notebook-retail-analysis

July 9, 2024

[2]:
```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from scipy import stats
import numpy as np
from datetime import datetime, timedelta

# Load data
purchase_behaviour_data = pd.read_csv('QVI_purchase_behaviour.csv')
transaction_data = pd.read_excel('QVI_transaction_data.xlsx', sheet_name='in')

# Function to filter out salsa products
def filter_salsa_products(data):
    salsa_keywords = ['salsa']
    return data[~data['PROD_NAME'].str.contains('|'.join(salsa_keywords),␣
 ↪case=False, na=False)]

# Filter out salsa products
transaction_data = filter_salsa_products(transaction_data)

# Function to identify outliers using Z-score
def z_score_outliers(data, threshold=3):
    z_scores = np.abs(stats.zscore(data.select_dtypes(include=np.number)))
    outliers = (z_scores > threshold).any(axis=1)
    return data[outliers]

# Identify outliers
purchase_behaviour_data_outliers = z_score_outliers(purchase_behaviour_data)
transaction_data_outliers = z_score_outliers(transaction_data)

# Function to detect outliers using IQR
def detect_outliers_iqr(data, feature):
    Q1 = data[feature].quantile(0.25)
    Q3 = data[feature].quantile(0.75)
    IQR = Q3 - Q1
    lower_bound = Q1 - 1.5 * IQR
    upper_bound = Q3 + 1.5 * IQR
```

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    outliers = data[(data[feature] < lower_bound) | (data[feature] >␣
 ↪upper_bound)]
    return outliers

# Apply the function to 'PROD_QTY' and 'TOT_SALES'
prod_qty_outliers = detect_outliers_iqr(transaction_data, 'PROD_QTY')
tot_sales_outliers = detect_outliers_iqr(transaction_data, 'TOT_SALES')

# Convert Excel serial date to datetime
def excel_date_to_datetime(excel_serial_date):
    excel_epoch = datetime(1899, 12, 30)  # Excel's epoch start (not 1900 due␣
 ↪to a historical error)
    delta = timedelta(days=excel_serial_date)
    return excel_epoch + delta

transaction_data['DATE'] = transaction_data['DATE'].
 ↪apply(excel_date_to_datetime)

# Plotting the distribution of high total sales over time
plt.figure(figsize=(10, 5))
plt.scatter(transaction_data.loc[transaction_data['TOT_SALES'] > 15, 'DATE'],
            transaction_data.loc[transaction_data['TOT_SALES'] > 15,␣
 ↪'TOT_SALES'], color='red')
plt.title('Distribution of High Total Sales Over Time')
plt.xlabel('Date')
plt.ylabel('Total Sales')
plt.ylim(0, 700)  # Adjusting the Y-axis limit for clarity
plt.show()

# Display the top 10 highest total sales
top_10_highest_sales = transaction_data.nlargest(10, 'TOT_SALES')
print("Top 10 Highest Total Sales:")
print(top_10_highest_sales)

# Display the top 10 lowest total sales
top_10_lowest_sales = transaction_data.nsmallest(10, 'TOT_SALES')
print("\nTop 10 Lowest Total Sales:")
print(top_10_lowest_sales)

# Feature Engineering
transaction_data['PACK_SIZE'] = transaction_data['PROD_NAME'].str.
 ↪extract('(\d+)g')
transaction_data['BRAND_NAME'] = transaction_data['PROD_NAME'].str.split().
 ↪str[0]

# Mapping of variations to standard brand names
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brand_mapping = {
    'Red': 'RRD',
    'RRD': 'RRD',  # Ensures 'RRD' stays as 'RRD'
    'Snbts': 'Sunbites',
    'Sunbites': 'Sunbites',  # Ensures 'Sunbites' stays as 'Sunbites'
    'Infuzions': 'Infzns',
    'Infzns': 'Infzns',  # Ensures 'Infzns' stays as 'Infzns'
    'WW': 'Woolworths',
    'Woolworths': 'Woolworths',  # Ensures 'Woolworths' stays as 'Woolworths'
    'Smiths': 'Smith',
    'Smith': 'Smith',  # Ensures 'Smith' stays as 'Smith'
    'NCC': 'Natural',
    'Natural': 'Natural',  # Ensures 'Natural' stays as 'Natural'
    'Dorito': 'Doritos',
    'Doritos': 'Doritos',  # Ensures 'Doritos' stays as 'Doritos'
    'Grain': 'GrnWves',
    'GrnWves': 'GrnWves'  # Ensures 'GrnWves' stays as 'GrnWves'
}

