US SAMPLE STORE RETAIL ANALYSIS

Melbourne Housing:max_bytes(150000):strip_icc():format(webp)/Walmart_exterior-8db53b3ec5c442f0a343fe01e6640090.jpg)

Step1: Importing Libaries

```
In [1]: #Library
    import numpy as np
    import pandas as pd
    import seaborn as sns
    import datetime
    import time
    import plotly.graph_objs as go
    import plotly.offline as py
    from plotly.offline import download_plotlyjs, init_notebook_mode, iplot

# py.init_notebook_mode(connected = True)
    pd.set_option('display.float_format', lambda x: f'{x:.1f}')
    df = pd.read_csv("us_superstore_sales.csv",encoding='latin-1')
```

First 5 Rows of the data

```
In [2]: df.head()
```

4

Out[2]:	Ro	w_ID	Order_ID	Order_Date	Ship_Date	Ship_Mode	Customer_ID	Customer_Name	Segment	Country	City	•••	Postal_Code	Region	Product_ID	Category	Sub_Catego
	0	1	CA- 2016- 152156	08/11/2016	11/11/2016	Second Class	CG-12520	Claire Gute	Consumer	United States	Henderson		42420	South	FUR-BO- 10001798	Furniture	Bookcas
	1	2	CA- 2016- 152156	08/11/2016	11/11/2016	Second Class	CG-12520	Claire Gute	Consumer	United States	Henderson		42420	South	FUR-CH- 10000454	Furniture	Cha
	2	3	CA- 2016- 138688	12/06/2016	16/06/2016	Second Class	DV-13045	Darrin Van Huff	Corporate	United States	Los Angeles		90036	West	OFF-LA- 10000240	Office Supplies	Labı
	3	4	US-2015- 108966	11/10/2015	18/10/2015	Standard Class	SO-20335	Sean O'Donnell	Consumer	United States	Fort Lauderdale		33311	South	FUR-TA- 10000577	Furniture	Tabl
	4	5	US-2015- 108966	11/10/2015	18/10/2015	Standard Class	SO-20335	Sean O'Donnell	Consumer	United States	Fort Lauderdale		33311	South	OFF-ST- 10000760	Office Supplies	Stora
	5 row:	s × 21	columns														
4																	>

Data Cleaning

```
In [4]: #Categorical Data
        df['Region'] = df['Region'].astype('category')
        df['Category'] = df['Category'].astype('category')
        df['Sub_Category'] = df['Sub_Category'].astype('category')
        df['Product_Name'] = df['Product_Name'].astype('category')
        df['Segment'] = df['Segment'].astype('category')
        df['Ship_Mode'] = df['Ship_Mode'].astype('category')
        df['Country'] = df['Country'].astype('category')
        df['State'] = df['State'].astype('category')
        df['City'] = df['City'].astype('category')
        #Integer Data
        df['Quantity'] = df['Quantity'].astype('int64')
        df['Discount'] = df['Discount'].fillna(0)
        df['Discount'] = df['Discount'].astype('float64')
        df['Sales'] = df['Sales'].fillna(0)
        df['Sales'] = df['Sales'].astype('float64')
        df['Profit'] = df['Profit'].fillna(0)
        df['Profit'] = df['Profit'].astype('float64')
```

```
#Date Data
df['Order_Date'] = pd.to_datetime(df['Order_Date'],format='%d/%m/%Y')
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9994 entries, 0 to 9993
Data columns (total 21 columns):
 # Column
             Non-Null Count Dtype
                 -----
 0 Row_ID 9994 non-null int64
1 Order_ID 9994 non-null object
 2 Order Date 9994 non-null datetime64[ns]
 3 Ship Date
                   9994 non-null object
 4 Ship Mode
                   9994 non-null category
 5 Customer_ID 9994 non-null object
 6 Customer_Name 9994 non-null object
    Segment 9994 non-null category
Country 9994 non-null category
City 9994 non-null category
State 9994 non-null category
 9
 10 State
 11 Postal_Code 9994 non-null int64
 12 Region 9994 non-null category
13 Product_ID 9994 non-null object
14 Category 9994 non-null category
 15 Sub_Category 9994 non-null category
 16 Product_Name 9994 non-null category
                   9994 non-null float64
 17 Sales
 18 Quantity 9994 non-null int64
 19 Discount 9994 non-null float64
 20 Profit 9994 non-null float64
dtypes: category(9), datetime64[ns](1), float64(3), int64(3), object(5)
memory usage: 1.1+ MB
```

