
mean	2181.288914
std	1706.499616
min	33.290000
25%	834.247400
50%	1794.331000
75%	3101.296400
max	13086.964800

✓ Data Information:

```
print(merged_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          11765 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         10188 non-null  object
 9   Outlet_Location_Type 14204 non-null  object
10   Outlet_Type          14204 non-null  object
11   Item_Outlet_Sales    8523 non-null   float64
dtypes: float64(4), int64(1), object(7)
memory usage: 1.3+ MB
None
```

✓ Explore Outlet Information:

```
print(merged_df['Outlet_Identifier'].value_counts())
print(merged_df.groupby('Outlet_Type')['Item_Outlet_Sales'].mean())
```

```
Outlet_Identifier
OUT027      1559
OUT013      1553
OUT049      1550
OUT046      1550
OUT035      1550
OUT045      1548
OUT018      1546
OUT017      1543
OUT010       925
OUT019       880
Name: count, dtype: int64
Outlet_Type
Grocery Store      339.828500
Supermarket Type1  2316.181148
Supermarket Type2  1995.498739
Supermarket Type3  3694.038558
Name: Item_Outlet_Sales, dtype: float64
```

✓ Analyze Item Information:

```
print(merged_df['Item_Type'].value_counts())
print(merged_df.groupby('Item_Type')['Item_Outlet_Sales'].mean())
```

```
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
Name: count, dtype: int64
Item_Type
Baking Goods            1952.971207
```

```

Breads                2204.132226
Breakfast              2111.808651
Canned                 2225.194904
Dairy                  2232.542597
Frozen Foods           2132.867744
Fruits and Vegetables  2289.009592
Hard Drinks            2139.221622
Health and Hygiene     2010.000265
Household              2258.784300
Meat                   2158.977911
Others                 1926.139702
Seafood                2326.065928
Snack Foods            2277.321739
Soft Drinks            2006.511735
Starchy Foods          2374.332773
Name: Item_Outlet_Sales, dtype: float64

```

Relationship between Item Visibility and Sales:

```
print(merged_df[['Item_Visibility', 'Item_Outlet_Sales']].corr())
```

```

Item_Visibility  Item_Outlet_Sales
Item_Visibility    1.000000        -0.128625
Item_Outlet_Sales  -0.128625         1.000000

```

Double-click (or enter) to edit

Sales Trends Analysis

```

sales_by_year = merged_df.groupby('Outlet_Establishment_Year')['Item_Outlet_Sales'].sum()
print(sales_by_year)

```

```

Outlet_Establishment_Year
1985    3.633620e+06
1987    2.142664e+06
1997    2.118395e+06
1998    1.883402e+05
1999    2.183970e+06
2002    2.036725e+06
2004    2.268123e+06
2007    2.167465e+06
2009    1.851823e+06
Name: Item_Outlet_Sales, dtype: float64

```

Sales Trends by Outlet Type

```

sales_by_outlet_type = merged_df.groupby('Outlet_Type')['Item_Outlet_Sales'].sum()
print(sales_by_outlet_type)

```

```

Outlet_Type
Grocery Store    3.680343e+05
Supermarket Type1  1.291734e+07
Supermarket Type2  1.851823e+06
Supermarket Type3  3.453926e+06
Name: Item_Outlet_Sales, dtype: float64

```

Double-click (or enter) to edit

Sales Trends by Item Type

```

sales_by_item_type = merged_df.groupby('Item_Type')['Item_Outlet_Sales'].sum()
print(sales_by_item_type)

```

```

Item_Type
Baking Goods    1.265525e+06
Breads          5.532372e+05

```

```

Breakfast      2.322990e+05
Canned         1.444151e+06
Dairy          1.522594e+06
Frozen Foods   1.825735e+06
Fruits and Vegetables  2.820060e+06
Hard Drinks    4.577934e+05
Health and Hygiene  1.045200e+06
Household      2.055494e+06
Meat           9.175656e+05
Others         3.255176e+05
Seafood        1.488682e+05
Snack Foods    2.732786e+06
Soft Drinks    8.928977e+05
Starchy Foods  3.514013e+05
Name: Item_Outlet_Sales, dtype: float64

