

Question1

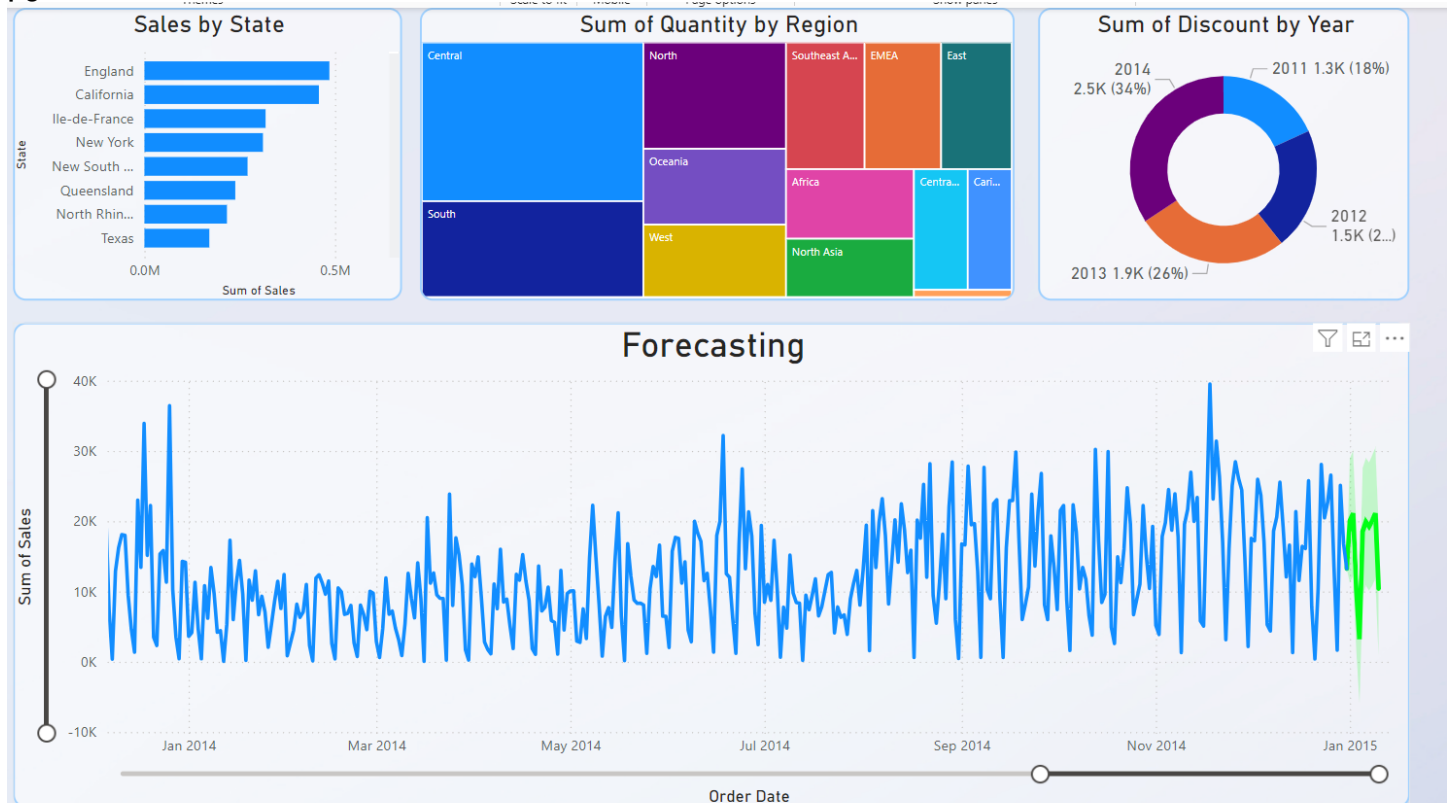
Use the provided sales data in Excel format and construct a dashboard in PowerBI

DASHBOARD

pg-1



pg-2



If any problem opening the files please use this link

https://drive.google.com/drive/folders/17VF3xbpNuMzjbQAKPIRXwJFBKaZV7R-P?usp=drive_link

Steps

- Preprocess the data : I removed the pin code column because it had too many missing numbers and I added an extra column called AOV(Average order value) to find the average order value of our customer
- Simple KPI cards to show revenue, profits, quantity sold and AOV
- Used a Slicer to organize the data on y-o-y basis and Segment basis
- Used line chart with date order to get monthly profits over the years
- Created a simple Forecasting module which can forecast the next possible price upto 15days with 95% confidence
- Used map to Legend = Region and bubble size for sales for country to get which country and region we have the most sales in we can use the map and look at it
- Used a treemap to check the quantity based on each region
- Sales Growth Rate: Show the sales growth rate on a quarterly or yearly basis to understand the business's growth trends.