# Apply the mapping to standardize brand names
transaction_data['BRAND_NAME'] = transaction_data['BRAND_NAME'].
 ↪replace(brand_mapping)

# Display the unique brands to ensure extraction and standardization worked␣
 ↪correctly
print(transaction_data['BRAND_NAME'].unique())

# Display the first few rows to verify
print(transaction_data[['PACK_SIZE', 'BRAND_NAME']].head())

# Remove Outliers
transaction_data_clean = transaction_data[~transaction_data.index.
 ↪isin(prod_qty_outliers.index)]
transaction_data_clean = transaction_data_clean[~transaction_data_clean.index.
 ↪isin(tot_sales_outliers.index)]

# Re-plotting without extreme outliers
plt.figure(figsize=(10, 5))
plt.scatter(transaction_data_clean['DATE'],
            transaction_data_clean['TOT_SALES'], color='blue')
plt.title('Distribution of Total Sales Over Time (Cleaned)')
plt.xlabel('Date')
plt.ylabel('Total Sales')
plt.show()

# Calculate total spending per customer
```

```
customer_spending = transaction_data_clean.
 ↪groupby('LYLTY_CARD_NBR')['TOT_SALES'].sum().
 ↪reset_index(name='TOTAL_SPENDING')

# Calculate average product quantity per transaction
average_qty = transaction_data_clean.groupby('LYLTY_CARD_NBR')['PROD_QTY'].
 ↪mean().reset_index(name='AVERAGE_QTY')

print(customer_spending.head())
print(average_qty.head())

# Merge customer spending and average quantity data
customer_data = pd.merge(customer_spending, average_qty, on='LYLTY_CARD_NBR')

# Merge with purchase behaviour data
customer_data = pd.merge(customer_data, purchase_behaviour_data,␣
 ↪on='LYLTY_CARD_NBR')

# Analyze total spending by customer segment
segment_spending = customer_data.groupby(['LIFESTAGE',␣
 ↪'PREMIUM_CUSTOMER'])['TOTAL_SPENDING'].sum().reset_index()

# Calculate the number of customers in each segment
customer_segments = customer_data.groupby(['LIFESTAGE',␣
 ↪'PREMIUM_CUSTOMER'])['LYLTY_CARD_NBR'].nunique().
 ↪reset_index(name='CUSTOMER_COUNT')

# Display the number of customers in each segment
print(customer_segments)

print(segment_spending.sort_values(by='TOTAL_SPENDING', ascending=False))

# Calculate total sales per product
product_sales = transaction_data_clean.groupby('PROD_NAME')['TOT_SALES'].sum().
 ↪reset_index(name='TOTAL_SALES')

# Identify top-performing products
top_products = product_sales.sort_values(by='TOTAL_SALES', ascending=False).
 ↪head(10)

print(top_products)

# Resample sales data to monthly frequency and calculate total sales per month
monthly_sales = transaction_data_clean.resample('M', on='DATE')['TOT_SALES'].
 ↪sum().reset_index()
```

```
# Plot monthly sales trends
plt.figure(figsize=(10, 5))
sns.lineplot(data=monthly_sales, x='DATE', y='TOT_SALES')
plt.title('Monthly Sales Trends')
plt.xlabel('Date')
plt.ylabel('Total Sales')
plt.show()

# Merge transaction data with purchase behaviour data to include LIFESTAGE and␣
 ↪PREMIUM_CUSTOMER
merged_data = pd.merge(transaction_data_clean, purchase_behaviour_data,␣
 ↪on='LYLTY_CARD_NBR')

# Group by LIFESTAGE, PREMIUM_CUSTOMER, and PROD_NAME to calculate total sales␣
 ↪for each product within each segment
segment_product_sales = merged_data.groupby(['LIFESTAGE', 'PREMIUM_CUSTOMER',␣
 ↪'PROD_NAME'])['TOT_SALES'].sum().reset_index()

# Function to get top products for each segment
def get_top_products(segment_data, top_n=3):
    return segment_data.sort_values(by='TOT_SALES', ascending=False).head(top_n)

# Apply the function to get the top 3 products for each segment
top_products_per_segment = (
    segment_product_sales.groupby(['LIFESTAGE', 'PREMIUM_CUSTOMER'],␣
 ↪group_keys=False)
    .apply(get_top_products)
    .reset_index(drop=True)
)

