## **Importing**

```
In [1]: import numpy as np
import pandas as pd

import calendar

from sklearn.preprocessing import StandardScaler
from sklearn.cluster import KMeans

import os
import seaborn as sns
import matplotlib.pyplot as plt
import plotly.express as px
import plotly.graph_objects as go
```

# Reading CSV files for all three dataset

```
In [2]: order = pd.read_csv("List_of_order.csv")
    order.head()
```

#### Out[2]:

|   | Order ID | Order Date | CustomerName | State          | City      |
|---|----------|------------|--------------|----------------|-----------|
| 0 | B-25601  | 1/4/2022   | Bharat       | Gujarat        | Ahmedabad |
| 1 | B-25602  | 1/4/2022   | Pearl        | Maharashtra    | Pune      |
| 2 | B-25603  | 3/4/2022   | Jahan        | Madhya Pradesh | Bhopal    |
| 3 | B-25604  | 3/4/2022   | Divsha       | Rajasthan      | Jaipur    |
| 4 | B-25605  | 5/4/2022   | Kasheen      | West Bengal    | Kolkata   |

```
In [3]: details = pd.read_csv("Order_Details.csv")
    details.head()
```

#### Out[3]:

|   | Order ID | Amount | Profit | Quantity | Category    | Sub-Category     |
|---|----------|--------|--------|----------|-------------|------------------|
| 0 | B-25601  | 1275   | -1148  | 7        | Furniture   | Bookcases        |
| 1 | B-25601  | 66     | -12    | 5        | Clothing    | Stole            |
| 2 | B-25601  | 8      | -2     | 3        | Clothing    | Hankerchief      |
| 3 | B-25601  | 80     | -56    | 4        | Electronics | Electronic Games |
| 4 | B-25602  | 168    | -111   | 2        | Electronics | Phones           |

```
In [4]: target = pd.read_csv("Sales_Target.csv")
    target.head()
```

#### Out[4]:

|   | Month of Order Date | Category  | Target |
|---|---------------------|-----------|--------|
| ( | <b>0</b> 18-Apr     | Furniture | 10400  |
|   | <b>1</b> 18-May     | Furniture | 10500  |
| : | <b>2</b> 18-Jun     | Furniture | 10600  |
| ; | <b>3</b> 18-Jul     | Furniture | 10800  |
|   | <b>4</b> 18-Aug     | Furniture | 10900  |

## Checking if there is any null values present in the dataset

```
In [6]: details.isnull().sum()
 Out[6]: Order ID
                         0
         Amount
                         0
         Profit
                         0
         Quantity
                         0
         Category
                          0
         Sub-Category
                         0
         dtype: int64
 In [7]: target.isnull().sum()
 Out[7]: Month of Order Date
                                0
         Category
                                0
         Target
                                0
         dtype: int64
 In [8]: target.shape
 Out[8]: (36, 3)
 In [9]: details.shape
 Out[9]: (1500, 6)
In [10]: order.shape
Out[10]: (500, 5)
```

