

Task 1

January 8, 2025

0.1 Data Importing

```
[1]: # Import Libraries
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt

# Import Data
transaction_data = pd.read_csv('QVI_transaction_data.csv')
transaction_data.head()
```

```
[1]:
```

	DATE	STORE_NBR	LYLTY_CARD_NBR	TXN_ID	PROD_NBR	\
0	43390	1	1000	1	5	
1	43599	1	1307	348	66	
2	43605	1	1343	383	61	
3	43329	2	2373	974	69	
4	43330	2	2426	1038	108	

	PROD_NAME	PROD_QTY	TOT_SALES
0	Natural Chip Compny SeaSalt175g	2	6.0
1	CCs Nacho Cheese 175g	3	6.3
2	Smiths Crinkle Cut Chips Chicken 170g	2	2.9
3	Smiths Chip Thinly S/Cream&Onion 175g	5	15.0
4	Kettle Tortilla ChpsHny&Jlpno Chili 150g	3	13.8

```
[2]: # Import Data
purchase_behaviour = pd.read_csv('QVI_purchase_behaviour.csv')
purchase_behaviour.head()
```

```
[2]:
```

	LYLTY_CARD_NBR	LIFESTAGE	PREMIUM_CUSTOMER
0	1000	YOUNG SINGLES/COUPLES	Premium
1	1002	YOUNG SINGLES/COUPLES	Mainstream
2	1003	YOUNG FAMILIES	Budget
3	1004	OLDER SINGLES/COUPLES	Mainstream
4	1005	MIDAGE SINGLES/COUPLES	Mainstream

0.2 High-Level Summary

```
[3]: # Summary Statistics
transaction_data.describe()
```

```
[3]:
```

	DATE	STORE_NBR	LYLTY_CARD_NBR	TXN_ID \
count	264836.000000	264836.00000	2.648360e+05	2.648360e+05
mean	43464.036260	135.08011	1.355495e+05	1.351583e+05
std	105.389282	76.78418	8.057998e+04	7.813303e+04
min	43282.000000	1.00000	1.000000e+03	1.000000e+00
25%	43373.000000	70.00000	7.002100e+04	6.760150e+04
50%	43464.000000	130.00000	1.303575e+05	1.351375e+05
75%	43555.000000	203.00000	2.030942e+05	2.027012e+05
max	43646.000000	272.00000	2.373711e+06	2.415841e+06

	PROD_NBR	PROD_QTY	TOT_SALES
count	264836.000000	264836.000000	264836.000000
mean	56.583157	1.907309	7.304200
std	32.826638	0.643654	3.083226
min	1.000000	1.000000	1.500000
25%	28.000000	2.000000	5.400000
50%	56.000000	2.000000	7.400000
75%	85.000000	2.000000	9.200000
max	114.000000	200.000000	650.000000

```
[4]: # Summary Statistics
purchase_behaviour.describe()
```

```
[4]:
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	LYLTY_CARD_NBR
count	7.263700e+04
mean	1.361859e+05
std	8.989293e+04
min	1.000000e+03
25%	6.620200e+04
50%	1.340400e+05
75%	2.033750e+05
max	2.373711e+06

0.3 Data Cleaning and Format Checks

```
[5]: # Check for Missing Values
print(transaction_data.isnull().sum())
print(purchase_behaviour.isnull().sum())
```

```
DATE          0
STORE_NBR     0
LYLTY_CARD_NBR 0
TXN_ID        0
```

```

PROD_NBR          0
PROD_NAME         0
PROD_QTY          0
TOT_SALES         0
dtype: int64
LYLTY_CARD_NBR    0
LIFESTAGE         0
PREMIUM_CUSTOMER  0
dtype: int64

```

```

[6]: # Check Data Type
print(transaction_data.dtypes)
print(purchase_behaviour.dtypes)

```