# Set display options to show all rows
pd.set_option('display.max_rows', None)

# Print the top products per segment
print(top_products_per_segment)

# Calculate the number of chips bought per customer by segment
chips_per_customer_segment = (
    merged_data.groupby(['LIFESTAGE', 'PREMIUM_CUSTOMER',␣
 ↪'LYLTY_CARD_NBR'])['PROD_QTY'].sum()
    .groupby(level=[0, 1]).mean().reset_index(name='AVG_QTY_PER_CUSTOMER')
)

# Display the number of chips bought per customer by segment
print(chips_per_customer_segment)
```

```
# Calculate the total amount spent on chips by each customer
customer_total_spent = transaction_data_clean.
 ↪groupby('LYLTY_CARD_NBR')['TOT_SALES'].sum().reset_index(name='TOTAL_SPENT')

# Calculate the total number of chips bought by each customer
customer_total_qty = transaction_data_clean.
 ↪groupby('LYLTY_CARD_NBR')['PROD_QTY'].sum().reset_index(name='TOTAL_QTY')

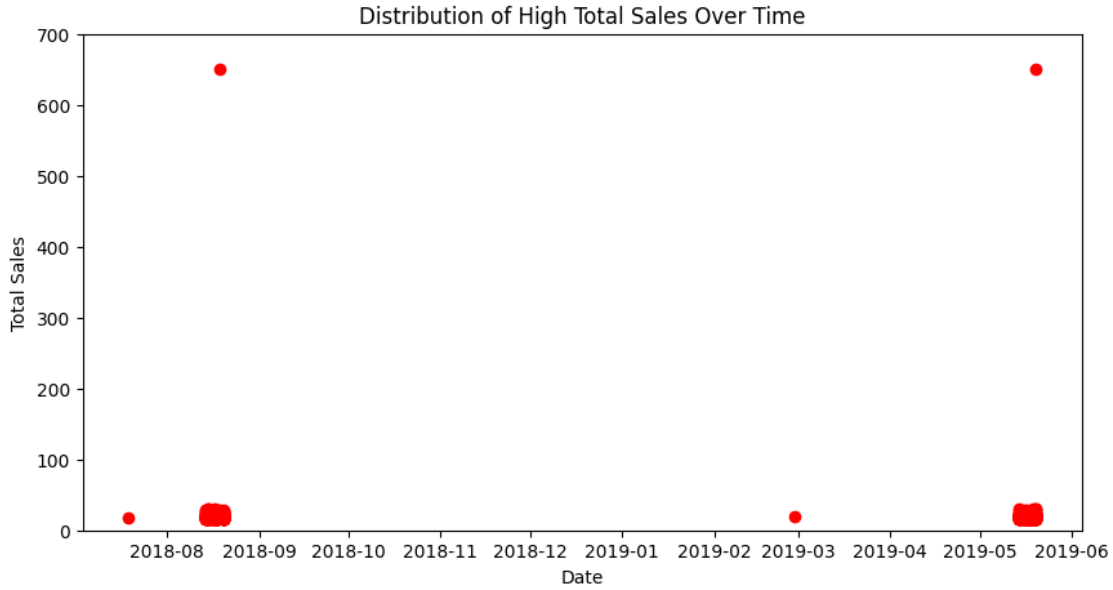
# Merge the total spending and total quantity data
customer_spending_qty = pd.merge(customer_total_spent, customer_total_qty,␣
 ↪on='LYLTY_CARD_NBR')

# Calculate the average chip price per customer
customer_spending_qty['AVG_CHIP_PRICE'] = customer_spending_qty['TOTAL_SPENT'] /
 ↪ customer_spending_qty['TOTAL_QTY']

# Merge with purchase behaviour data to get customer segments
customer_segment_data = pd.merge(customer_spending_qty,␣
 ↪purchase_behaviour_data, on='LYLTY_CARD_NBR')

# Calculate the average chip price by customer segment
avg_chip_price_segment = customer_segment_data.groupby(['LIFESTAGE',␣
 ↪'PREMIUM_CUSTOMER'])['AVG_CHIP_PRICE'].mean().reset_index()