PYTHON DOCUMENTATION

OBJECTIVE

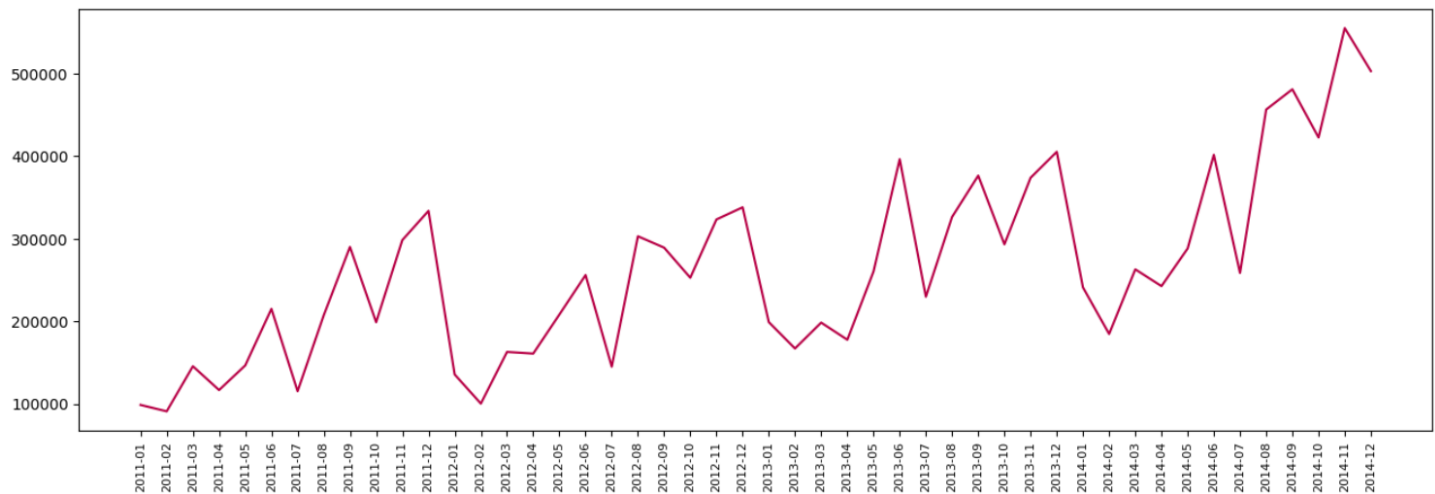
- Overall sales trend
- Sales in Countries
- Top 10 products by sales
- Total Sales by Segment
- Most Selling Products
- Most preferred Ship Mode
- Most Profitable Category and Sub-Category