Solving Business Questions using EDA

- 1. How much did by year the store make and sell between 2011 and 2014?
- 2. How much did by month the store make and sell between 2011 and 2014?
- 3. Which was the most profitable category, and which sold the most?
- 4. Which was the most profitable sub-category, and which sold the most?
- 5. Which was the most profitable segment, and which sold the most?
- 6. Which country bought the most, and which made the most profit?
- 7. Which country bought the less, and which made the less profit? There was negative profit?

```
In [5]: # Sales and profit by year

df['Order_Date'] = pd.to_datetime(df['Order_Date'])

df['Order_Year'] = df['Order_Date'].dt.year

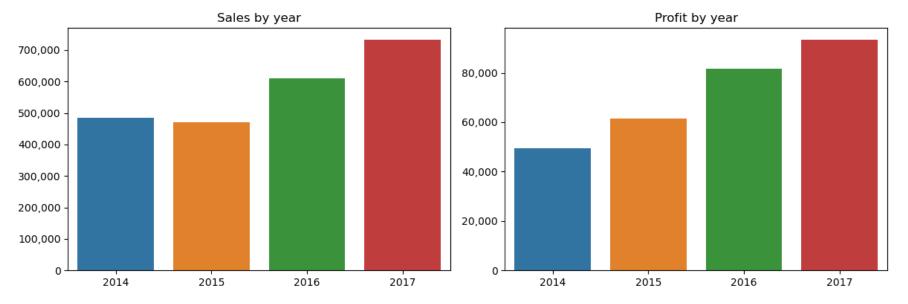
grup_y = df.groupby(['Order_Year']).sum().reset_index()
grup_y
```

```
Out[5]:
            Order_Year
                        Row ID Postal Code
                                               Sales Quantity Discount
                                                                        Profit
         0
                 2014 9904015
                                 113271247 484247.5
                                                        7581
                                                                 315.5 49544.0
                 2015 10413696
                                 111208247 470532.5
                                                        7979
                                                                 327.1 61618.6
         2
                 2016 12778804
                                 141003420 609205.6
                                                        9837
                                                                  400.3 81795.2
         3
                 2017 16848500
                                 186089738 733215.3
                                                        12476
                                                                 518.2 93439.3
```

Q1. What is the Sales and Profit growth Year-on-Year?

To answer this question, we can group the data by 'Sales' and calculate the mean price for each Year:

```
# Sales by year
 plt.figure(figsize=(12,4), tight_layout=True)
 plt.subplot(1,2,1)
 g1 = sns.barplot(x= 'Order_Year', y= 'Sales', data= grup_y)
 g1.set(xlabel=None, ylabel=None, title='Sales by year')
 current_values = plt.gca().get_yticks()
 plt.gca().set\_yticklabels(['{:,.0f}'.format(x) for x in current\_values])
 # Profit by year
 plt.subplot(1,2,2)
 g2 = sns.barplot(x= 'Order_Year', y= 'Profit', data= grup_y)
 g2.set(xlabel=None, ylabel=None, title='Profit by year')
 current_values = plt.gca().get_yticks()
 plt.gca().set_yticklabels(['{:,.0f}'.format(x) for x in current_values])
 # plt.savefig('fig1.png')
 plt.show()
C:\Users\Mohan Sharma\AppData\Local\Temp\ipykernel_20640\3970194423.py:7: UserWarning:
FixedFormatter should only be used together with FixedLocator
C:\Users\Mohan Sharma\AppData\Local\Temp\ipykernel 20640\3970194423.py:13: UserWarning:
FixedFormatter should only be used together with FixedLocator
```



Q1. What region is having the highest sale in US?