# Display the average chip price by customer segment
print(avg_chip_price_segment)
```



Top 10 Highest Total Sales:

	DATE	STORE_NBR	LYLTY_CARD_NBR	TXN_ID	PROD_NBR	\
69762	2018-08-19	226	226000	226201	4	
69763	2019-05-20	226	226000	226210	4	
5179	2018-08-15	94	94148	93390	14	
55558	2019-05-14	190	190113	190914	14	
69496	2018-08-15	49	49303	45789	14	
117850	2019-05-19	194	194308	194516	14	
150683	2019-05-20	118	118021	120799	14	
171815	2018-08-17	24	24095	20797	14	
184969	2019-05-20	44	44350	40394	14	
72	2018-08-19	96	96203	96025	7	

	PROD_NAME	PROD_QTY	TOT_SALES
69762	Dorito Corn Chp Supreme 380g	200	650.0
69763	Dorito Corn Chp Supreme 380g	200	650.0
5179	Smiths Crnkle Chip Orgnl Big Bag 380g	5	29.5
55558	Smiths Crnkle Chip Orgnl Big Bag 380g	5	29.5
69496	Smiths Crnkle Chip Orgnl Big Bag 380g	5	29.5
117850	Smiths Crnkle Chip Orgnl Big Bag 380g	5	29.5
150683	Smiths Crnkle Chip Orgnl Big Bag 380g	5	29.5
171815	Smiths Crnkle Chip Orgnl Big Bag 380g	5	29.5
184969	Smiths Crnkle Chip Orgnl Big Bag 380g	5	29.5
72	Smiths Crinkle Original 330g	5	28.5

Top 10 Lowest Total Sales:

	DATE	STORE_NBR	LYLTY_CARD_NBR	TXN_ID	PROD_NBR	\
13	2018-08-17	13	13213	12447	92	
145	2019-05-15	197	197172	197097	72	
173	2019-05-19	236	236247	240056	92	
215	2018-10-18	1	1411	476	92	
224	2018-12-14	2	2256	866	55	
525	2018-12-18	20	20311	17286	55	
981	2019-03-05	50	50034	46157	95	
1694	2019-06-15	90	90016	88697	72	
2619	2018-08-21	131	131211	135532	92	
2684	2018-11-07	136	136066	138453	92	

	PROD_NAME	PROD_QTY	TOT_SALES
13	WW Crinkle Cut Chicken 175g	1	1.7
145	WW Crinkle Cut Original 175g	1	1.7
173	WW Crinkle Cut Chicken 175g	1	1.7
215	WW Crinkle Cut Chicken 175g	1	1.7
224	Snbts Whlgrn Crisps Cheddr&Mstrd 90g	1	1.7
525	Snbts Whlgrn Crisps Cheddr&Mstrd 90g	1	1.7
981	Sunbites Whlegrn Crisps Frch/Onin 90g	1	1.7
1694	WW Crinkle Cut Original 175g	1	1.7
2619	WW Crinkle Cut Chicken 175g	1	1.7

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2684          WW Crinkle Cut      Chicken 175g         1        1.7
['Natural' 'CCs' 'Smith' 'Kettle' 'GrnWves' 'Doritos' 'Twisties'
 'Woolworths' 'Thins' 'Burger' 'Cheezels' 'Infzns' 'RRD' 'Pringles'
 'Tyrrells' 'Cobs' 'French' 'Tostitos' 'Cheetos' 'Sunbites']
  PACK_SIZE BRAND_NAME
0       175    Natural
1       175        CCs
2       170      Smith
3       175      Smith
4       150     Kettle
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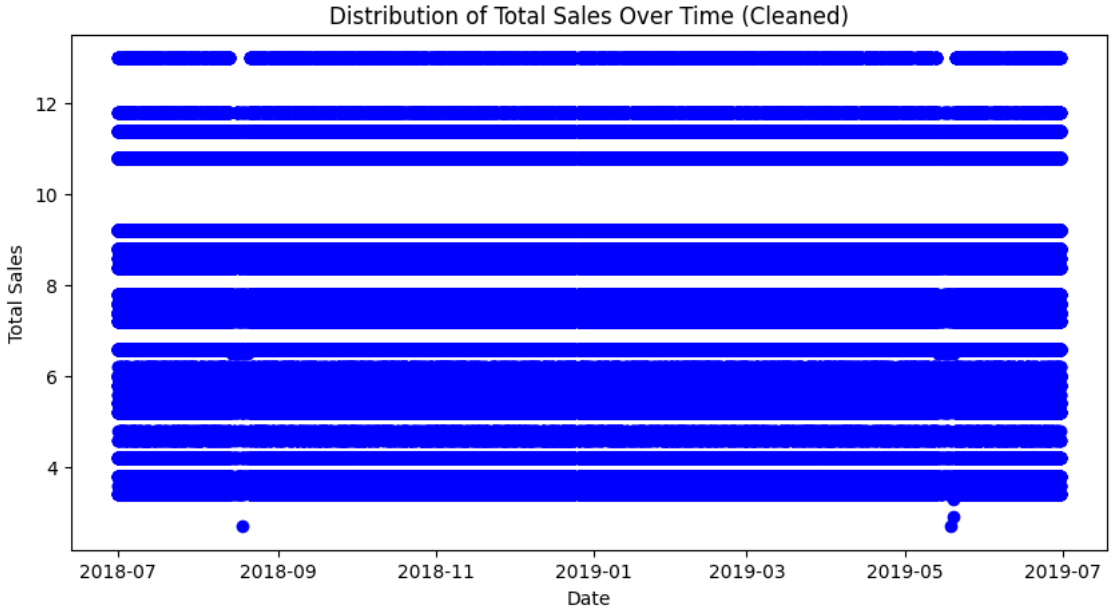