#### Changing the columns type to appropriate data type

```
In [11]: order.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 500 entries, 0 to 499
         Data columns (total 5 columns):
            Column
                            Non-Null Count Dtype
                                            ____
             Order ID
                            500 non-null
                                            object
             Order Date
                            500 non-null
                                            object
              CustomerName 500 non-null
                                            object
          3
              State
                            500 non-null
                                            object
          4 City
                            500 non-null
                                            object
         dtvpes: object(5)
         memory usage: 19.7+ KB
In [12]: # Changing the Order Date variable to datetime data type
         order['Order Date'] = order['Order Date'].astype('datetime64[ns]')
         C:\Users\hp\anaconda3\lib\site-packages\pandas\core\tools\datetimes.py:1063: UserWarning: Parsing '13-04-2022' in D
         D/MM/YYYY format. Provide format or specify infer datetime format=True for consistent parsing.
           cache array = maybe cache(arg, format, cache, convert listlike)
         C:\Users\hp\anaconda3\lib\site-packages\pandas\core\tools\datetimes.py:1063: UserWarning: Parsing '15-04-2022' in D
         D/MM/YYYY format. Provide format or specify infer datetime format=True for consistent parsing.
           cache array = maybe cache(arg, format, cache, convert listlike)
         C:\Users\hp\anaconda3\lib\site-packages\pandas\core\tools\datetimes.py:1063: UserWarning: Parsing '17-04-2022' in D
         D/MM/YYYY format. Provide format or specify infer datetime format=True for consistent parsing.
           cache array = maybe cache(arg, format, cache, convert listlike)
         C:\Users\hp\anaconda3\lib\site-packages\pandas\core\tools\datetimes.py:1063: UserWarning: Parsing '18-04-2022' in D
         D/MM/YYYY format. Provide format or specify infer datetime format=True for consistent parsing.
           cache array = maybe cache(arg, format, cache, convert listlike)
         C:\Users\hp\anaconda3\lib\site-packages\pandas\core\tools\datetimes.py:1063: UserWarning: Parsing '20-04-2022' in D
         D/MM/YYYY format. Provide format or specify infer datetime format=True for consistent parsing.
           cache array = maybe cache(arg, format, cache, convert listlike)
         C:\Users\hp\anaconda3\lib\site-packages\pandas\core\tools\datetimes.py:1063: UserWarning: Parsing '22-04-2022' in D
         D/MM/YYYY format. Provide format or specify infer_datetime_format=True for consistent parsing.
           cache_array = _maybe_cache(arg, format, cache, convert_listlike)
         C:\Users\hp\anaconda3\lib\site-packages\pandas\core\tools\datetimes.py:1063: UserWarning: Parsing '23-04-2022' in D
```

```
In [13]: order.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 500 entries, 0 to 499
         Data columns (total 5 columns):
                            Non-Null Count Dtype
              Column
              _____
                                            ____
              Order ID
                            500 non-null
                                            object
                            500 non-null
                                            datetime64[ns]
              Order Date
             CustomerName 500 non-null
                                            object
              State
                            500 non-null
                                            object
             City
                            500 non-null
                                            object
          4
         dtypes: datetime64[ns](1), object(4)
         memory usage: 19.7+ KB
In [14]: # Cleaning the detail dataset
         details.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 1500 entries, 0 to 1499
         Data columns (total 6 columns):
                            Non-Null Count Dtype
              Column
              Order ID
                                            object
                            1500 non-null
          1
              Amount
                            1500 non-null
                                            int64
             Profit
                            1500 non-null
                                            int64
                            1500 non-null
                                            int64
              Ouantity
             Category
                            1500 non-null
                                            object
              Sub-Category 1500 non-null
                                            object
         dtypes: int64(3), object(3)
         memory usage: 70.4+ KB
```

```
In [15]: # Chaning the Category and Sub-category variable to categorical data type
        details['Category'] = details['Category'].astype('category')
        details['Sub-Category'] = details['Sub-Category'].astype('category')
        details.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 1500 entries, 0 to 1499
         Data columns (total 6 columns):
                          Non-Null Count Dtype
             Column
                           _____
             Order ID
                                          object
                          1500 non-null
         1 Amount2 Profit
                          1500 non-null
                                          int64
                          1500 non-null
                                         int64
             Quantity
                          1500 non-null
                                         int64
          3
          4 Category
                          1500 non-null category
          5 Sub-Category 1500 non-null category
         dtypes: category(2), int64(3), object(1)
        memory usage: 50.7+ KB
In [16]: target.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 36 entries, 0 to 35
        Data columns (total 3 columns):
                                 Non-Null Count Dtype
             Column
             Month of Order Date 36 non-null
                                                object
                                 36 non-null
            Category
                                                object
                        36 non-null
            Target
                                                int64
         dtypes: int64(1), object(2)
         memory usage: 992.0+ bytes
```

```
In [17]: | target['Category'] = target['Category'].astype('category')
         target.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 36 entries, 0 to 35
         Data columns (total 3 columns):
            Column
                                  Non-Null Count Dtype
             Month of Order Date 36 non-null
                                                  object
                                  36 non-null
          1 Category
                                                  category
                                  36 non-null
                                                  int64
            Target
         dtypes: category(1), int64(1), object(1)
         memory usage: 872.0+ bytes
In [18]: # Cleanded Details data
         details.head()
```

# Out[18]:

|   | Order ID | Amount | Profit | Quantity | Category    | Sub-Category     |
|---|----------|--------|--------|----------|-------------|------------------|
| 0 | B-25601  | 1275   | -1148  | 7        | Furniture   | Bookcases        |
| 1 | B-25601  | 66     | -12    | 5        | Clothing    | Stole            |
| 2 | B-25601  | 8      | -2     | 3        | Clothing    | Hankerchief      |
| 3 | B-25601  | 80     | -56    | 4        | Electronics | Electronic Games |
| 4 | B-25602  | 168    | -111   | 2        | Flectronics | Phones           |

In [19]: # Cleaned Order Data
 order.head()

Out[19]:

| City      | State          | CustomerName | Order Date | Order ID |   |
|-----------|----------------|--------------|------------|----------|---|
| Ahmedabad | Gujarat        | Bharat       | 2022-01-04 | B-25601  | 0 |
| Pune      | Maharashtra    | Pearl        | 2022-01-04 | B-25602  | 1 |
| Bhopal    | Madhya Pradesh | Jahan        | 2022-03-04 | B-25603  | 2 |
| Jaipur    | Rajasthan      | Divsha       | 2022-03-04 | B-25604  | 3 |
| Kolkata   | West Bengal    | Kasheen      | 2022-05-04 | B-25605  | 4 |

In [20]: # Cleaned Target Dataset
target.head()