```

✎ Explore Item Fat Content and Sales:

```
print(merged_df.groupby('Item_Fat_Content')['Item_Outlet_Sales'].mean())
```

```

Item_Fat_Content
LF      2073.551928
Low Fat  2164.477336
Regular  2235.186702
low fat  2087.740737
reg      1962.192268
Name: Item_Outlet_Sales, dtype: float64

```

✎ Analyze Outlet Location Type and Sales:

```
print(merged_df.groupby('Outlet_Location_Type')['Item_Outlet_Sales'].mean())
```

```

Outlet_Location_Type
Tier 1      1876.909159
Tier 2     2323.990559
Tier 3     2279.627651
Name: Item_Outlet_Sales, dtype: float64

```

✎ Identify Missing Values:

```
print(merged_df.isnull().sum())
```

```

Item_Identifier      0
Item_Weight          2439
Item_Fat_Content      0
Item_Visibility      0
Item_Type            0
Item_MRP             0
Outlet_Identifier    0
Outlet_Establishment_Year  0
Outlet_Size          4016
Outlet_Location_Type  0
Outlet_Type          0
Item_Outlet_Sales    5681
dtype: int64

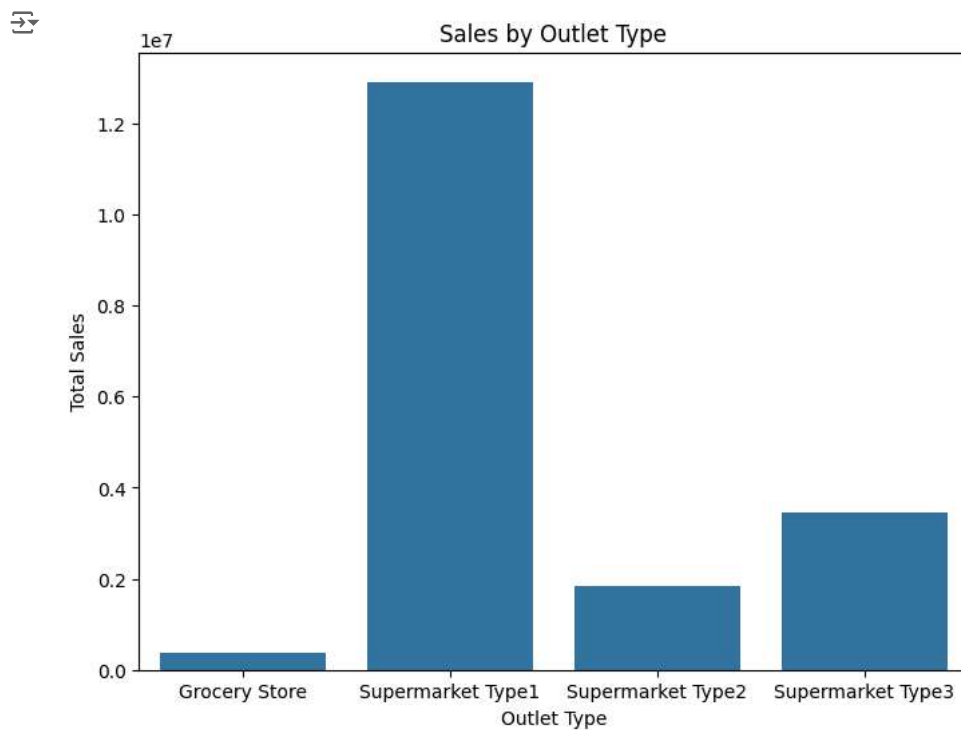
```

