```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import plotly.express as px
```

In [25]: transactions= pd.read_csv('QVI_transaction_data.csv')
behaviour= pd.read_csv('QVI_purchase_behaviour.csv')

In [3]: behaviour.head()

Out[3]:

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

In [4]: transactions.head(3)

Out[4]:

	DATE	STORE_NBR	LYLTY_CARD_NBR	TXN_ID	PROD_NBR	PROD_NAME	PROD_QTY	TOT_SALES
0	43390	1	1000	1	5	Natural Chip Compny SeaSalt175g	2	6.0
1	43599	1	1307	348	66	CCs Nacho Cheese 175g	3	6.3
2	43605	1	1343	383	61	Smiths Crinkle Cut Chips Chicken 170g	2	2.9

```
In [5]: # Look for Empty cells
        for i in transactions.columns:
            empty=np.mean(transactions[i].isnull())
            print( f'{i} -- {empty}% null values')
        DATE -- 0.0% null values
        STORE NBR -- 0.0% null values
        LYLTY CARD NBR -- 0.0% null values
        TXN ID -- 0.0% null values
        PROD NBR -- 0.0% null values
        PROD NAME -- 0.0% null values
        PROD QTY -- 0.0% null values
        TOT SALES -- 0.0% null values
In [6]: for i in behaviour.columns:
            empty=np.mean(behaviour[i].isnull())
            print( f'{i} -- {empty}% null values')
        LYLTY CARD NBR -- 0.0% null values
        LIFESTAGE -- 0.0% null values
        PREMIUM CUSTOMER -- 0.0% null values
```

Data has no nulls

In [7]: print(transactions.dtypes) print(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	-

```
In [8]: # Change Date column to Dates from numbers
    import datetime
    base_date = datetime.datetime(1900, 1, 1)

def date_change(provided_number):
        return base_date + datetime.timedelta(days=provided_number)

transactions['DATE']= transactions['DATE'].apply(date_change)

transactions.head()
```

Out[8]:

	DATE	STORE_NBR	LYLTY_CARD_NBR	TXN_ID	PROD_NBR	PROD_NAME	PROD_QTY	TOT_SALES
0	2018-10-19	1	1000	1	5	Natural Chip Compny SeaSalt175g	2	6.0
1	2019-05-16	1	1307	348	66	CCs Nacho Cheese 175g	3	6.3
2	2019-05-22	1	1343	383	61	Smiths Crinkle Cut Chips Chicken 170g	2	2.9
3	2018-08-19	2	2373	974	69	Smiths Chip Thinly S/Cream&Onion 175g	5	15.0
4	2018-08-20	2	2426	1038	108	Kettle Tortilla ChpsHny&Jlpno Chili 150g	3	13.8

```
In [9]: transactions['PROD_NAME']=transactions.PROD_NAME.str.strip()
```

In [10]: #Join behavior and transactions dataframes

df=pd.merge(transactions,behaviour)
df=pd.DataFrame(df)
df.head()

Out[10]:

	DATE	STORE_NBR	LYLTY_CARD_NBR	TXN_ID	PROD_NBR	PROD_NAME	PROD_QTY	TOT_SALES	LIFESTAGE	PREMIUM_CUST
0	2018- 10-19	1	1000	1	5	Natural Chip Compny SeaSalt175g	2	6.0	YOUNG SINGLES/COUPLES	Pr
1	2019- 05-16	1	1307	348	66	CCs Nacho Cheese 175g	3	6.3	MIDAGE SINGLES/COUPLES	E
2	2018- 11-12	1	1307	346	96	WW Original Stacked Chips 160g	2	3.8	MIDAGE SINGLES/COUPLES	E
3	2019- 03-11	1	1307	347	54	CCs Original 175g	1	2.1	MIDAGE SINGLES/COUPLES	E
4	2019- 05-22	1	1343	383	61	Smiths Crinkle Cut Chips Chicken 170g	2	2.9	MIDAGE SINGLES/COUPLES	E

