Name - Bhagyashri Vinit Dumbre

Task-1

 $\ensuremath{\text{\#}}$ Import pandas, matplotlib and seaborn

import pandas as pd

import numpy as np

import matplotlib.pyplot as plt

import seaborn as sns

%matplotlib inline

import first transaction_data

Transaction_data = pd.read_csv("/QVI_transaction_data.csv")

Transaction_data

	DATE	STORE_NBR	LYLTY_CARD_NBR	TXN_ID	PROD_NBR	PROD_NAME	SIZE	PROD_QTY	TOT_SALE
0	10/17/2018	1	1000	1	5	Natural Chip Compny SeaSalt175g	175g	2	6
1	5/14/2019	1	1307	348	66	CCs Nacho Cheese 175g	175g	3	6
2	5/20/2019	1	1343	383	61	Smiths Crinkle Cut Chips Chicken 170g	170g	2	2
3	8/17/2018	2	2373	974	69	Smiths Chip Thinly S/Cream&Onion 175g	175g	5	15
4	8/18/2018	2	2426	1038	108	Kettle Tortilla ChpsHny&Jlpno Chili 150g	150g	3	13

Kettle Sweet

Purchase_behaviour = pd.read_csv("/QVI_purchase_behaviour.csv") Purchase_behaviour

	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
72632	2370651	MIDAGE SINGLES/COUPLES	Mainstream
72633	2370701	YOUNG FAMILIES	Mainstream
72634	2370751	YOUNG FAMILIES	Premium
72635	2370961	OLDER FAMILIES	Budget
72636	2373711	YOUNG SINGLES/COUPLES	Mainstream

72637 rows × 3 columns

Transaction_data .head(10)

	DATE	STORE_NBR	LYLTY_CARD_NBR	TXN_ID	PROD_NBR	PROD_NAME	SIZE	PROD_QTY	TOT_SALES
0	10/17/2018	1	1000	1	5	Natural Chip Compny SeaSalt175g	175g	2	6.0
1	5/14/2019	1	1307	348	66	CCs Nacho Cheese 175g	175g	3	6.3
2	5/20/2019	1	1343	383	61	Smiths Crinkle Cut Chips Chicken 170g	170g	2	2.9
3	8/17/2018	2	2373	974	69	Smiths Chip Thinly S/Cream&Onion 175g	175g	5	15.0
4	8/18/2018	2	2426	1038	108	Kettle Tortilla ChpsHny&Jlpno Chili 150g	150g	3	13.8

Transaction_data .tail(10)

[#] import 2nd purchase behaviour data

		DATE	STORE_NBR	LYLTY_CARD_NBR	TXN_ID	PROD_NBR	PROD_NAME	SIZE	PROD_QTY	TOT_S#
	264826	3/25/2019	272	272194	269908	75	Cobs Popd Sea Salt Chips 110g	110g	2	
	264827	8/28/2018	272	272197	269911	104	Infuzions Thai SweetChili	110g	2	
Purch	ase_behav	viour.head(10)							
	LYLTY	_CARD_NBR		LIFESTAGE	PREMIUM	_CUSTOMER				
	0	1000	YOUNG SIN	NGLES/COUPLES		Premium				

```
1002 YOUNG SINGLES/COUPLES
                                                Mainstream
1
2
            1003
                          YOUNG FAMILIES
                                                    Budget
3
            1004
                  OLDER SINGLES/COUPLES
                                                Mainstream
                                                 Mainstream
                 MIDAGE SINGLES/COUPLES
                  YOUNG SINGLES/COUPLES
5
            1007
                                                    Budget
6
            1009
                            NEW FAMILIES
                                                   Premium
                 YOUNG SINGLES/COUPLES
                                                 Mainstream
7
            1010
8
            1011
                  OLDER SINGLES/COUPLES
                                                 Mainstream
9
            1012
                           OLDER FAMILIES
                                                 Mainstream
```

Transaction_data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 264836 entries, 0 to 264835
Data columns (total 9 columns):
                    Non-Null Count
 # Column
                                    Dtype
 0
    DATE
                    264836 non-null object
    STORE_NBR
                    264836 non-null int64
 1
    LYLTY_CARD_NBR 264836 non-null int64
    TXN_ID
                    264836 non-null int64
    PROD_NBR
                    264836 non-null int64
                    264836 non-null object
    PROD_NAME
 5
                    264836 non-null object
 6
    SIZE
    PROD_QTY
                    264836 non-null int64
    TOT_SALES
                    264836 non-null float64
dtypes: float64(1), int64(5), object(3)
```

Purchase_behaviour.info()

memory usage: 18.2+ MB

Transaction_data.columns

Purchase_behaviour.columns