✎ Analyze Sales by Outlet Type:

```

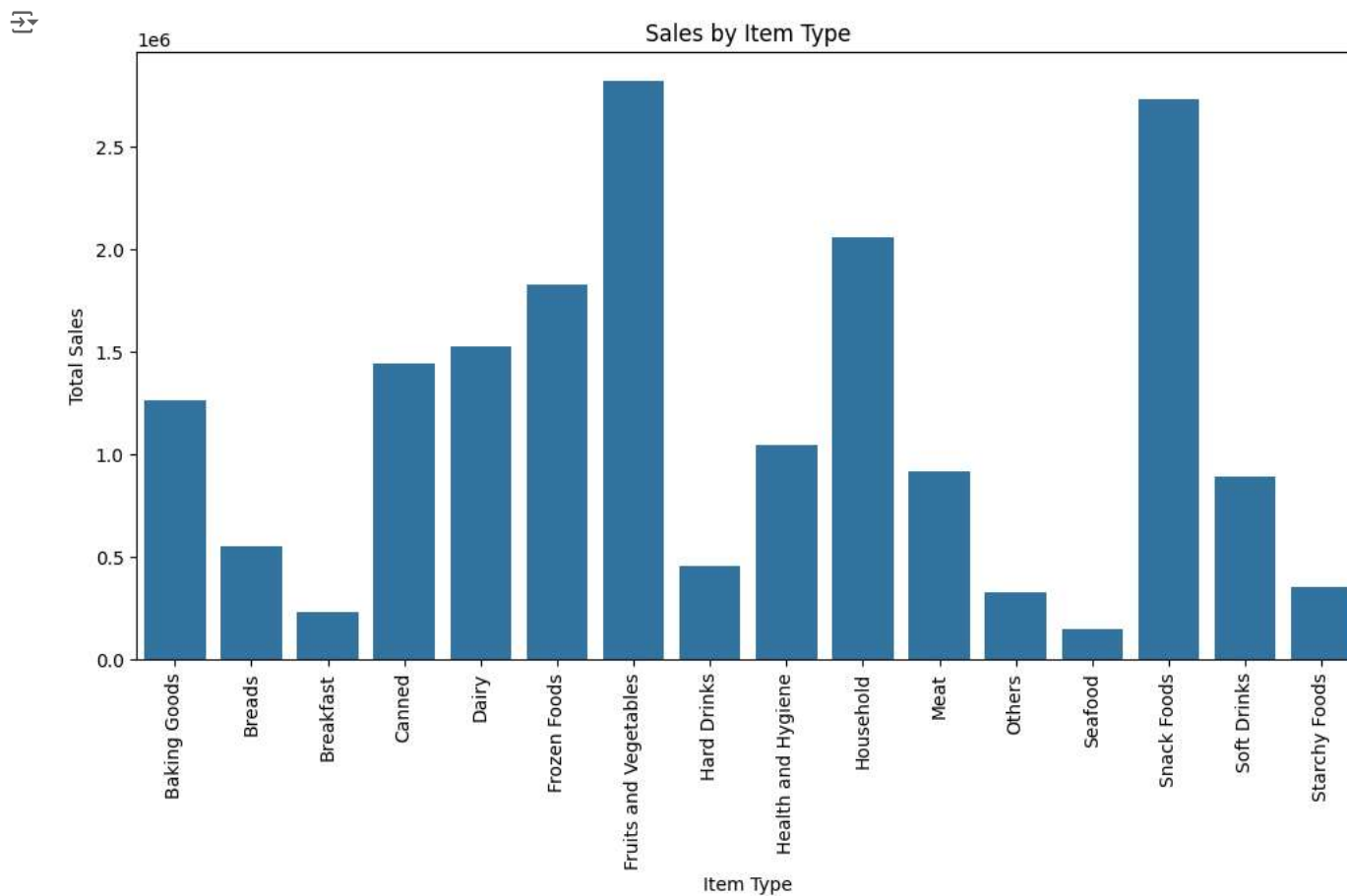
outlet_sales = merged_df.groupby('Outlet_Type')['Item_Outlet_Sales'].sum()
plt.figure(figsize=(8, 6))
sns.barplot(x=outlet_sales.index, y=outlet_sales.values)
plt.xlabel('Outlet Type')
plt.ylabel('Total Sales')
plt.title('Sales by Outlet Type')
plt.show()