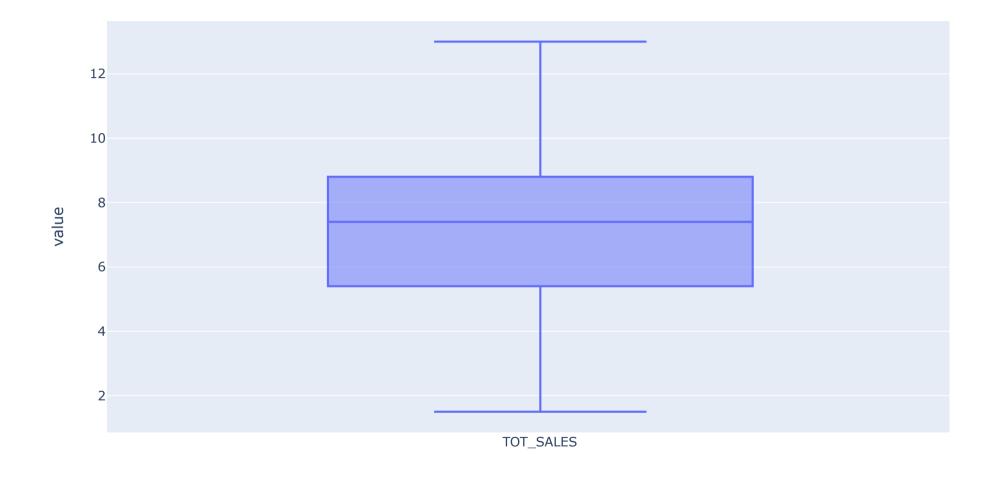
```
In [11]: # Check for outliers in total sales

IQR= df['PROD_QTY'].quantile(0.75)-df['PROD_QTY'].quantile(0.25)
lower_bound = df['PROD_QTY'].quantile(0.25)- 1.5*IQR
upper_bound = df['PROD_QTY'].quantile(0.75)+ 1.5*IQR

outliers_low= df[(df['PROD_QTY']<lower_bound)]
outliers_high= df[(df['PROD_QTY']>upper_bound)]

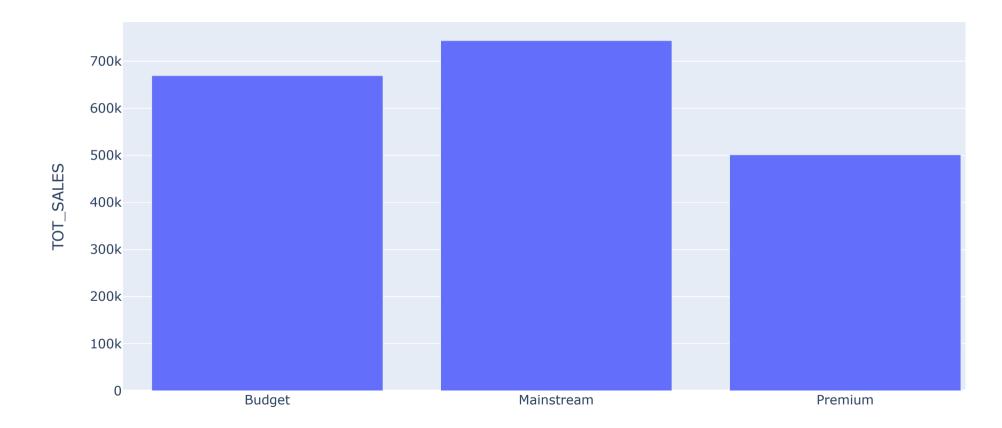
df_no_outliers= df[~(df['PROD_QTY']>upper_bound)]
```

In [12]: px.box(df_no_outliers['TOT_SALES'])

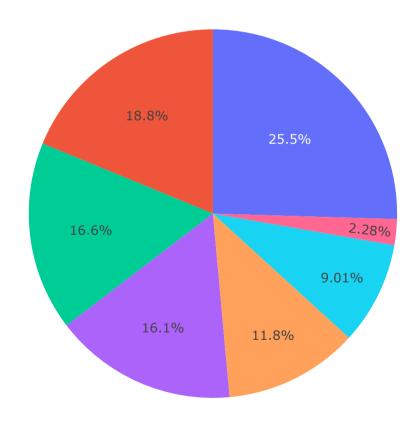


```
In [13]: grouped=df_no_outliers.groupby('PREMIUM_CUSTOMER')['TOT_SALES'].sum().reset_index()
px.bar(grouped,x='PREMIUM_CUSTOMER',y='TOT_SALES', title='Type of Customers vs ')
```

Type of Customers vs



Groups that make up Premium Customers

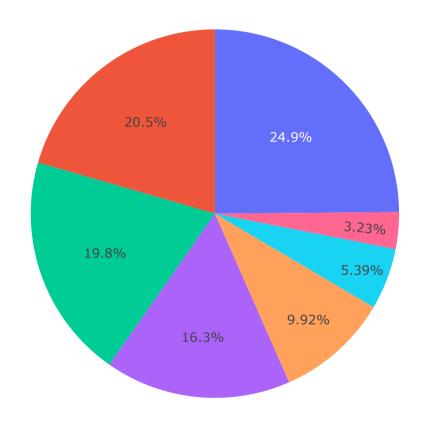


OLDER SINGL RETIREES YOUNG FAMIL OLDER FAMIL

MIDAGE SING

YOUNG SINGL
NEW FAMILIES

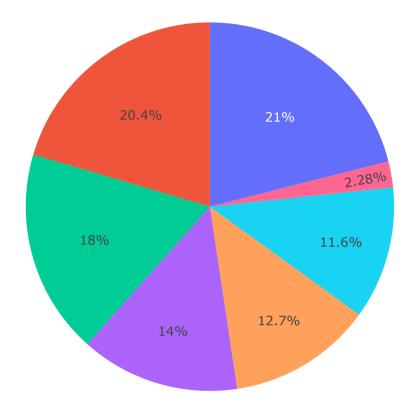
Groups that make up Budget Customers



OLDER FAMILI YOUNG FAMIL OLDER SINGL RETIREES YOUNG SINGL

MIDAGE SING NEW FAMILIE!