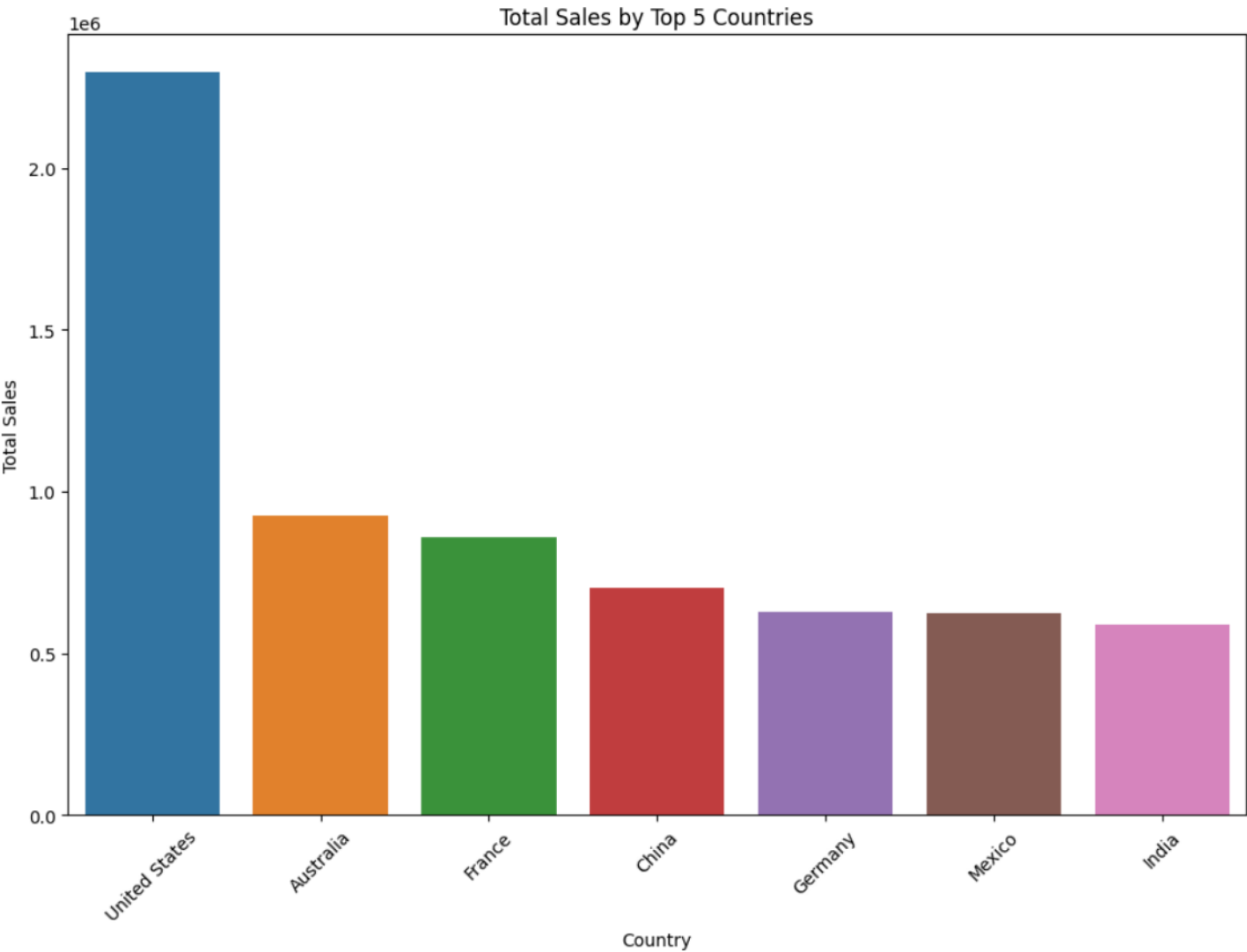
```
df.describe().round()
```

	Row ID	Sales	Quantity	Discount	Profit	Shipping Cost
count	51290.0	51290.0	51290.0	51290.0	51290.0	51290.0
mean	25646.0	246.0	3.0	0.0	29.0	26.0
std	14806.0	488.0	2.0	0.0	174.0	57.0
min	1.0	0.0	1.0	0.0	-6600.0	0.0
25%	12823.0	31.0	2.0	0.0	0.0	3.0
50%	25646.0	85.0	3.0	0.0	9.0	8.0
75%	38468.0	251.0	5.0	0.0	37.0	24.0
max	51290.0	22638.0	14.0	1.0	8400.0	934.0

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overall sales trend





dzvg503uh

August 2, 2024

NapQueen(ANARX) GLOBAL SUPERSTORE ANALYSIS

OBJECTIVE

Overall sales trend

Sales in Countries

Top 10 products by sales

Total Sales by Segment

Most Selling Products

Most preferred Ship Mode

Most Profitable Category and Sub-Category

IMPORTING REQUIRED LIBRARIES

```
[3]: # Data Manipulation
import pandas as pd

# Data Visualisation
import matplotlib.pyplot as plt
%matplotlib inline

import seaborn as sns
```

IMPORTING THE DATASET

```
[4]: # Importing dataset
df = pd.read_csv('Global-Superstore - Global-Superstore.csv.csv')
```

DATA AUDIT

You can't make your data work for you until you know what data you're talking about.

To get a quick idea of what the data looks like, we can call the head function on the data frame. By default, this returns the top five rows, but it can take in a parameter of how many rows to return.

```
[5]: # First five rows of the dataset
df.head()
```

```
[5]:
```

	Row ID	Order ID	Order Date	Ship Date	Ship Mode	Customer ID \
0	32298	CA-2012-124891	7/31/2012	7/31/2012	Same Day	RH-19495
1	26341	IN-2013-77878	2/5/2013	2/7/2013	Second Class	JR-16210
2	25330	IN-2013-71249	10/17/2013	10/18/2013	First Class	CR-12730
3	13524	ES-2013-1579342	1/28/2013	1/30/2013	First Class	KM-16375
4	47221	SG-2013-4320	11/5/2013	11/6/2013	Same Day	RH-9495

	Customer Name	Segment	City	State ... \
0	Rick Hansen	Consumer	New York City	New York ...
1	Justin Ritter	Corporate	Wollongong	New South Wales ...
2	Craig Reiter	Consumer	Brisbane	Queensland ...
3	Katherine Murray	Home Office	Berlin	Berlin ...
4	Rick Hansen	Consumer	Dakar	Dakar ...

	Product ID	Category	Sub-Category \
0	TEC-AC-10003033	Technology	Accessories
1	FUR-CH-10003950	Furniture	Chairs
2	TEC-PH-10004664	Technology	Phones
3	TEC-PH-10004583	Technology	Phones
4	TEC-SHA-10000501	Technology	Copiers

	Product Name	Sales Quantity \
0	Plantronics CS510 - Over-the-Head monaural Wir...	2309.650 7
1	Novimex Executive Leather Armchair, Black	3709.395 9
2	Nokia Smart Phone, with Caller ID	5175.171 9
3	Motorola Smart Phone, Cordless	2892.510 5
4	Sharp Wireless Fax, High-Speed	2832.960 8

	Discount	Profit	Shipping Cost	Order Priority
0	0.0	762.1845	933.57	Critical
1	0.1	-288.7650	923.63	Critical
2	0.1	919.9710	915.49	Medium
3	0.1	-96.5400	910.16	Medium
4	0.0	311.5200	903.04	Critical