```
   LYLTY_CARD_NBR  TOTAL_SPENDING
0            1000             6.0
1            1010             8.8
2            1011             6.2
3            1013             4.2
4            1025             6.0
   LYLTY_CARD_NBR  AVERAGE_QTY
0            1000          2.0
1            1010          2.0
2            1011          2.0
3            1013          2.0
4            1025          2.0
               LIFESTAGE PREMIUM_CUSTOMER  CUSTOMER_COUNT
0   MIDAGE SINGLES/COUPLES          Budget            1232
1   MIDAGE SINGLES/COUPLES      Mainstream            2918
2   MIDAGE SINGLES/COUPLES         Premium            1997
```

3	NEW FAMILIES	Budget	911
4	NEW FAMILIES	Mainstream	701
5	NEW FAMILIES	Premium	495
6	OLDER FAMILIES	Budget	4148
7	OLDER FAMILIES	Mainstream	2511
8	OLDER FAMILIES	Premium	2015
9	OLDER SINGLES/COUPLES	Budget	4325
10	OLDER SINGLES/COUPLES	Mainstream	4271
11	OLDER SINGLES/COUPLES	Premium	4165
12	RETIREES	Budget	3817
13	RETIREES	Mainstream	5475
14	RETIREES	Premium	3372
15	YOUNG FAMILIES	Budget	3545
16	YOUNG FAMILIES	Mainstream	2386
17	YOUNG FAMILIES	Premium	2124
18	YOUNG SINGLES/COUPLES	Budget	2795
19	YOUNG SINGLES/COUPLES	Mainstream	6669
20	YOUNG SINGLES/COUPLES	Premium	1896

	LIFESTAGE	PREMIUM_CUSTOMER	TOTAL_SPENDING
6	OLDER FAMILIES	Budget	150009.9
13	RETIREES	Mainstream	135028.2
19	YOUNG SINGLES/COUPLES	Mainstream	134727.4
15	YOUNG FAMILIES	Budget	123723.0
9	OLDER SINGLES/COUPLES	Budget	120494.4
10	OLDER SINGLES/COUPLES	Mainstream	117243.2
11	OLDER SINGLES/COUPLES	Premium	116249.8
12	RETIREES	Budget	98685.1
7	OLDER FAMILIES	Mainstream	92179.3
14	RETIREES	Premium	85324.0
16	YOUNG FAMILIES	Mainstream	81967.9
1	MIDAGE SINGLES/COUPLES	Mainstream	79566.1
17	YOUNG FAMILIES	Premium	74650.9
8	OLDER FAMILIES	Premium	71766.1
18	YOUNG SINGLES/COUPLES	Budget	50729.4
2	MIDAGE SINGLES/COUPLES	Premium	50664.6
20	YOUNG SINGLES/COUPLES	Premium	34708.6
0	MIDAGE SINGLES/COUPLES	Budget	31061.9
3	NEW FAMILIES	Budget	18899.0
4	NEW FAMILIES	Mainstream	14711.3
5	NEW FAMILIES	Premium	9935.0

	PROD_NAME	TOTAL_SALES
11	Dorito Corn Chp Supreme 380g	37232.0
79	Smiths Crnkle Chip Orgnl Big Bag 380g	33901.4
71	Smiths Crinkle Chips Salt & Vinegar 330g	32809.2
31	Kettle Mozzarella Basil & Pesto 175g	32443.2
70	Smiths Crinkle Original 330g	32398.8
6	Cheezels Cheese 330g	32045.4
12	Doritos Cheese Supreme 330g	31441.2

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37   Kettle Sweet Chilli And Sour Cream 175g      31158.0
33      Kettle Sea Salt     And Vinegar 175g      30682.8
30          Kettle Honey Soy    Chicken 175g      30682.8
```
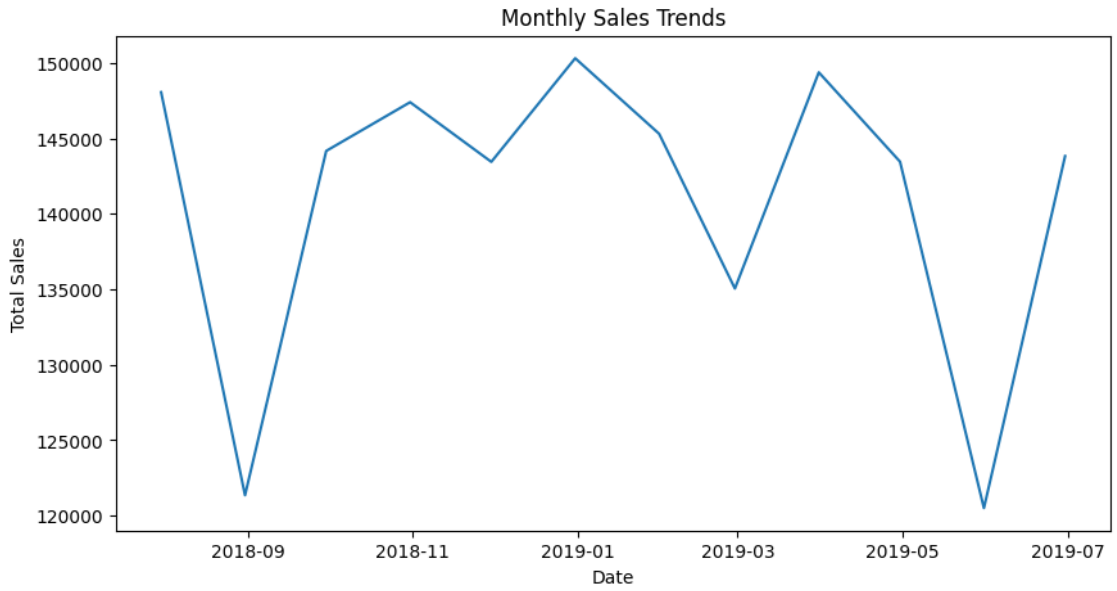

```
/tmp/ipykernel_3111/2346987776.py:153: FutureWarning: 'M' is deprecated and will
be removed in a future version, please use 'ME' instead.
  monthly_sales = transaction_data_clean.resample('M',
on='DATE')['TOT_SALES'].sum().reset_index()
```




```
                LIFESTAGE PREMIUM_CUSTOMER  \
0   MIDAGE SINGLES/COUPLES           Budget
1   MIDAGE SINGLES/COUPLES           Budget
2   MIDAGE SINGLES/COUPLES           Budget
3   MIDAGE SINGLES/COUPLES       Mainstream
4   MIDAGE SINGLES/COUPLES       Mainstream
5   MIDAGE SINGLES/COUPLES       Mainstream
6   MIDAGE SINGLES/COUPLES          Premium
7   MIDAGE SINGLES/COUPLES          Premium
8   MIDAGE SINGLES/COUPLES          Premium
9             NEW FAMILIES           Budget
10            NEW FAMILIES           Budget
11            NEW FAMILIES           Budget
12            NEW FAMILIES       Mainstream
13            NEW FAMILIES       Mainstream
14            NEW FAMILIES       Mainstream
15            NEW FAMILIES          Premium
16            NEW FAMILIES          Premium
17            NEW FAMILIES          Premium
```