```
Index(['LYLTY_CARD_NBR', 'LIFESTAGE', 'PREMIUM_CUSTOMER'], dtype='object')
```

Transaction_data.shape

(264836, 9)

Purchase_behaviour.shape

(72637, 3)

 $\mbox{\#}$ merge both data files in order to do findings.

left_merged = pd.merge(Transaction_data, Purchase_behaviour, how="outer", on=["LYLTY_CARD_NBR"])
left_merged

		DATE	STORE_NBR	LYLTY_CARD_NBR	TXN_ID	PROD_NBR	PROD_NAME	SIZE	PROD_QTY	TOT_SALES
	0	10/17/2018	1	1000	1	5	Natural Chip Compny SeaSalt175g	175g	2	6.0
	1	5/14/2019	1	1307	348	66	CCs Nacho Cheese 175g	175g	3	6.3
	2	11/10/2018	1	1307	346	96	WW Original Stacked Chips 160g	160g	2	3.8
	3	3/9/2019	1	1307	347	54	CCs Original 175g	175g	1	2.1
	4	5/20/2019	1	1343	383	61	Smiths Crinkle Cut Chips Chicken 170g	170g	2	2.9
£+ ~	ongod	chano								

 ${\tt left_merged.shape}$

(264836, 11)

to describe data

left_merged1 = left_merged.iloc[:,6:9]
left_merged1.describe()

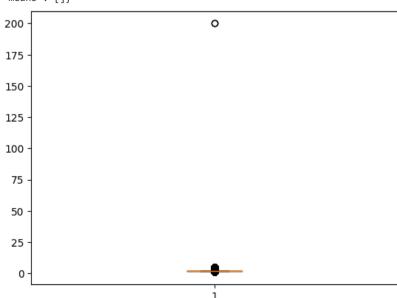
	PROD_QTY	TOT_SALES
count	264836.000000	264836.000000
mean	1.907309	7.304200
std	0.643654	3.083226
min	1.000000	1.500000
25%	2.000000	5.400000
50%	2.000000	7.400000
75%	2.000000	9.200000
max	200.000000	650.000000

type of data left_merged.dtypes

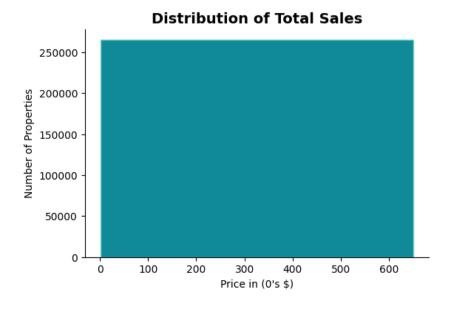
> DATE object STORE_NBR int64 LYLTY_CARD_NBR int64 TXN_ID PROD_NBR int64 int64 PROD_NAME object SIZE object PROD_QTY
> TOT_SALES
> LIFESTAGE int64 float64 object PREMIUM_CUSTOMER object dtype: object

outliers finding in product quantity plt.boxplot(left_merged1['PROD_QTY'],vert = True)

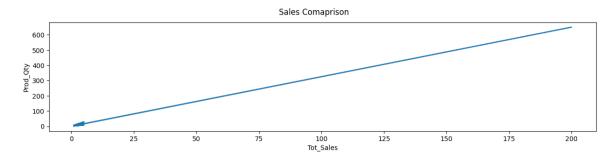
'means': []}



```
# outliers finding in total sales
plt.boxplot(left_merged1['TOT_SALES'],vert = True)
     {'whiskers': [<matplotlib.lines.Line2D at 0x7f266b59cc40>,
      <matplotlib.lines.Line2D at 0x7f266b59cee0>],
      'caps': [<matplotlib.lines.Line2D at 0x7f266b59d180>,
      <matplotlib.lines.Line2D at 0x7f266b59d420>],
      'boxes': [<matplotlib.lines.Line2D at 0x7f266b59c9a0>],
      'medians': [<matplotlib.lines.Line2D at 0x7f266b59d6c0>],
      'fliers': [<matplotlib.lines.Line2D at 0x7f266b59d960>],
      'means': []}
                                          0
      600
      500
      400
      300
      200
      100
        0
```



```
fig = plt.figure(figsize=(15,3))
plt.plot(left_merged1.PROD_QTY, left_merged.TOT_SALES)
plt.xlabel('Tot_Sales')
plt.ylabel('Prod_Qty')
plt.suptitle('Sales Comaprison')
plt.show()
```



from above line graph distribution as total sales increases product qty also increased.