#### Out[20]:

|   | Month of Order Date | Category  | Target |
|---|---------------------|-----------|--------|
| 0 | 18-Apr              | Furniture | 10400  |
| 1 | 18-May              | Furniture | 10500  |
| 2 | 18-Jun              | Furniture | 10600  |
| 3 | 18-Jul              | Furniture | 10800  |
| 4 | 18-Aug              | Furniture | 10900  |

Making a new dataframe containing the Amount, Profit and Quantity of the different orders. Then joining it with the Order datasets by taking Order ID as the Primary Key.

```
In [21]: profits = details.groupby('Order ID').sum().reset_index()
profits.head()
```

#### Out[21]:

|   | Order ID | Amount | Profit | Quantity |
|---|----------|--------|--------|----------|
| 0 | B-25601  | 1429   | -1218  | 19       |
| 1 | B-25602  | 3889   | 975    | 22       |
| 2 | B-25603  | 2025   | -180   | 25       |
| 3 | B-25604  | 222    | 22     | 11       |
| 4 | B-25605  | 75     | 0      | 7        |

Now I will merge two datasets order details and list of order As this two dataset have common relation between them using Order ID

```
In [22]: df = pd.merge(order, profits)
    df.head()
```

#### Out[22]:

|   | Order ID | Order Date | CustomerName | State          | City      | Amount | Profit | Quantity |
|---|----------|------------|--------------|----------------|-----------|--------|--------|----------|
| C | B-25601  | 2022-01-04 | Bharat       | Gujarat        | Ahmedabad | 1429   | -1218  | 19       |
| 1 | B-25602  | 2022-01-04 | Pearl        | Maharashtra    | Pune      | 3889   | 975    | 22       |
| 2 | B-25603  | 2022-03-04 | Jahan        | Madhya Pradesh | Bhopal    | 2025   | -180   | 25       |
| 3 | B-25604  | 2022-03-04 | Divsha       | Rajasthan      | Jaipur    | 222    | 22     | 11       |
| 4 | B-25605  | 2022-05-04 | Kasheen      | West Bengal    | Kolkata   | 75     | 0      | 7        |

## **Sales Trend Analysis**

Trend analysis is to find patterns in data, such as ups & downs. A "trend" is an upwards or downwards shift in a data set over time. In retail, this analysis of past trends in sales or revenue; allows to predict the future market. This analysis useful for budgeting and forecasting. Total sales of any business on a trend line may obtain some significant information

#### Out[23]:

|   | Order ID | Amount | Profit | Quantity |
|---|----------|--------|--------|----------|
| 0 | B-25601  | 1429   | -1218  | 19       |
| 1 | B-25602  | 3889   | 975    | 22       |
| 2 | B-25603  | 2025   | -180   | 25       |
| 3 | B-25604  | 222    | 22     | 11       |
| 4 | B-25605  | 75     | 0      | 7        |

the code extracts year and month information from the 'Order Date' column, adds corresponding columns to the DataFrame, groups the data by year and month, calculates the sum for each group, and sorts the resulting DataFrame.

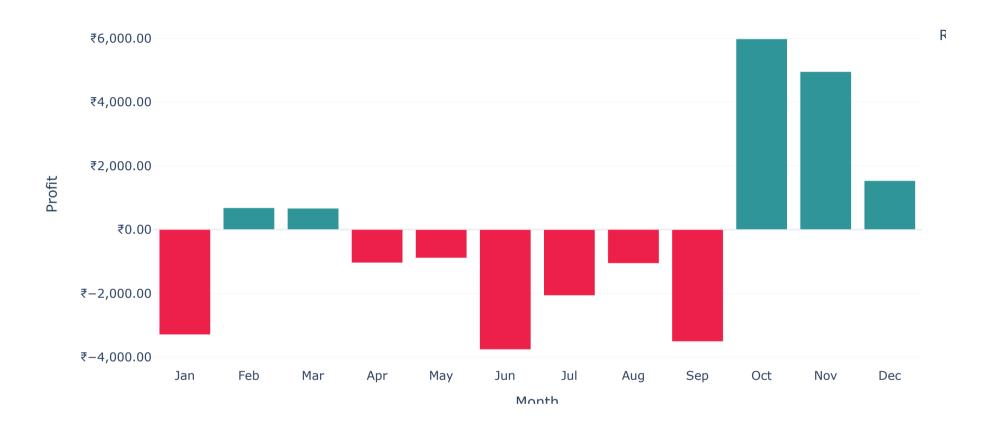
```
In [24]: df['Year'] = pd.DatetimeIndex(df['Order Date']).year
    df['Month_Number'] = pd.DatetimeIndex(df['Order Date']).month
    df['Month'] = df['Month_Number'].apply(lambda x: calendar.month_abbr[x])
    year_month = df.groupby(['Year', 'Month', 'Month_Number']).sum().sort_values(['Year', 'Month_Number'])
    year_month
```