18	OLDER FAMILIES	Budget
19	OLDER FAMILIES	Budget
20	OLDER FAMILIES	Budget
21	OLDER FAMILIES	Mainstream
22	OLDER FAMILIES	Mainstream
23	OLDER FAMILIES	Mainstream
24	OLDER FAMILIES	Premium
25	OLDER FAMILIES	Premium
26	OLDER FAMILIES	Premium
27	OLDER SINGLES/COUPLES	Budget
28	OLDER SINGLES/COUPLES	Budget
29	OLDER SINGLES/COUPLES	Budget
30	OLDER SINGLES/COUPLES	Mainstream
31	OLDER SINGLES/COUPLES	Mainstream
32	OLDER SINGLES/COUPLES	Mainstream
33	OLDER SINGLES/COUPLES	Premium
34	OLDER SINGLES/COUPLES	Premium
35	OLDER SINGLES/COUPLES	Premium
36	RETIREES	Budget
37	RETIREES	Budget
38	RETIREES	Budget
39	RETIREES	Mainstream
40	RETIREES	Mainstream
41	RETIREES	Mainstream
42	RETIREES	Premium
43	RETIREES	Premium
44	RETIREES	Premium
45	YOUNG FAMILIES	Budget
46	YOUNG FAMILIES	Budget
47	YOUNG FAMILIES	Budget
48	YOUNG FAMILIES	Mainstream
49	YOUNG FAMILIES	Mainstream
50	YOUNG FAMILIES	Mainstream
51	YOUNG FAMILIES	Premium
52	YOUNG FAMILIES	Premium
53	YOUNG FAMILIES	Premium
54	YOUNG SINGLES/COUPLES	Budget
55	YOUNG SINGLES/COUPLES	Budget
56	YOUNG SINGLES/COUPLES	Budget
57	YOUNG SINGLES/COUPLES	Mainstream
58	YOUNG SINGLES/COUPLES	Mainstream
59	YOUNG SINGLES/COUPLES	Mainstream
60	YOUNG SINGLES/COUPLES	Premium
61	YOUNG SINGLES/COUPLES	Premium
62	YOUNG SINGLES/COUPLES	Premium