[5 rows x 24 columns]

```
[6]: # Last five rows of the dataset
df.tail()
```

```
[6]:
```

	Row ID	Order ID	Order Date	Ship Date	Ship Mode \
51285	29002	IN-2014-62366	6/19/2014	6/19/2014	Same Day
51286	35398	US-2014-102288	6/20/2014	6/24/2014	Standard Class
51287	40470	US-2013-155768	12/2/2013	12/2/2013	Same Day
51288	9596	MX-2012-140767	2/18/2012	2/22/2012	Standard Class
51289	6147	MX-2012-134460	5/22/2012	5/26/2012	Second Class

	Customer ID	Customer Name	Segment	City	State	...	\
51285	KE-16420	Katrina Edelman	Corporate	Kure	Hiroshima	...	
51286	ZC-21910	Zuschuss Carroll	Consumer	Houston	Texas	...	
51287	LB-16795	Laurel Beltran	Home Office	Oxnard	California	...	
51288	RB-19795	Ross Baird	Home Office	Valinhos	S<o Paulo	...	
51289	MC-18100	Mick Crebagga	Consumer	Tipitapa	Managua	...	

	Product ID	Category	Sub-Category	\
51285	OFF-FA-10000746	Office Supplies	Fasteners	
51286	OFF-AP-10002906	Office Supplies	Appliances	
51287	OFF-EN-10001219	Office Supplies	Envelopes	
51288	OFF-BI-10000806	Office Supplies	Binders	
51289	OFF-PA-10004155	Office Supplies	Paper	

	Product Name	Sales	Quantity	\
51285	Advantus Thumb Tacks, 12 Pack	65.100	5	
51286	Hoover Replacement Belt for Commercial Guardsm...	0.444	1	
51287	#10- 4 1/8" x 9 1/2" Security-Tint Envelopes	22.920	3	
51288	Acco Index Tab, Economy	13.440	2	
51289	Eaton Computer Printout Paper, 8.5 x 11	61.380	3	

	Discount	Profit	Shipping Cost	Order Priority
51285	0.0	4.5000	0.01	Medium
51286	0.8	-1.1100	0.01	Medium
51287	0.0	11.2308	0.01	High
51288	0.0	2.4000	0.00	Medium
51289	0.0	1.8000	0.00	High

[5 rows x 24 columns]

```
[7]: # Shape of the dataset
df.shape
```

```
[7]: (51290, 24)
```

```
[8]: # Columns present in the dataset
df.columns
```

```
[8]: Index(['Row ID', 'Order ID', 'Order Date', 'Ship Date', 'Ship Mode',
        'Customer ID', 'Customer Name', 'Segment', 'City', 'State', 'Country',
        'Postal Code', 'Market', 'Region', 'Product ID', 'Category',
        'Sub-Category', 'Product Name', 'Sales', 'Quantity', 'Discount',
        'Profit', 'Shipping Cost', 'Order Priority'],
        dtype='object')
```

```
[9]: # A concise summary of the dataset
df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 51290 entries, 0 to 51289
Data columns (total 24 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Row ID                 51290 non-null  int64
1   Order ID               51290 non-null  object
2   Order Date             51290 non-null  object
3   Ship Date              51290 non-null  object
4   Ship Mode               51290 non-null  object
5   Customer ID            51290 non-null  object
6   Customer Name          51290 non-null  object
7   Segment                51290 non-null  object
8   City                   51290 non-null  object
9   State                  51290 non-null  object
10  Country                 51290 non-null  object
11  Postal Code             9994 non-null   float64
12  Market                  51290 non-null  object
13  Region                  51290 non-null  object
14  Product ID              51290 non-null  object
15  Category                51290 non-null  object
16  Sub-Category            51290 non-null  object
17  Product Name            51290 non-null  object
18  Sales                   51290 non-null  float64
19  Quantity                51290 non-null  int64
20  Discount                51290 non-null  float64
21  Profit                  51290 non-null  float64
22  Shipping Cost           51290 non-null  float64
23  Order Priority           51290 non-null  object
dtypes: float64(5), int64(2), object(17)
memory usage: 9.4+ MB

```

Now we can do further analysis on our data to answer our questions. Before that, we should see if there are any missing values in our data set. To check if there are any missing values in the entire data set we use the `isnull` function, then see if there are any values.

```

[10]: # Checking missing values
      df.isna().sum()

```

```

[10]: Row ID                0
      Order ID              0
      Order Date            0
      Ship Date              0
      Ship Mode              0
      Customer ID           0
      Customer Name         0
      Segment               0
      City                  0

```



```

State          0
Country        0
Postal Code    41296
Market         0
Region         0
Product ID     0
Category       0
Sub-Category   0
Product Name   0
Sales          0
Quantity       0
Discount       0
Profit         0
Shipping Cost  0
Order Priority  0
dtype: int64

```

Postal code has many missing values since we have city address we dont require postal code and we cant fill postal codes with other vales so drop postal code column

```

[12]: # Drop the 'Postal Code' column
df = df.drop(columns=['Postal Code'])

```

Next, we can look at some descriptive statistics of the data frame with the describe method.