## Out[24]:

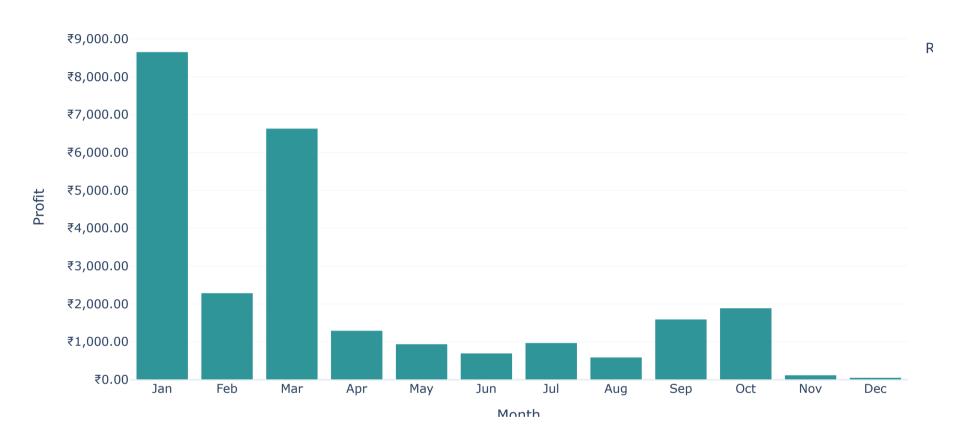
|      |       |              | Amount | Profit | Quantity |
|------|-------|--------------|--------|--------|----------|
| Year | Month | Month_Number |        |        |          |
| 2022 | Jan   | 1            | 18035  | -3296  | 203      |
|      | Feb   | 2            | 6566   | 685    | 58       |
|      | Mar   | 3            | 7434   | 669    | 144      |
|      | Apr   | 4            | 26170  | -1043  | 337      |
|      | May   | 5            | 20422  | -891   | 306      |
|      | Jun   | 6            | 17406  | -3759  | 353      |
|      | Jul   | 7            | 15682  | -2065  | 239      |
|      | Aug   | 8            | 45269  | -1059  | 601      |
|      | Sep   | 9            | 20210  | -3509  | 310      |
|      | Oct   | 10           | 32758  | 5979   | 414      |
|      | Nov   | 11           | 38858  | 4955   | 433      |
|      | Dec   | 12           | 23892  | 1535   | 209      |
| 2023 | Jan   | 1            | 50448  | 8655   | 640      |
|      | Feb   | 2            | 15894  | 2291   | 253      |
|      | Mar   | 3            | 39700  | 6633   | 485      |
|      | Apr   | 4            | 11079  | 1295   | 106      |
|      | May   | 5            | 4390   | 943    | 63       |
|      | Jun   | 6            | 3392   | 700    | 52       |
|      | Jul   | 7            | 5116   | 975    | 67       |
|      | Aug   | 8            | 6557   | 594    | 83       |
|      | Sep   | 9            | 5583   | 1597   | 70       |
|      | Oct   | 10           | 14147  | 1892   | 149      |
|      | Nov   | 11           | 2235   | 122    | 35       |
|      | Dec   | 12           | 259    | 57     | 5        |

```
In [25]: year month = year month.reset index()
         year_month["Color"] = np.where(year_month["Profit"]<0, 'Loss', 'Profit')</pre>
         year_month_2022 = year_month[year_month['Year']==2022]
         fig = px.bar(year month 2022, x='Month Number', y='Profit', color='Color',
                      title="Monthly Profit in 2022",
                      labels=dict(Month Number="Month", Profit="Profit", Color="Results"),
                      color discrete map={
                           'Loss': '#EC2049',
                           'Profit': '#2F9599'},
                      hover data=["Month", "Profit"],
                      template='plotly white')
         fig.update layout(yaxis tickprefix = '₹', yaxis tickformat = ',.2f')
         fig.update layout(
             xaxis = dict(
                 tickvals = [1, 2, 3, 4, 5, 6, 7,8,9, 10, 11, 12],
                 ticktext = ['Jan', 'Feb', 'Mar', 'Apr', 'May', 'Jun', 'Jul', 'Aug', 'Sep', 'Oct', 'Nov', 'Dec']
         fig.show()
```

