

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

# Step 1: Load the Dataset
url = "https://github.com/akki8087/Big-Mart-Sales/raw/refs/heads/master/Train.csv"
data = pd.read_csv(url)

# Display basic information about the dataset
print("Dataset Overview:")
print(data.head())
print("\nSummary Statistics:")
print(data.describe())
print("\nDataset Info:")
print(data.info())

# Select relevant columns for the analysis
columns_of_interest = ['Item_Outlet_Sales', 'Outlet_Size', 'Item_Type', 'Outlet_Establishment_Year', 'Outlet_Location_Type']
data = data[columns_of_interest]

# Step 2: Compute the Correlation Matrix
# Convert categorical variables to numerical representations
encoded_data = pd.get_dummies(data, drop_first=True)
correlation_matrix = encoded_data.corr()

# Step 3: Visualize the Correlation Matrix
plt.figure(figsize=(12, 8))
sns.heatmap(correlation_matrix, annot=True, fmt=".2f", cmap="coolwarm", cbar=True)
plt.title("Correlation Matrix for Big Mart Sales")
plt.show()

# Step 4: Interpret the Results
# Identify strong correlations (absolute value > 0.5)
strong_correlations = correlation_matrix['Item_Outlet_Sales'][abs(correlation_matrix['Item_Outlet_Sales']) > 0.5]
print("\nStrong Correlations with Sales:")
print(strong_correlations)