```
# year wise and month wise total sales
left_merged["DATE_YEAR"]= pd.DatetimeIndex(left_merged['DATE']).year
left_merged["DATE_MONTH"]= pd.DatetimeIndex(left_merged['DATE']).month
```

pivot_table = left_merged.pivot_table(index=['DATE_YEAR', 'DATE_MONTH'], aggfunc={'TOT_SALES':sum})
pivot_table.reset_index().sort_values(['DATE_YEAR', 'DATE_MONTH'])

	DATE_YEAR	DATE_MONTH	TOT_SALES
0	2018	7	165275.30
1	2018	8	158731.05
2	2018	9	160522.00
3	2018	10	164415.70
4	2018	11	160233.70
5	2018	12	167913.40
6	2019	1	162642.30
7	2019	2	150665.00
8	2019	3	166265.20
9	2019	4	159845.10
10	2019	5	157367.65
11	2019	6	160538.60

```
# yearly Total Sales
left_merged["DATE_YEAR"]= pd.DatetimeIndex(left_merged['DATE']).year
pivot_table2 = left_merged.pivot_table(index=['DATE_YEAR'], aggfunc={'TOT_SALES':sum})
pivot_table2.reset_index().sort_values(['DATE_YEAR'])
```

	DATE_YEAR	TOT_SALES
0	2018	977091.15
1	2019	957323.85