# Monthly Profit in 2022



# Monthly Profit in 2023

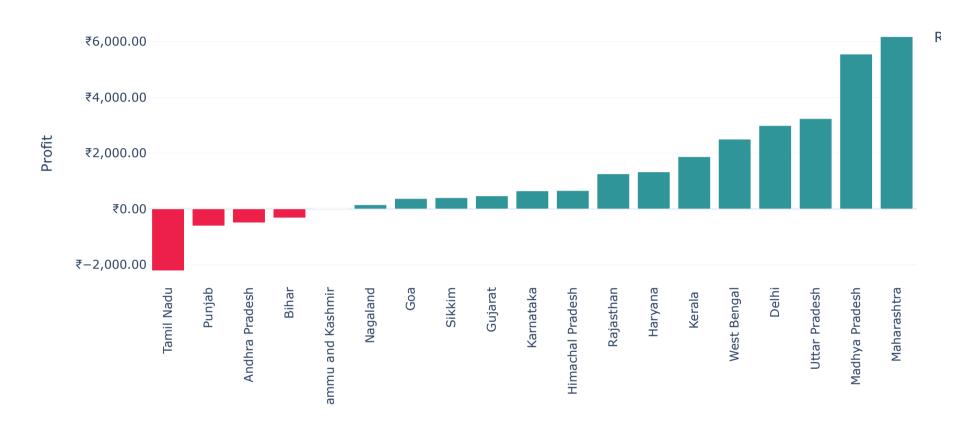


### Out[27]:

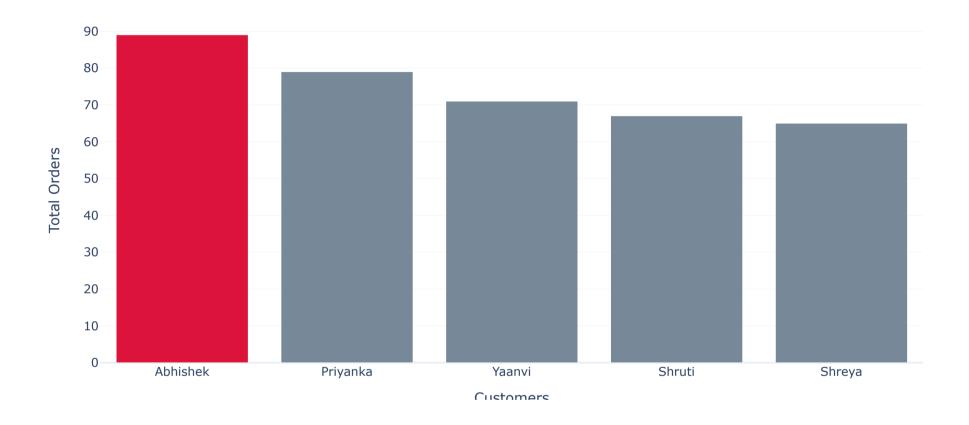
|    | State             | Total Orders |
|----|-------------------|--------------|
| 16 | Tamil Nadu        | 8            |
| 15 | Sikkim            | 12           |
| 3  | Goa               | 14           |
| 5  | Haryana           | 14           |
| 6  | Himachal Pradesh  | 14           |
| 7  | Jammu and Kashmir | 14           |
| 0  | Andhra Pradesh    | 15           |
| 12 | Nagaland          | 15           |
| 9  | Kerala            | 16           |
| 1  | Bihar             | 16           |
| 8  | Karnataka         | 21           |
| 17 | Uttar Pradesh     | 22           |
| 2  | Delhi             | 22           |
| 18 | West Bengal       | 22           |
| 13 | Punjab            | 25           |
| 4  | Gujarat           | 27           |
| 14 | Rajasthan         | 32           |
| 11 | Maharashtra       | 90           |
| 10 | Madhya Pradesh    | 101          |

```
In [28]: profit_by_state = df.groupby('State').sum().reset_index().sort_values(['Profit'])
         profit by state["Color"] = np.where(profit by state["Profit"]<0, 'Loss', 'Profit')</pre>
         fig = px.bar(profit by state, x='State', y='Profit',
                      color='Color', color discrete map={
                           'Loss': '#EC2049',
                           'Profit': '#2F9599'},
                      title="Profit by State",
                      labels=dict(Color="Results"),
                      template='plotly white')
         # Disabling Zoom
         fig.layout.xaxis.fixedrange = True
         fig.layout.yaxis.fixedrange = True
         fig.update layout(yaxis tickprefix = '₹', yaxis tickformat = ',.2f')
         fig.update xaxes(
                 tickangle = -90,
                 title text = "States",
         fig.show()
```