	PROD_NAME	TOT_SALES
0	Dorito Corn Chp Supreme 380g	715.0

1	Kettle Mozzarella	Basil & Pesto	175g	637.2
2	Doritos Cheese	Supreme	330g	592.8
3	Smiths Crinkle Chips	Salt & Vinegar	330g	1995.0
4		Cheezels Cheese	330g	1903.8
5	Smiths Crnkle Chip	Orgnl Big Bag	380g	1840.8
6	Dorito Corn Chp	Supreme	380g	1131.0
7	Smiths Crnkle Chip	Orgnl Big Bag	380g	1109.2
8		Cheezels Cheese	330g	1026.0
9	Doritos Cheese	Supreme	330g	467.4
10	Dorito Corn Chp	Supreme	380g	455.0
11	Kettle Honey Soy	Chicken	175g	453.6
12	Dorito Corn Chp	Supreme	380g	370.5
13	Kettle Mozzarella	Basil & Pesto	175g	345.6
14		Kettle Chilli	175g	324.0
15	Smiths Crnkle Chip	Orgnl Big Bag	380g	259.6
16	Doritos Corn Chips	Cheese Supreme	170g	228.8
17	Smiths Crinkle	Original	330g	228.0
18	Dorito Corn Chp	Supreme	380g	3048.5
19	Smiths Crinkle Chips	Salt & Vinegar	330g	3043.8
20		Cheezels Cheese	330g	2918.4
21	Dorito Corn Chp	Supreme	380g	1943.5
22	Smiths Crinkle Chips	Salt & Vinegar	330g	1869.6
23	Smiths Crnkle Chip	Orgnl Big Bag	380g	1746.4
24	Smiths Crinkle	Original	330g	1504.8
25	Dorito Corn Chp	Supreme	380g	1501.5
26	Smiths Crinkle Chips	Salt & Vinegar	330g	1436.4
27	Dorito Corn Chp	Supreme	380g	2600.0
28	Smiths Crnkle Chip	Orgnl Big Bag	380g	2489.8
29	Kettle Sea Salt	And Vinegar	175g	2440.8
30	Dorito Corn Chp	Supreme	380g	2730.0
31	Smiths Crnkle Chip	Orgnl Big Bag	380g	2336.4
32	Doritos Cheese	Supreme	330g	2245.8
33	Dorito Corn Chp	Supreme	380g	2730.0
34		Cheezels Cheese	330g	2485.2
35	Smiths Crinkle	Original	330g	2416.8
36		Kettle Chilli	175g	2311.2
37	Smiths Crnkle Chip	Orgnl Big Bag	380g	2183.0
38	Kettle Mozzarella	Basil & Pesto	175g	2138.4
39	Dorito Corn Chp	Supreme	380g	2886.0
40	Smiths Crinkle	Original	330g	2724.6
41	Kettle Honey Soy	Chicken	175g	2700.0
42	Dorito Corn Chp	Supreme	380g	2145.0
43	Kettle Mozzarella	Basil & Pesto	175g	1879.2
44	Smiths Crnkle Chip	Orgnl Big Bag	380g	1817.2
45	Dorito Corn Chp	Supreme	380g	2574.0
46	Smiths Crinkle	Original	330g	2428.2
47		Kettle Original	175g	2408.4
48	Smiths Crnkle Chip	Orgnl Big Bag	380g	1663.8

```
49          Dorito Corn Chp     Supreme 380g     1612.0
50                      Cheezels Cheese 330g     1527.6
51          Dorito Corn Chp     Supreme 380g     1592.5
52                      Cheezels Cheese 330g     1539.0
53                      Kettle Original 175g     1468.8
54          Dorito Corn Chp     Supreme 380g     1079.0
55          Doritos Cheese      Supreme 330g     1014.6
56      Kettle Sea Salt     And Vinegar 175g     1004.4
57          Dorito Corn Chp     Supreme 380g     3445.0
58    Smiths Crnkle Chip  Orgnl Big Bag 380g     3174.2
59    Kettle Mozzarella   Basil & Pesto 175g     3142.8
60          Dorito Corn Chp     Supreme 380g      754.0
61    Smiths Crnkle Chip  Orgnl Big Bag 380g      731.6
62    Kettle Mozzarella   Basil & Pesto 175g      712.8
                LIFESTAGE PREMIUM_CUSTOMER  AVG_QTY_PER_CUSTOMER