This shows some descriptive statistics on the data set. Notice, it only shows the statistics on the numerical columns. From here you can see the following statistics:

- Row count, which aligns to what the shape attribute showed us.
- The mean, or average.
- The standard deviation, or how spread out the data is.
- The minimum and maximum value of each column
- The number of items that fall within the first, second, and third percentiles.

```

[13]: # Generating descriptive statistics summary
df.describe().round()

```

```

[13]:      Row ID    Sales  Quantity  Discount  Profit  Shipping Cost
count  51290.0  51290.0   51290.0   51290.0  51290.0         51290.0
mean    25646.0    246.0         3.0         0.0     29.0           26.0
std     14806.0    488.0         2.0         0.0    174.0           57.0
min         1.0      0.0         1.0         0.0   -6600.0           0.0
25%     12823.0     31.0         2.0         0.0      0.0            3.0
50%     25646.0     85.0         3.0         0.0      9.0            8.0
75%     38468.0    251.0         5.0         0.0     37.0           24.0
max     51290.0  22638.0        14.0         1.0   8400.0          934.0

```

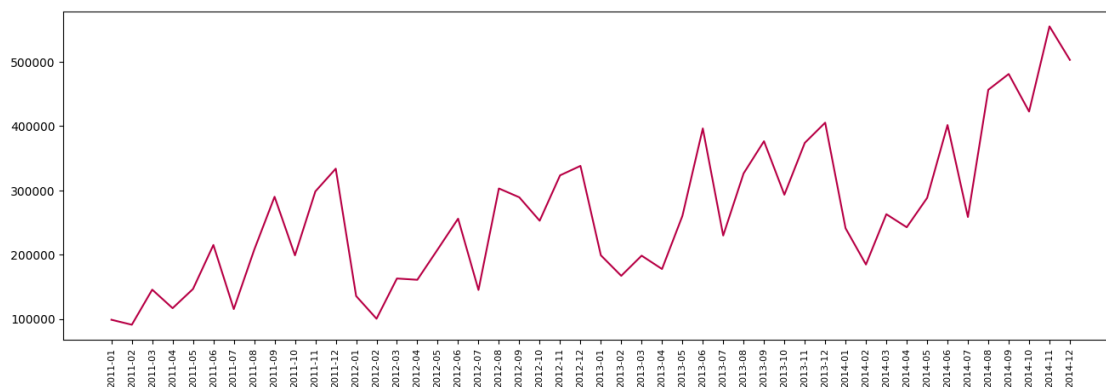
EXPLORATORY DATA ANALYSIS

- WHAT IS THE OVERALL SALES TREND?

```
[16]: # Getting month year from order_date
df['Order Date'] = pd.to_datetime(df['Order Date'])
df['month_year'] = df['Order Date'].apply(lambda x: x.strftime('%Y-%m'))
```

```
[20]: # Step 3: Group by 'month_year' and sum 'Sales'
df_temp = df.groupby('month_year')['Sales'].sum().reset_index()
```

```
[22]: # Setting the figure size
plt.figure(figsize=(16, 5))
plt.plot(df_temp['month_year'], df_temp['Sales'], color='#b80045')
plt.xticks(rotation='vertical', size=8)
plt.show()
```