```

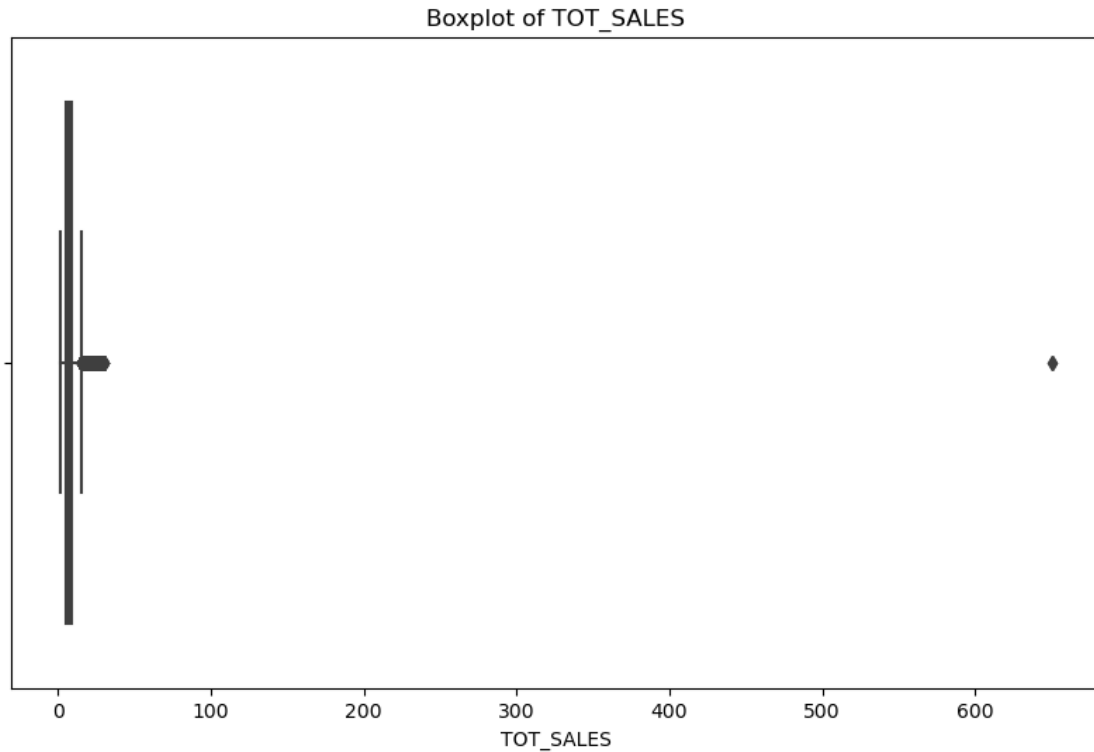
DATE              int64
STORE_NBR         int64
LYLTY_CARD_NBR    int64
TXN_ID            int64
PROD_NBR          int64
PROD_NAME         object
PROD_QTY          int64
TOT_SALES         float64
dtype: object
LYLTY_CARD_NBR    int64
LIFESTAGE         object
PREMIUM_CUSTOMER  object
dtype: object

```

```

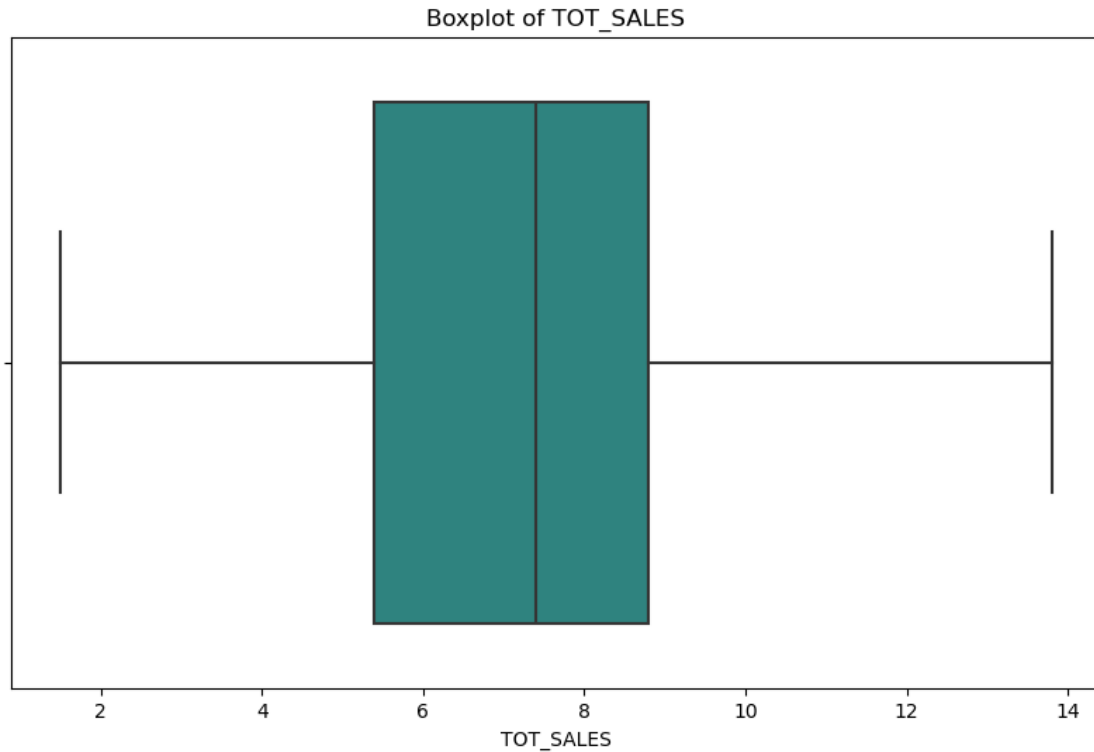
[47]: # Check for Outliers
plt.figure(figsize=(10, 6))
sns.boxplot(x=transaction_data['TOT_SALES'])
plt.title('Boxplot of TOT_SALES')
plt.xlabel('TOT_SALES')
plt.show()

```