```
g=sns.relplot(x='DATE_MONTH',
               y='DATE_YEAR',
               data=pivot_table,
               kind = 'scatter',
               size = 'TOT_SALES',
               #style = 'time'
plt.show()
        2019.0
        2018.8
     DATE_YEAR
5.8102
                                                                   TOT_SALES
                                                                       153000
                                                                       156000
                                                                       159000
                                                                       162000
                                                                       165000
        2018.2
        2018.0
                                                             12
                                      6
                                  DATE_MONTH
```

from above scatter plot we find out that in 2018 in 12 th month there were highest sales and # in 2019 in 3rd month there were highest sales

pivot_table3=left_merged.pivot_table(index=['PROD_NAME','SIZE'], aggfunc={'TOT_SALES':sum})
pivot_table3.reset_index().sort_values(['TOT_SALES'], ascending = False)

	PROD_NAME	SIZE	TOT_SALES
11	Dorito Corn Chp Supreme 380g	380g	40352.0
87	Smiths Crnkle Chip Orgnl Big Bag 380g	380g	36367.6
77	Smiths Crinkle Chips Salt & Vinegar 330g	330g	34804.2
33	Kettle Mozzarella Basil & Pesto 175g	175g	34457.4
86	Smiths Crinkle Original 330g	330g	34302.6
91	Sunbites Whlegrn Crisps Frch/Onin 90g	90g	4600.2
106	WW Crinkle Cut Original 175g	175g	4532.2
114	Woolworths Mild Salsa 300g	300g	4234.5
113	Woolworths Medium Salsa 300g	300g	4050.0
46	Natural Chip Compny SeaSalt175g	175g	6.0
115 ro	ws × 3 columns		

 $\mbox{\#}$ from above pivot table it is find that Dorito Corn Chp Supreme 380g $\,$ have highest sales

	PROD_NAME	SIZE	PROD_QTY
11	Dorito Corn Chp Supreme 380g	380g	6509
33	Kettle Mozzarella Basil & Pesto 175g	175g	6381
42	Kettle Tortilla ChpsHny&Jlpno Chili 150g	150g	6309
8	Cobs Popd Sea Salt Chips 110g	110g	6277
10	Cobs Popd Swt/Chlli &Sr/Cream Chips 110g	110g	6256
113	Woolworths Medium Salsa 300g	300g	2700
43	NCC Sour Cream & Garden Chives 175g	175g	2682
106	WW Crinkle Cut Original 175g	175g	2666
21	French Fries Potato Chips 175g	175g	2643

pivot_merged = pd.merge(pivot_table3, pivot_table4, how="outer", on=["PROD_NAME","SIZE"])
pivot_merged
pivot_merged1 = pivot_merged.reset_index().sort_values(['TOT_SALES'], ascending = False)
pivot_merged1

	PROD_NAME	SIZE	TOT_SALES	PROD_QTY
11	Dorito Corn Chp Supreme 380g	380g	40352.0	6509
87	Smiths Crnkle Chip Orgnl Big Bag 380g	380g	36367.6	6164
77	Smiths Crinkle Chips Salt & Vinegar 330g	330g	34804.2	6106
33	Kettle Mozzarella Basil & Pesto 175g	175g	34457.4	6381
86	Smiths Crinkle Original 330g	330g	34302.6	6018
91	Sunbites Whlegrn Crisps Frch/Onin 90g	90g	4600.2	2706
106	WW Crinkle Cut Original 175g	175g	4532.2	2666
114	Woolworths Mild Salsa 300g	300g	4234.5	2823
113	Woolworths Medium Salsa 300g	300g	4050.0	2700
46	Natural Chip Compny SeaSalt175g	175g	6.0	2

115 rows × 4 columns

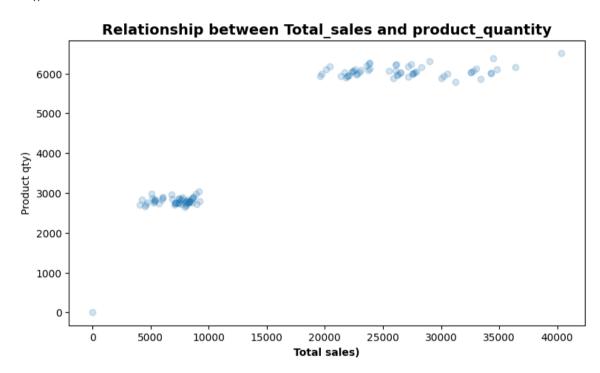
pivot_merged1.head(20)

	PROD_NAME	SIZE	TOT_SALES	PROD_QTY
11	Dorito Corn Chp Supreme 380g	380g	40352.0	6509
87	Smiths Crnkle Chip Orgnl Big Bag 380g	380g	36367.6	6164
77	Smiths Crinkle Chips Salt & Vinegar 330g	330g	34804.2	6106
33	Kettle Mozzarella Basil & Pesto 175g	175g	34457.4	6381
86	Smiths Crinkle Original 330g	330g	34302.6	6018