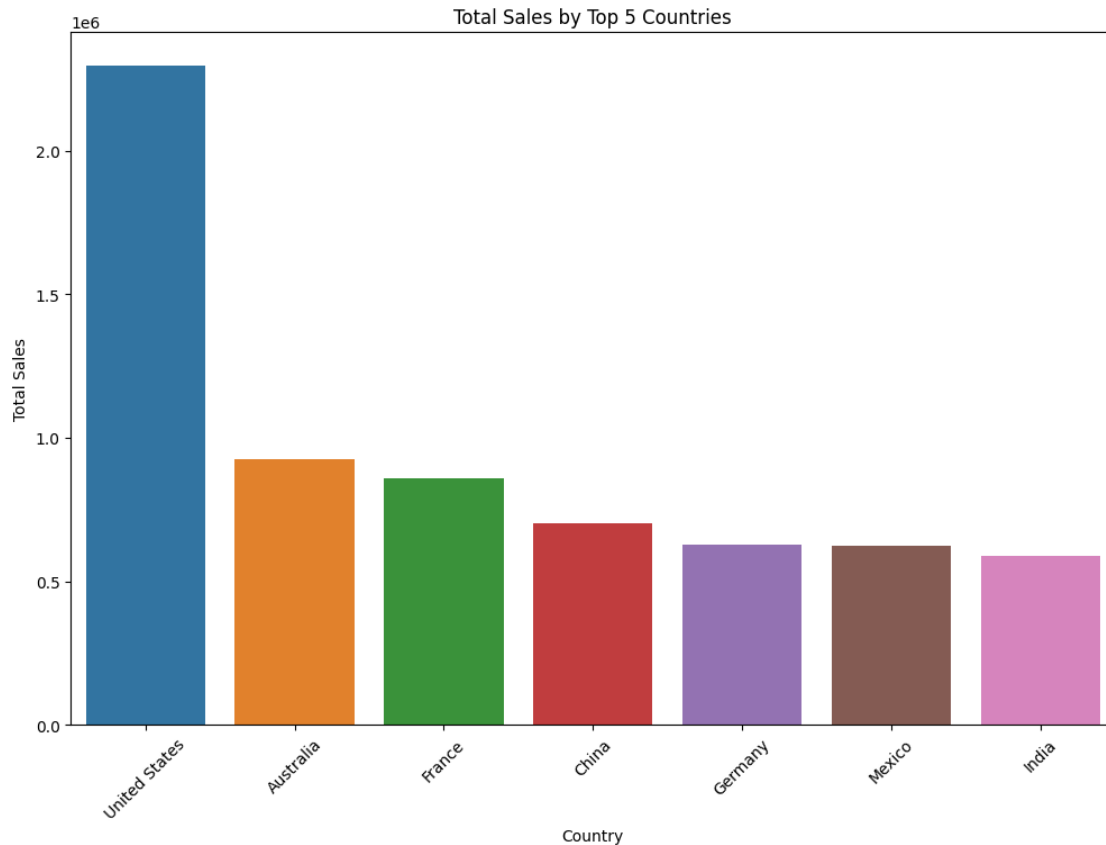


- What are the sales in Countries?

```
[37]: # Group by 'Country' and sum 'Sales'
country_sales = df.groupby('Country')['Sales'].sum().reset_index()

# Sort the dataframe in descending order and select the top 5 countries
top_countries = country_sales.sort_values(by='Sales', ascending=False).head(7)

# Plotting the data
plt.figure(figsize=(12, 8))
sns.barplot(x='Country', y='Sales', data=top_countries, estimator=sum)
plt.title('Total Sales by Top 5 Countries')
plt.xlabel('Country')
plt.ylabel('Total Sales')
plt.xticks(rotation=45)
plt.show()
```



- WHICH ARE THE TOP 10 PRODUCTS BY SALES?

```
[24]: # Grouping products by sales
prod_sales = df.groupby('Product Name')['Sales'].sum().reset_index()

# Sorting the dataframe in descending order
prod_sales.sort_values(by=['Sales'], inplace=True, ascending=False)

# Top 10 products by sales
prod_sales[:10]
```