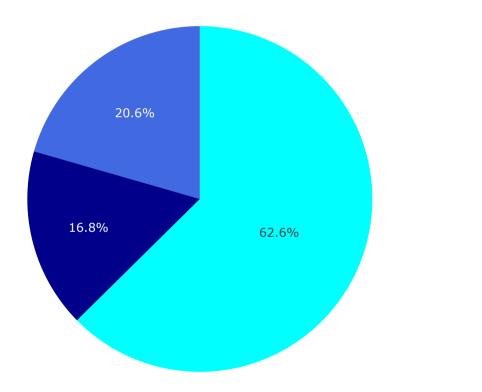
## Profit by State



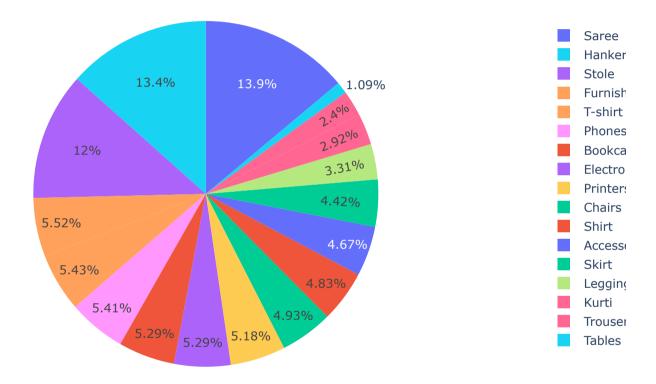
Top 5 Customers



## Total Quantity Sold per Category

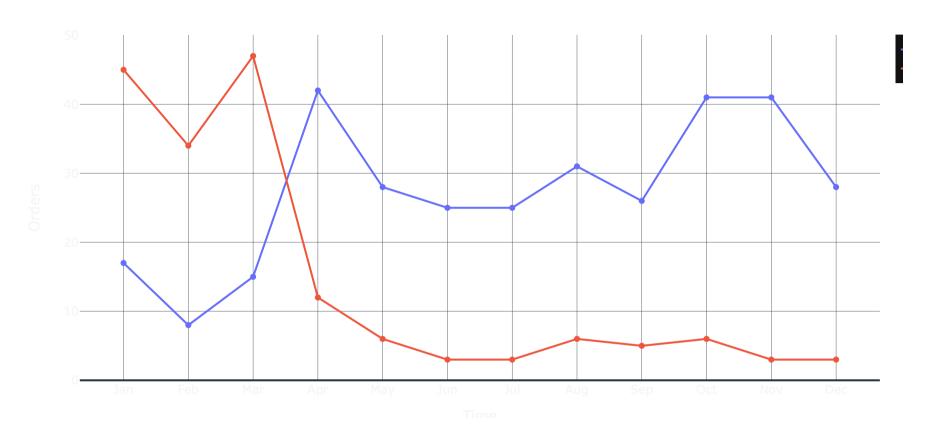


## Total Quantity Sold per Sub-Category

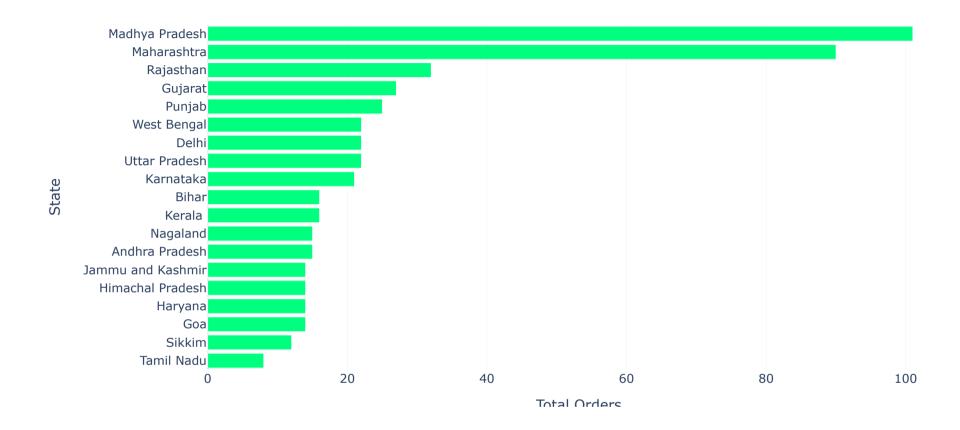


```
Analysing-E-commerce-data - Jupyter Notebook
In [32]:
         date_orders = order.groupby('Order Date').size().reset_index(name="Orders")
         date orders['Month'] = pd.DatetimeIndex(date orders['Order Date']).month
         date orders['Year'] = pd.DatetimeIndex(date orders['Order Date']).year
         date orders 2022 = date orders[date orders['Year']==2022]
         date orders 2023 = date orders[date orders['Year']==2023]
         month 2022 = date orders 2022.groupby('Month').sum().reset index()
         month 2023 = date orders 2023.groupby('Month').sum().reset index()
         fig = go.Figure()
         fig.add trace(go.Scatter(
             name='2022',
             x=month 2022['Month'],
             y=month 2022['Orders'],
             connectgaps=True # override default to connect the gaps
         ))
         fig.add trace(go.Scatter(
             name='2023',
             x=month 2023['Month'],
             y=month 2023['Orders'],
             connectgaps=True # override default to connect the gaps
         ))
         fig.update layout(title text='Monthly Quantity Sold',
                          template='plotly dark')
         fig.update xaxes(title text='Time')
         fig.update yaxes(title text='Orders')
         fig.update layout(
             xaxis = dict(
                 tickvals = [1, 2, 3, 4, 5, 6, 7,8,9, 10, 11, 12],
                 ticktext = ['Jan', 'Feb', 'Mar', 'Apr', 'May', 'Jun', 'Jul', 'Aug', 'Sep', 'Oct', 'Nov', 'Dec']
         fig.layout.xaxis.fixedrange = True
         fig.layout.yaxis.fixedrange = True
```