# Observations based on the heatmap and correlations
print("\nObservations:")
print("1. Variables with a strong correlation to sales can be further explored.")
print("2. Additional analysis can include examining categorical features like Outlet_Size and Item_Type for patterns.")
```



Dataset Overview:

	Item_Identifier	Item_Weight	Item_Fat_Content	Item_Visibility	\
0	FDA15	9.30	Low Fat	0.016047	
1	DRC01	5.92	Regular	0.019278	
2	FDN15	17.50	Low Fat	0.016760	
3	FDX07	19.20	Regular	0.000000	
4	NCD19	8.93	Low Fat	0.000000	

	Item_Type	Item_MRP	Outlet_Identifier	\
0	Dairy	249.8092	OUT049	
1	Soft Drinks	48.2692	OUT018	
2	Meat	141.6180	OUT049	
3	Fruits and Vegetables	182.0950	OUT010	
4	Household	53.8614	OUT013	

	Outlet_Establishment_Year	Outlet_Size	Outlet_Location_Type	\
0	1999	Medium	Tier 1	
1	2009	Medium	Tier 3	
2	1999	Medium	Tier 1	
3	1998	NaN	Tier 3	
4	1987	High	Tier 3	

	Outlet_Type	Item_Outlet_Sales
0	Supermarket Type1	3735.1380
1	Supermarket Type2	443.4228
2	Supermarket Type1	2097.2700
3	Grocery Store	732.3800
4	Supermarket Type1	994.7052

Summary Statistics:

	Item_Weight	Item_Visibility	Item_MRP	Outlet_Establishment_Year	\
count	7060.000000	8523.000000	8523.000000	8523.000000	
mean	12.857645	0.066132	140.992782	1997.831867	
std	4.643456	0.051598	62.275067	8.371760	
min	4.555000	0.000000	31.290000	1985.000000	
25%	8.773750	0.026989	93.826500	1987.000000	
50%	12.600000	0.053931	143.012800	1999.000000	
75%	16.850000	0.094585	185.643700	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

Dataset Info:

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 8523 entries, 0 to 8522

Data columns (total 12 columns):

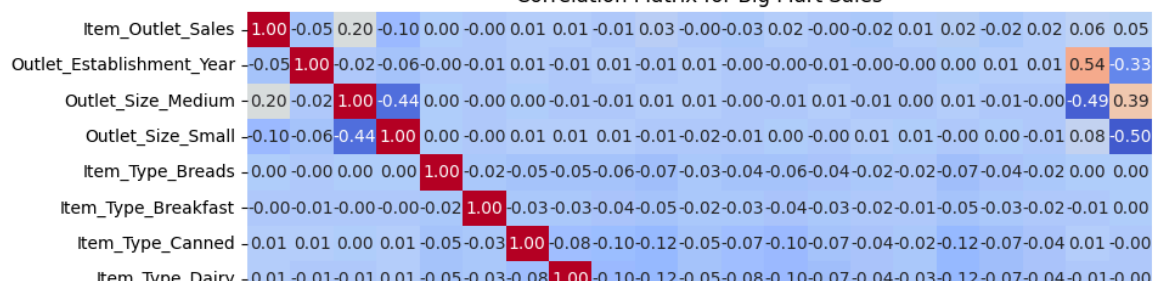
#	Column	Non-Null Count	Dtype
0	Item_Identifier	8523 non-null	object
1	Item_Weight	7060 non-null	float64
2	Item_Fat_Content	8523 non-null	object
3	Item_Visibility	8523 non-null	float64
4	Item_Type	8523 non-null	object
5	Item_MRP	8523 non-null	float64
6	Outlet_Identifier	8523 non-null	object
7	Outlet_Establishment_Year	8523 non-null	int64
8	Outlet_Size	6113 non-null	object
9	Outlet_Location_Type	8523 non-null	object
10	Outlet_Type	8523 non-null	object
11	Item_Outlet_Sales	8523 non-null	float64

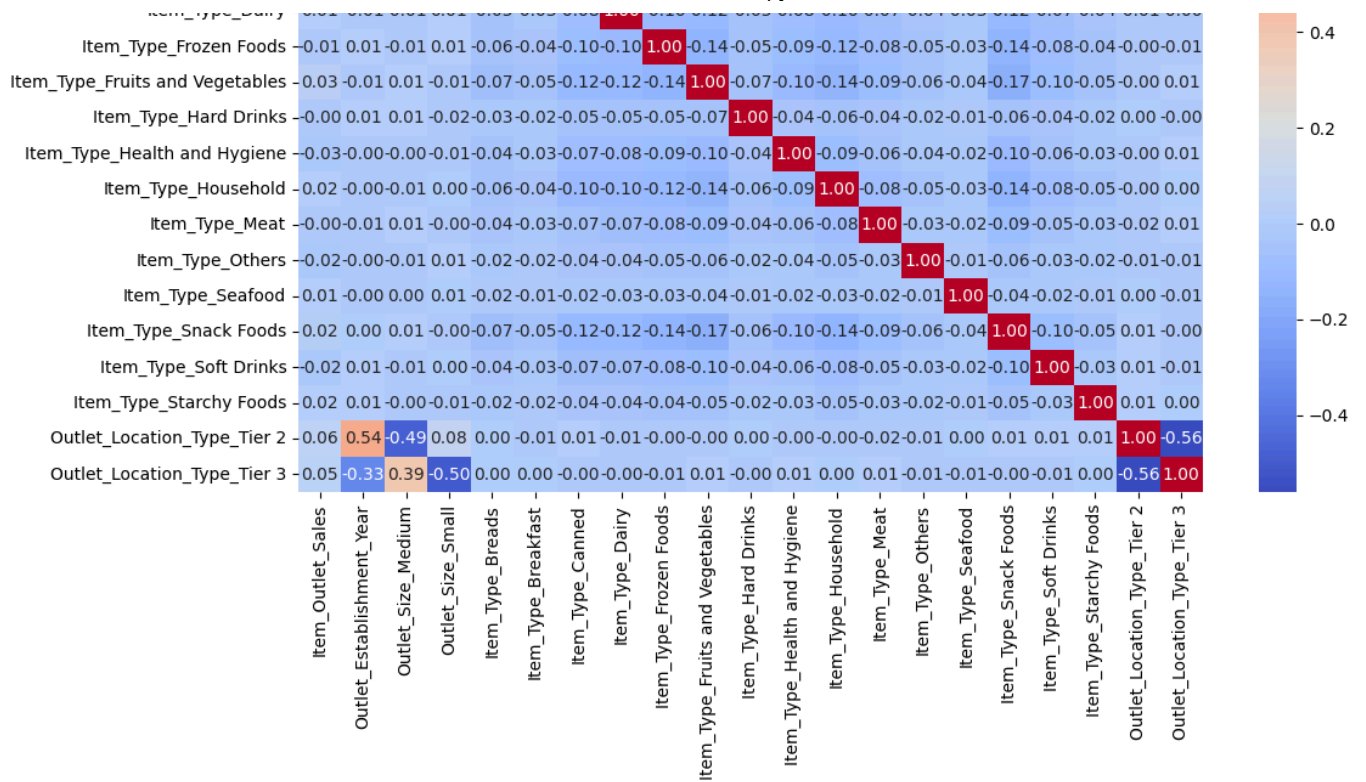
dtypes: float64(4), int64(1), object(7)

memory usage: 799.2+ KB

None

Correlation Matrix for Big Mart Sales





Strong Correlations with Sales:

Item_Outlet_Sales 1.0

Name: Item_Outlet_Sales, dtype: float64