Groups that make up Mainstream Customers

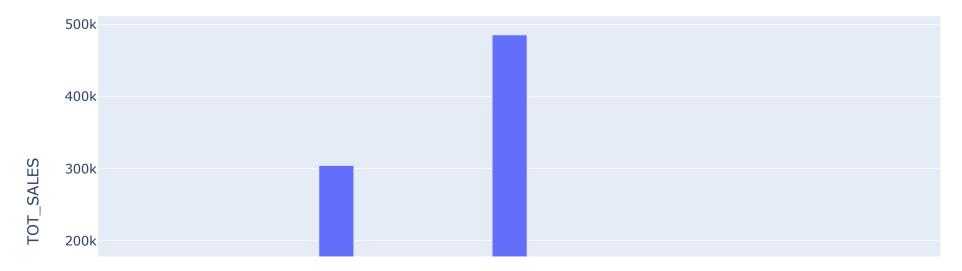


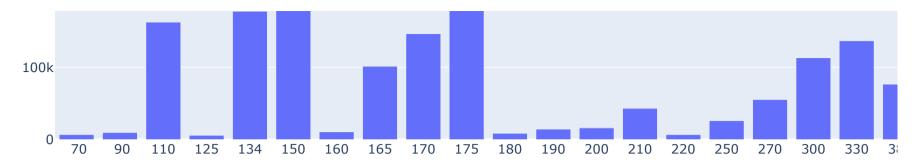
In [15]: df['weight_g']= df['PROD_NAME'].str[-4:-1]

RETIREES
YOUNG SINGL
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YOUNG FAMIL
MIDAGE SING
NEW FAMILIES

Sales for weight of chips

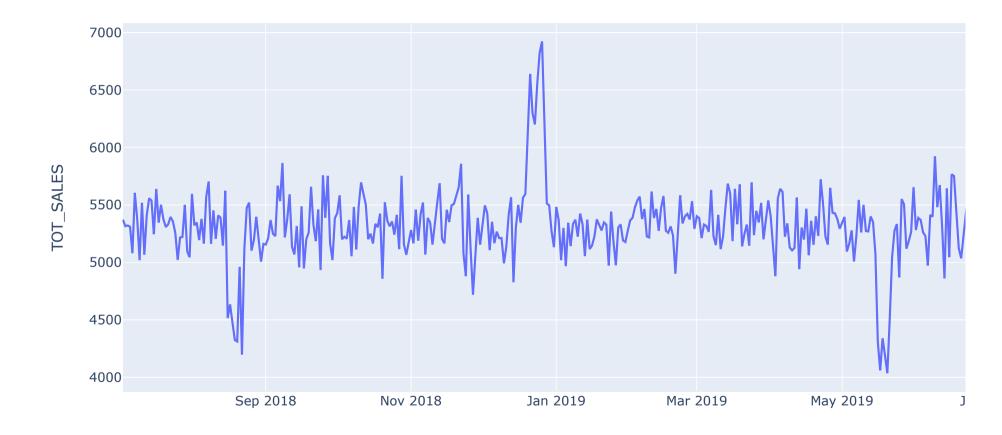




In [20]: date=df.groupby('DATE')['TOT_SALES'].sum().reset_index()

In [21]: px.line(date,x='DATE',y='TOT_SALES', title='Sales Over Time with All Sales Data')

Sales Over Time with All Sales Data

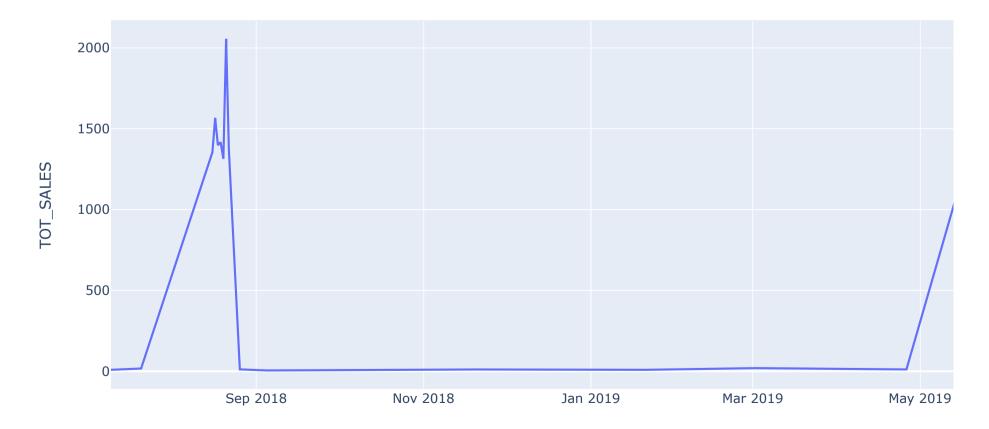


```
In [22]: date_no_outlier=df_no_outliers.groupby('DATE')['TOT_SALES'].sum().reset_index()
    px.line(date_no_outlier,x='DATE',y='TOT_SALES', title='Sales over time with outliers removed')
```

Sales over time with outliers removed



Spikes in sales over time



Spikes might be due to preparations for Labor Day (Sept 4) and Memorial Day (May 29)

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