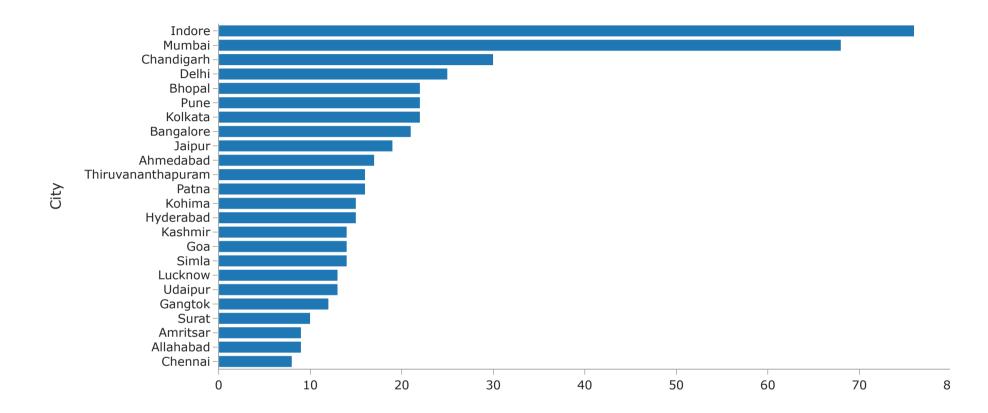
## Monthly Quantity Sold



## Total Orders by State

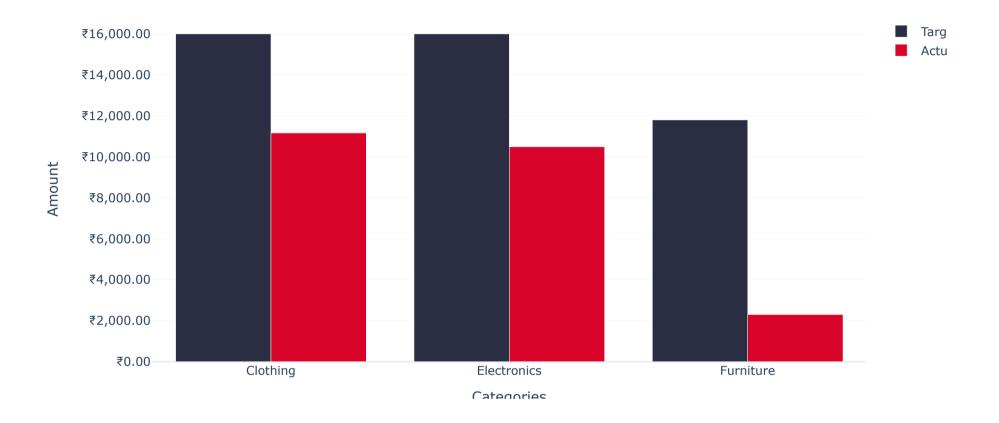


# Total Orders by City



```
In [35]: target category = target.groupby('Category').max().reset index()
         details category = details.groupby('Category').sum().reset index()
         target category['Actual Amount'] = details category['Profit']
         fig = go.Figure(data=[
             go.Bar(name='Target', x=target_category['Category'], y=target_category['Target'],
                   marker color='#2b2d42'),
             go.Bar(name='Actual Amount', x=target category['Category'], y=target category['Actual Amount'],
                   marker color='#d90429')
         ])
         fig.update_layout(title_text='Actual vs Target Sales',
                          template='plotly white')
         fig.update xaxes(title text='Categories')
         fig.update yaxes(title text='Amount')
         fig.update layout(yaxis tickprefix = '₹', yaxis tickformat = ',.2f')
         fig.layout.xaxis.fixedrange = True
         fig.layout.yaxis.fixedrange = True
         fig.show()
```