```
[75]: # Remove Outliers
Q1 = filtered_transaction_data['TOT_SALES'].quantile(0.25)
Q3 = filtered_transaction_data['TOT_SALES'].quantile(0.75)
IQR = Q3 - Q1
lower_bound = Q1 - (1.5 * IQR)
upper_bound = Q3 + (1.5 * IQR)
filtered_transaction_data =
    ↪ filtered_transaction_data[(filtered_transaction_data['TOT_SALES'] >
    ↪ lower_bound) & (filtered_transaction_data['TOT_SALES'] < upper_bound)]

plt.figure(figsize=(10, 6))
sns.boxplot(x=filtered_transaction_data['TOT_SALES'], palette='viridis')
plt.title('Boxplot of TOT_SALES')
plt.xlabel('TOT_SALES')
plt.show()
```



0.4 Data Analysis and Data Visualization

```
[61]: # Extract pack size
filtered_transaction_data['PACK_SIZE'] = filtered_transaction_data['PROD_NAME'].
    ↪str.extract('(\d+)').astype(float)

# Extract brand name
filtered_transaction_data['BRAND'] = filtered_transaction_data['PROD_NAME'].str.
    ↪split().str[0]

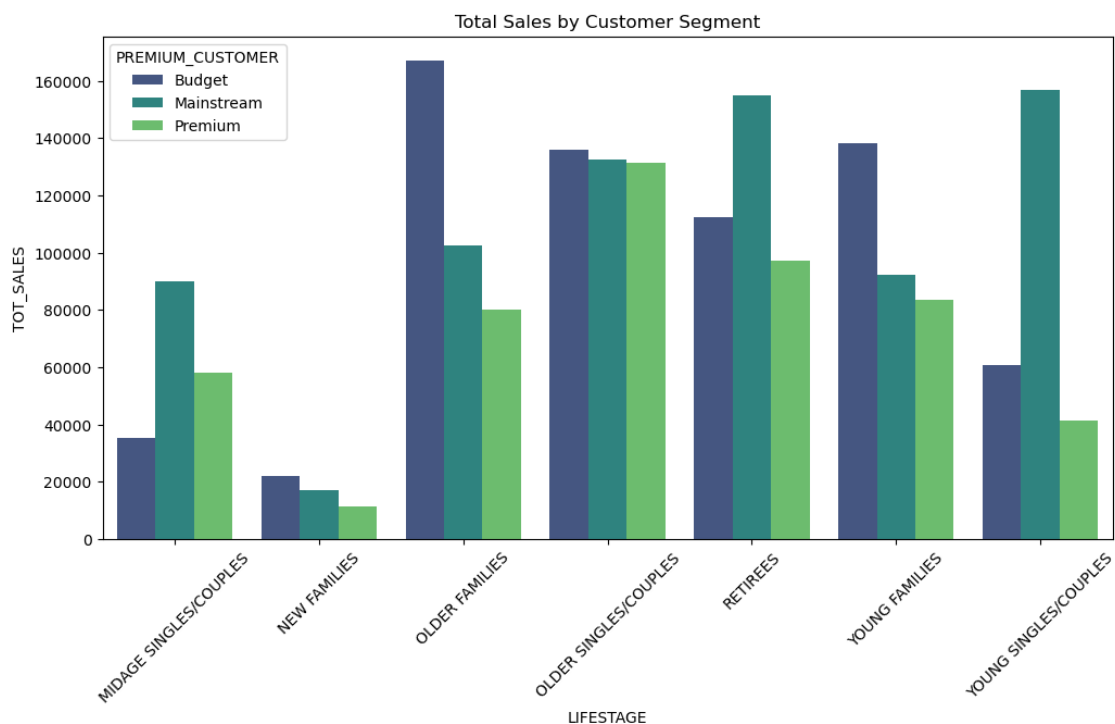
filtered_transaction_data[['PACK_SIZE', 'BRAND']].head()
```

```
[61]:  PACK_SIZE  BRAND
0      175.0  Natural
1      175.0    CCs
2      170.0  Smiths
4      150.0  Kettle
5      300.0    Old
```