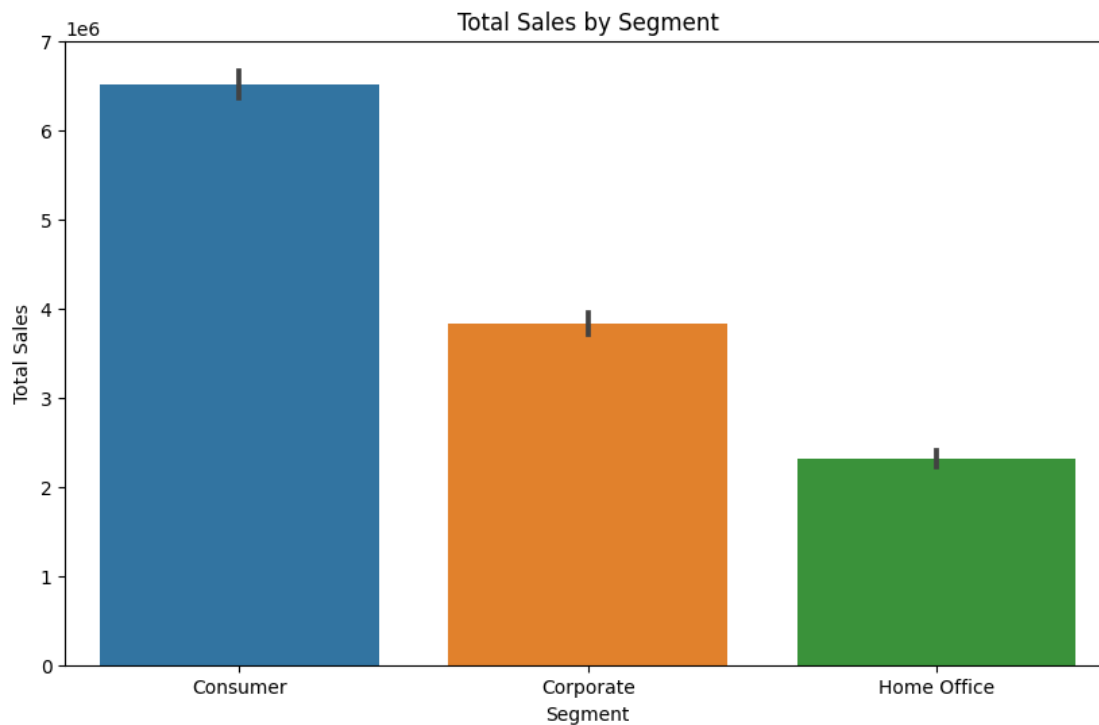
```
[24]:
```

	Product Name	Sales
310	Apple Smart Phone, Full Size	86935.7786
970	Cisco Smart Phone, Full Size	76441.5306
2415	Motorola Smart Phone, Full Size	73156.3030
2501	Nokia Smart Phone, Full Size	71904.5555
866	Canon imageCLASS 2200 Advanced Copier	61599.8240
1837	Hon Executive Leather Armchair, Adjustable	58193.4841
2631	Office Star Executive Leather Armchair, Adjust...	50661.6840
1714	Harbour Creations Executive Leather Armchair, ...	50121.5160
2988	Samsung Smart Phone, Cordless	48653.4600

-

Total Sales by Segment

```
[38]: plt.figure(figsize=(10, 6))
sns.barplot(x='Segment', y='Sales', data=df, estimator=sum)
plt.title('Total Sales by Segment')
plt.xlabel('Segment')
plt.ylabel('Total Sales')
plt.show()
```



- WHICH ARE THE MOST SELLING PRODUCTS?

```
[26]: # Grouping products by Quantity
best_selling_prods = pd.DataFrame(df.groupby('Product Name')['Quantity'].sum().
    ↪reset_index())

# Sorting the dataframe in descending order
best_selling_prods.sort_values(by=['Quantity'], inplace=True, ascending=False)

# Most selling products
best_selling_prods[:10]
```

```
[26]:
```

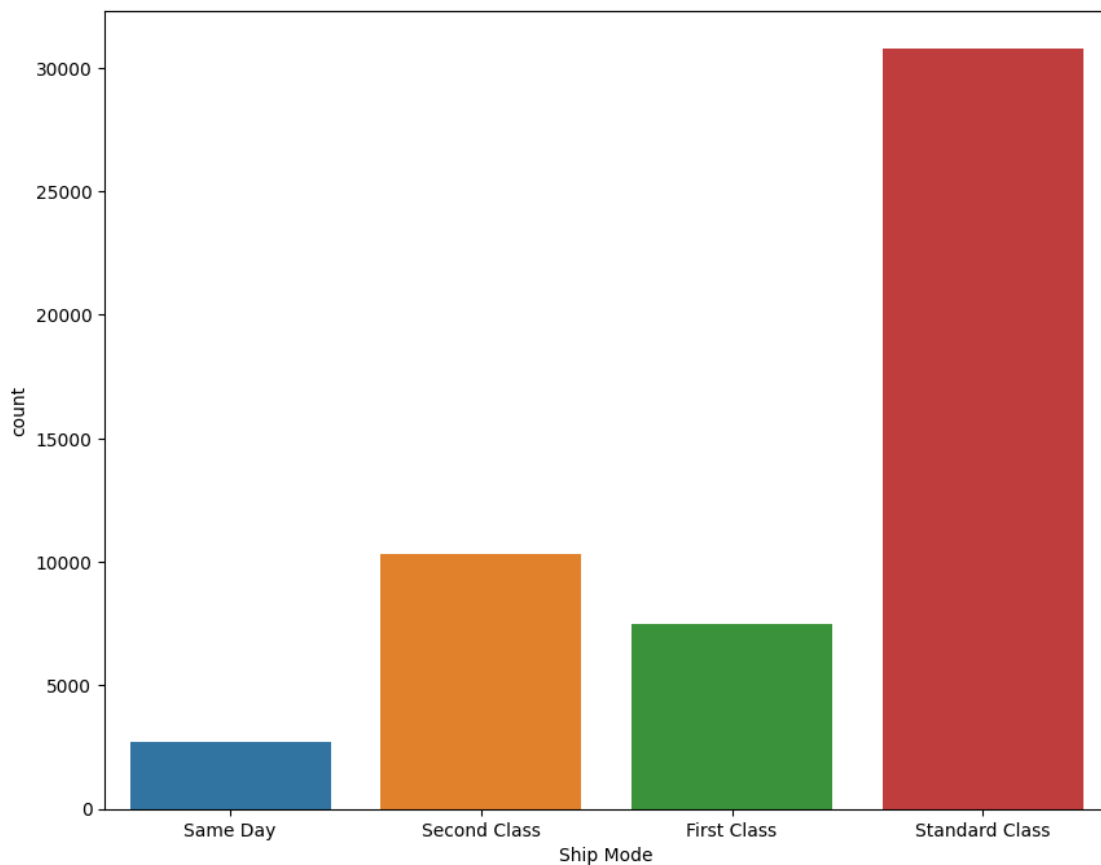
	Product Name	Quantity
3275	Staples	876
894	Cardinal Index Tab, Clear	337
1210	Eldon File Cart, Single Width	321
2840	Rogers File Cart, Single Width	262
3070	Sanford Pencil Sharpener, Water Color	259
3335	Stockwell Paper Clips, Assorted Sizes	253
446	Avery Index Tab, Clear	252
1981	Ibico Index Tab, Clear	251
3179	Smead File Cart, Single Width	250
3266	Stanley Pencil Sharpener, Water Color	242