6	Cheezels Cheese 330g	330g	34296.9	6017
12	Doritos Cheese Supreme 330g	330g	33390.6	5858
39	Kettle Sweet Chilli And Sour Cream 175g	175g	33031.8	6120
34	Kettle Original 175g	175g	32740.2	6064
35	Kettle Sea Salt And Vinegar 175g	175g	32589.0	6035
32	Kettle Honey Soy Chicken 175g	175g	32578.2	6033
31	Kettle Chilli 175g	175g	31271.4	5792
49	Old El Paso Salsa Dip Chnky Tom Ht300g	300g	30513.3	5986
50	Old El Paso Salsa Dip Tomato Med 300g	300g	30237.9	5929
51	Old El Paso Salsa Dip Tomato Mild 300g	300g	30033.9	5890
42	Kettle Tortilla ChpsHny&Jlpno Chili 150g	150g	29021.4	6309
37	Kettle Sensations Camembert & Fig 150g	150g	28308.4	6157
102	Twisties Chicken270g	270g	27853.0	6055
40	Kettle Tortilla ChpsBtroot&Ricotta 150g	150g	27770.2	6037
41	Kettle Tortilla ChpsFeta&Garlic 150g	150g	27627.6	6008

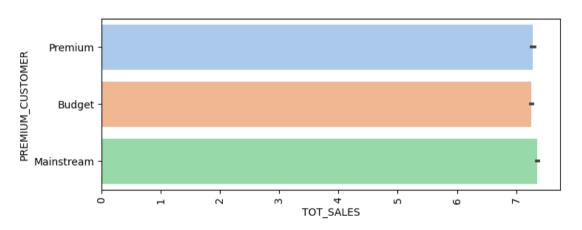
[#] from above pivot table it is find that Dorito Corn Chp Supreme 380g have highest sales with highest prod qty but Kettle Mozzarella Basil & Pesto 175g 175g having lowest sales

[#] Smiths Crnkle Chip Orgnl Big Bag 380g and Smiths Crinkle Chips Salt & Vinegar 330g with higher prod qty.
if we increase the prod qty of Smiths Crnkle Chip Orgnl Big Bag 380g and Smiths Crinkle Chips Salt & Vinegar 330g we will get more sales.

	PROD_NAME	SIZE	TOT_SALES	PROD_QTY
82	Smiths Crinkle Cut French OnionDip 150g	150g	7046.0	2710
26	Infuzions Mango Chutny Papadums 70g	70g	6852.0	2855
0	Burger Rings 220g	220g	6831.0	2970
3	CCs Tasty Cheese 175g	175g	6069.0	2890
2	CCs Original 175g	175g	6048.0	2880
1	CCs Nacho Cheese 175g	175g	5961.9	2839
7	Cheezels Cheese Box 125g	125g	5733.0	2730
111	WW Supreme Cheese Corn Chips 200g	200g	5390.3	2837
108	WW Original Corn Chips 200g	200g	5367.5	2825
109	WW Original Stacked Chips 160g	160g	5323.8	2802
110	WW Sour Cream &OnionStacked Chips 160g	160g	5323.8	2802
107	WW D/Style Chip Sea Salt 200g	200g	5249.7	2763
112	Woolworths Cheese Rings 190g	190g	5169.6	2872
90	Snbts Whlgrn Crisps Cheddr&Mstrd 90g	90g	5076.2	2986
105	WW Crinkle Cut Chicken 175g	175g	4702.2	2766
91	Sunbites Whlegrn Crisps Frch/Onin 90g	90g	4600.2	2706
106	WW Crinkle Cut Original 175g	175g	4532.2	2666
114	Woolworths Mild Salsa 300g	300g	4234.5	2823
113	Woolworths Medium Salsa 300g	300g	4050.0	2700
46	Natural Chip Compny SeaSalt175g	175g	6.0	2



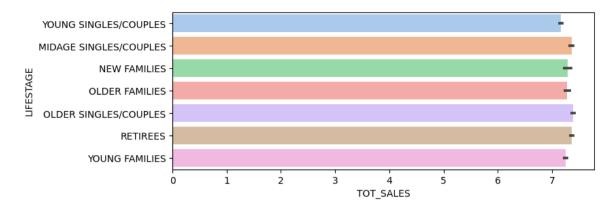
```
plt.figure(figsize = (8,3))
pal = sns.color_palette("pastel")
sns.barplot(x="TOT_SALES",y="PREMIUM_CUSTOMER",data=left_merged, palette=pal)
plt.xticks(rotation = 90, fontsize = 10)
plt.show()
```



from above chart i come to conclusion that Mainstream premium customer gives us more sales.

7. 6. (6. . (6. 3))

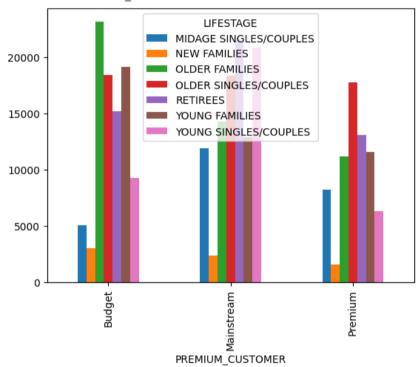
plt.tigure(tigsize = (8,3))
pal = sns.color_palette("pastel")
sns.barplot(x="TOT_SALES",y="LIFESTAGE",data=left_merged, palette=pal)
plt.show()



From above chart i came to know about midage/single couples, older/single couples, retirees get more sales.

pd.crosstab(left_merged.PREMIUM_CUSTOMER,left_merged.LIFESTAGE).plot(kind="bar")

<Axes: xlabel='PREMIUM_CUSTOMER'>



- $\mbox{\tt\#}$ from Budget lifestage Older families and young families give higher sales
- # from Mainstream Retirees, young/single couples give more sales.
- # from Premium Older/single couples , retirees give more sales.

left_merged["PREMIUM_CUSTOMER"].value_counts()
left_merged.PREMIUM_CUSTOMER.value_counts().plot(kind="pie")

<Axes: ylabel='PREMIUM_CUSTOMER'>