```
[67]: # Merge transaction data with purchase behavior data
merged_data = filtered_transaction_data.merge(purchase_behaviour,
    ↪on='LYLTY_CARD_NBR', how='left')
```

```
[77]: # Summarize Customer Segments
segment_summary = merged_data.groupby(['LIFESTAGE', 'PREMIUM_CUSTOMER']).agg({
    'TOT_SALES': 'sum',
    'PROD_QTY': 'sum',
    'TXN_ID': 'nunique'
}).rename(columns={'TXN_ID': 'UNIQUE_TRANSACTIONS'})
segment_summary['AVG_SPEND_PER_TRANSACTION'] = (
    segment_summary['TOT_SALES'] / segment_summary['UNIQUE_TRANSACTIONS']
)
segment_summary = segment_summary.reset_index()

# Plot total sales by customer segment
plt.figure(figsize=(12, 6))
sns.barplot(data=segment_summary, x='LIFESTAGE', y='TOT_SALES',
            hue='PREMIUM_CUSTOMER', palette='viridis')
plt.title("Total Sales by Customer Segment")
plt.xticks(rotation=45)
plt.show()
```



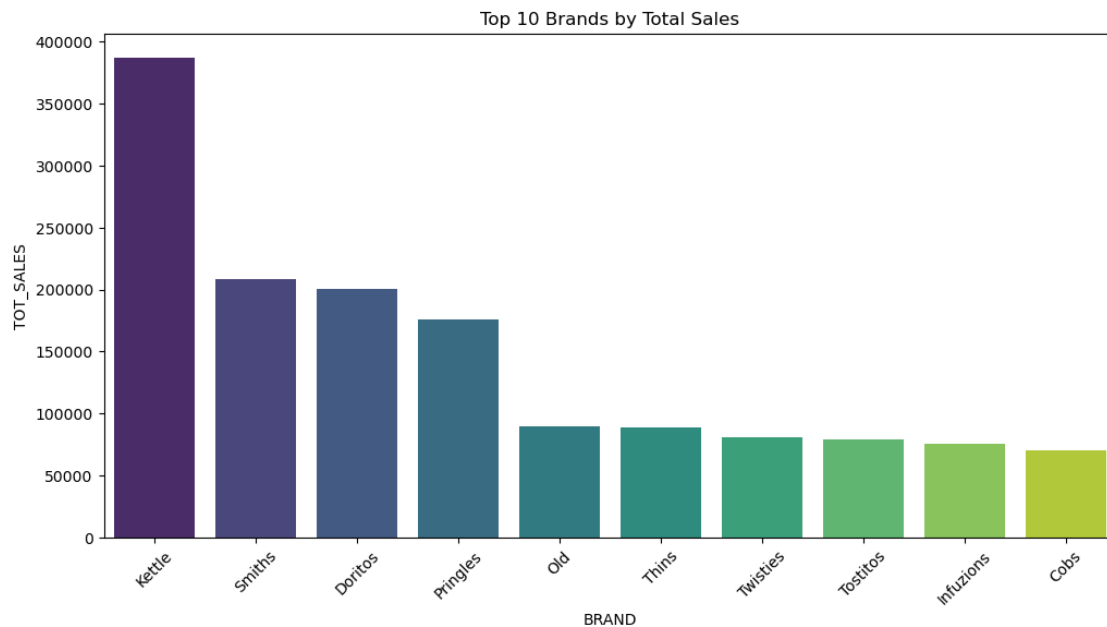
```
[71]: # Identify Top Brands
top_brands = merged_data.groupby('BRAND').agg({'TOT_SALES': 'sum'}).sort_values(
    by='TOT_SALES', ascending=False
).head(10)
```

```

top_brands = top_brands.reset_index()

# Plot top brands
plt.figure(figsize=(12, 6))
sns.barplot(data=top_brands, x='BRAND', y='TOT_SALES', palette='viridis')
plt.title("Top 10 Brands by Total Sales")
plt.xticks(rotation=45)
plt.show()

```



```

[73]: # Analyze Pack Size Popularity
pack_size_summary = merged_data.groupby('PACK_SIZE').agg({'TOT_SALES': 'sum'}).
    ↪sort_index()
pack_size_summary = pack_size_summary.reset_index()

# Plot pack size popularity
plt.figure(figsize=(12, 6))
sns.barplot(data=pack_size_summary, x='PACK_SIZE', y='TOT_SALES',
    ↪palette='viridis')
plt.title("Pack Size Popularity by Total Sales")
plt.xticks(rotation=45)
plt.show()

```