```



✓ Analyze Sales by Item Type:

```
item_sales = merged_df.groupby('Item_Type')['Item_Outlet_Sales'].sum()
plt.figure(figsize=(12, 6))
sns.barplot(x=item_sales.index, y=item_sales.values)
plt.xlabel('Item Type')
plt.ylabel('Total Sales')
plt.title('Sales by Item Type')
plt.xticks(rotation=90)
plt.show()
```



✓ Analyze Sales by Outlet Location Type:

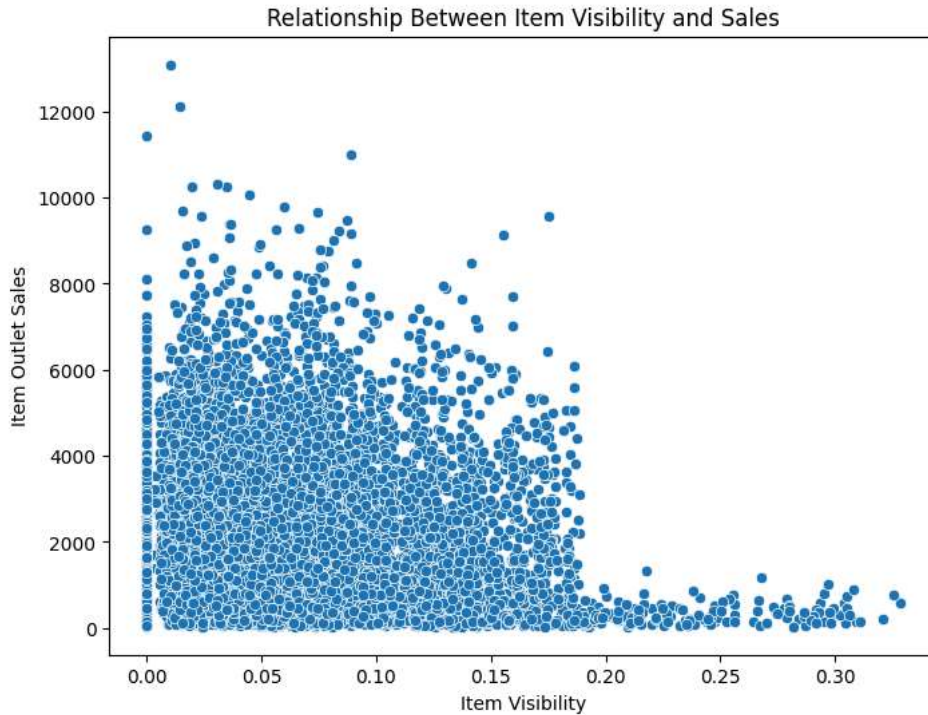
```
location_sales = merged_df.groupby('Outlet_Location_Type')['Item_Outlet_Sales'].sum()
plt.figure(figsize=(8, 6))
sns.barplot(x=location_sales.index, y=location_sales.values)
plt.xlabel('Outlet Location Type')
plt.ylabel('Total Sales')
plt.title('Sales by Outlet Location Type')
plt.show()
```

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Sales by Outlet Location Type

▼ Analyze Relationship Between Item Visibility and Sales:

```
plt.figure(figsize=(8, 6))
sns.scatterplot(x='Item_Visibility', y='Item_Outlet_Sales', data=merged_df)
plt.xlabel('Item Visibility')
plt.ylabel('Item Outlet Sales')
plt.title('Relationship Between Item Visibility and Sales')
plt.show()
```



▼ Analyze Relationship Between Item MRP and Sales:

```
plt.figure(figsize=(8, 6))
sns.scatterplot(x='Item_MRP', y='Item_Outlet_Sales', data=merged_df)
plt.xlabel('Item MRP')
plt.ylabel('Item Outlet Sales')
plt.title('Relationship Between Item MRP and Sales')
plt.show()
```