## Actual vs Target Sales



#### Out[36]:

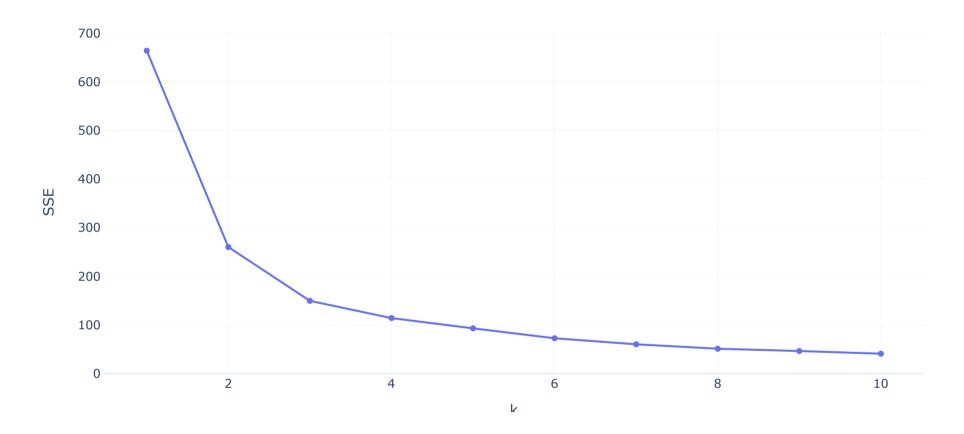
|   | CustomerName | Amount | Quantity |
|---|--------------|--------|----------|
| 0 | Aakanksha    | 74     | 8        |
| 1 | Aarushi      | 4701   | 49       |
| 2 | Aashna       | 1931   | 32       |
| 3 | Aastha       | 3276   | 28       |
| 4 | Aayush       | 556    | 18       |

```
In [37]: # Standardizing
         customer_seg2 = customer_seg[['Amount', 'Quantity']]
         scaler = StandardScaler()
         scaler.fit(customer seg2)
         customers normalized = scaler.transform(customer seg2)
         customers_normalized
         # Elbow Method to find best number of clusters
         sse = \{\}
         for k in range(1, 11):
             kmeans = KMeans(n clusters=k, random state=42)
             kmeans.fit(customers normalized)
             sse[k] = kmeans.inertia # SSE to closest cluster centroid
         # Plotting SSE
         fig = go.Figure()
         fig.add trace(go.Scatter(
             x=list(sse.keys()),
             y=list(sse.values()),
             connectgaps=True # override default to connect the gaps
         ))
         fig.update layout(title text='The Elbow Method',
                          template='plotly white')
         fig.update xaxes(title text='k')
         fig.update yaxes(title text='SSE')
         fig.layout.xaxis.fixedrange = True
         fig.layout.yaxis.fixedrange = True
         fig.show()
```

C:\Users\hp\anaconda3\lib\site-packages\sklearn\cluster\\_kmeans.py:1036: UserWarning:

KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP NUM THREADS=2.

## The Elbow Method



```
In [38]: # KMeans
model = KMeans(n_clusters=3)
model.fit(customers_normalized)
customer_seg['Cluster'] = model.labels_ + 1
customer_seg['Cluster'] = customer_seg['Cluster'].astype('category')
customer_seg.head()
```

#### Out[38]:

|   | CustomerName | Amount | Quantity | Cluster |
|---|--------------|--------|----------|---------|
| 0 | Aakanksha    | 74     | 8        | 1       |
| 1 | Aarushi      | 4701   | 49       | 2       |
| 2 | Aashna       | 1931   | 32       | 3       |
| 3 | Aastha       | 3276   | 28       | 3       |
| 4 | Aayush       | 556    | 18       | 1       |

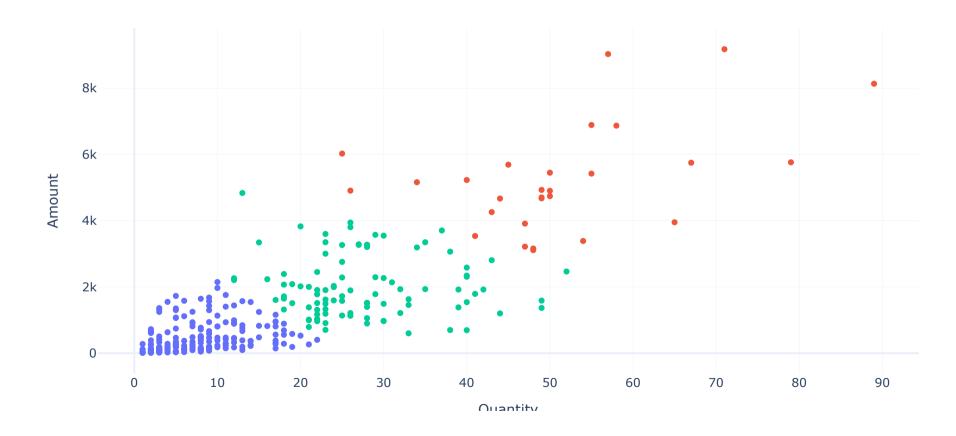
```
In [39]: customer_seg.groupby('Cluster').agg({
    'Amount':'mean',
    'Quantity':'count'}).round(2)
```

#### Out[39]:

#### **Amount Quantity**

# 1 451.36 208 2 5237.32 28 3 1989.32 96

## Amount vs Quantity - Customer Segmentation



```
In [41]: details.columns
Out[41]: Index(['Order ID', 'Amount', 'Profit', 'Quantity', 'Category', 'Sub-Category'], dtype='object')
```

```
In [42]: # constructing a heatmap to understand the correlation

plt.figure(figsize=(10,10))
sns.heatmap(details.corr(), cmap='Blues',annot = True)
```

## Out[42]: <AxesSubplot:>