- WHAT IS THE MOST PREFERRED SHIP MODE?

```
[27]: # Setting the figure size
plt.figure(figsize=(10, 8))

# countplot: Show the counts of observations in each categorical bin using bars
sns.countplot(x='Ship Mode', data=df)

# Display the figure
plt.show()
```



- WHICH ARE THE MOST PROFITABLE CATEGORY AND SUB-CATEGORY?

```
[31]: # Grouping products by Category and Sub-Category
cat_subcat = pd.DataFrame(df.groupby(['Category', 'Sub-Category'])['Profit'].
    ↪sum())

# Sorting the values
cat_subcat.sort_values(['Category', 'Profit'], ascending=False)
```

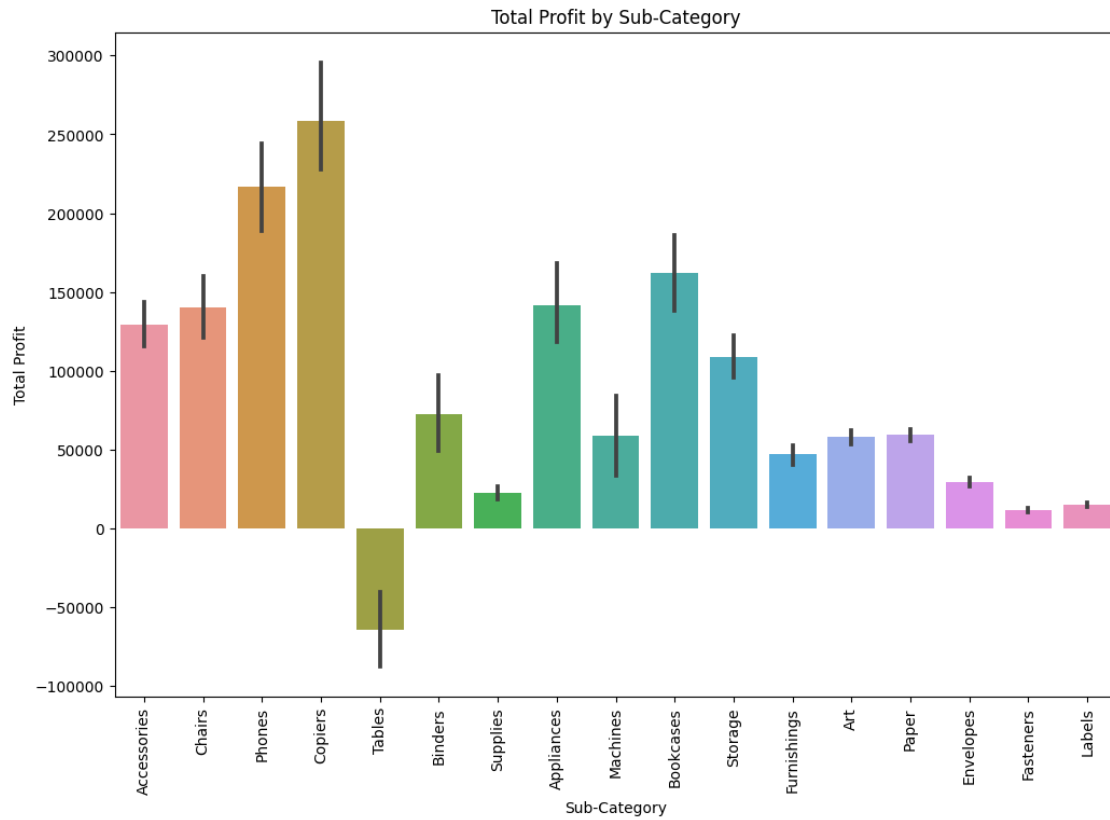
```
[31]:
```

	Category	Sub-Category	Profit
	Technology	Copiers	258567.54818
		Phones	216717.00580
		Accessories	129626.30620
		Machines	58867.87300
	Office Supplies	Appliances	141680.58940
		Storage	108461.48980
		Binders	72449.84600
		Paper	59207.68270
		Art	57953.91090
		Envelopes	29601.11630
		Supplies	22583.26310
		Labels	15010.51200
		Fasteners	11525.42410
	Furniture	Bookcases	161924.41950
		Chairs	140396.26750
		Furnishings	46967.42550
		Tables	-64083.38870

-

Graphical values

```
[39]: # Visualization 4: Profit by Sub-Category
plt.figure(figsize=(12, 8))
sns.barplot(x='Sub-Category', y='Profit', data=df, estimator=sum)
plt.title('Total Profit by Sub-Category')
plt.xlabel('Sub-Category')
plt.ylabel('Total Profit')
plt.xticks(rotation=90)
plt.show()
```



here we can see that tablets are loss making for us so its better for the company to stop selling tablets as we are losing money in those