0   MIDAGE SINGLES/COUPLES           Budget              6.689935
1   MIDAGE SINGLES/COUPLES       Mainstream              6.828650
2   MIDAGE SINGLES/COUPLES          Premium              6.685028
3             NEW FAMILIES           Budget              5.251372
4             NEW FAMILIES       Mainstream              5.298146
5             NEW FAMILIES          Premium              5.147475
6           OLDER FAMILIES           Budget              9.641755
7           OLDER FAMILIES       Mainstream              9.824771
8           OLDER FAMILIES          Premium              9.568238
9    OLDER SINGLES/COUPLES           Budget              7.161618
10   OLDER SINGLES/COUPLES       Mainstream              7.168345
11   OLDER SINGLES/COUPLES          Premium              7.149580
12                RETIREES           Budget              6.561174
13                RETIREES       Mainstream              6.378082
14                RETIREES          Premium              6.440688
15          YOUNG FAMILIES           Budget              9.266573
16          YOUNG FAMILIES       Mainstream              9.217100
17          YOUNG FAMILIES          Premium              9.347458
18   YOUNG SINGLES/COUPLES           Budget              4.868694
19   YOUNG SINGLES/COUPLES       Mainstream              4.946169
20   YOUNG SINGLES/COUPLES          Premium              4.900844
                LIFESTAGE PREMIUM_CUSTOMER  AVG_CHIP_PRICE
0   MIDAGE SINGLES/COUPLES           Budget        3.820015
1   MIDAGE SINGLES/COUPLES       Mainstream        4.069664
2   MIDAGE SINGLES/COUPLES          Premium        3.837673
3             NEW FAMILIES           Budget        3.949506
4             NEW FAMILIES       Mainstream        3.964496
5             NEW FAMILIES          Premium        3.902643
6           OLDER FAMILIES           Budget        3.847186
7           OLDER FAMILIES       Mainstream        3.831940
8           OLDER FAMILIES          Premium        3.812338
9    OLDER SINGLES/COUPLES           Budget        3.938227
10   OLDER SINGLES/COUPLES       Mainstream        3.868939
```

11	OLDER SINGLES/COUPLES	Premium	3.969827
12	RETIREES	Budget	3.998533
13	RETIREES	Mainstream	3.895328
14	RETIREES	Premium	3.983547
15	YOUNG FAMILIES	Budget	3.867016
16	YOUNG FAMILIES	Mainstream	3.826046
17	YOUNG FAMILIES	Premium	3.849675
18	YOUNG SINGLES/COUPLES	Budget	3.730527
19	YOUNG SINGLES/COUPLES	Mainstream	4.155596
20	YOUNG SINGLES/COUPLES	Premium	3.732626

```
/tmp/ipykernel_3111/2346987776.py:176: DeprecationWarning:
DataFrameGroupBy.apply operated on the grouping columns. This behavior is
deprecated, and in a future version of pandas the grouping columns will be
excluded from the operation. Either pass `include_groups=False` to exclude the
groupings or explicitly select the grouping columns after groupby to silence
this warning.
  .apply(get_top_products)
```