To answer this question, we can group the data by 'Region' and calculate the Total Sale for each region:

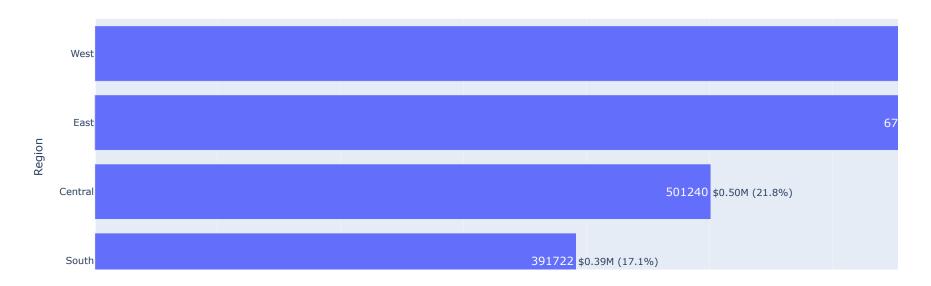
```
In [8]:
        import plotly.graph_objects as go
        import plotly.express as px
        # Group the data by region and calculate total sales
        sales_by_region = df.groupby('Region')['Sales'].sum().reset_index()
        # Convert sales to million dollars
        sales by region['Sales Million'] = sales by region['Sales'] / 1 000 000
        # Sort the data by sales in descending order
        sales_by_region = sales_by_region.sort_values('Sales', ascending=True)
        # Create a horizontal bar chart for sales by region
        fig = go.Figure(data=go.Bar(y=sales_by_region['Region'], x=sales_by_region['Sales_Million'], orientation='h'))
        # Add labels inside the bars with increased font size
        fig.update_traces(text=round(sales_by_region['Sales']), textposition='inside', textfont_size=14)
        # Add percentages outside the bars with increased font size
        total_sales = sales_by_region['Sales_Million'].sum()
        fig.update_layout(annotations=[
            go.layout.Annotation(
                y=r, x=s,
                text=f"${s:.2f}M\n({s/total_sales*100:.1f}%)",
                showarrow=False,
                xanchor='left',
                yanchor='middle',
```

```
font=dict(size=12)
  ) for r, s in zip(sales_by_region['Region'], sales_by_region['Sales_Million'])
])

# Set the chart title and axes LabeLs
fig.update_layout(
   title='Total Sales by Region',
   xaxis_title='Sales (Million $)',
   yaxis_title='Region'
)

# Show the chart
fig.show()
```

Total Sales by Region



Inference from the Total Sales Bar Plot By Region;

According to the plot

- Sount region have maximum sales and total contribution of 31.6%.
- West region have minimum sales and total contribution of 17.1 %

Q2. What Category is having the highest sale in US?

To answer this question, we can group the data by 'Category' and calculate the Total Sale for each region:

```
In [9]: import plotly.graph_objects as go
        import plotly.express as px
        # Group the data by Category and calculate total sales
         sales_by_region = df.groupby('Category')['Sales'].sum().reset_index()
        # Convert sales to million dollars
         sales by region['Sales Million'] = sales by region['Sales'] / 1 000 000
        # Sort the data by sales in descending order
        sales_by_region = sales_by_region.sort_values('Sales', ascending=True)
        # Create a horizontal bar chart for sales by Category
        fig = go.Figure(data=go.Bar(y=sales_by_region['Category'], x=sales_by_region['Sales_Million'], orientation='h'))
        # Add labels inside the bars with increased font size
        fig.update_traces(text=round(sales_by_region['Sales']), textposition='inside', textfont_size=14)
        # Add percentages outside the bars with increased font size
        total_sales = sales_by_region['Sales_Million'].sum()
        fig.update_layout(annotations=[
            go.layout.Annotation(
                y=r, x=s,
                text=f"${s:.2f}M\n({s/total sales*100:.1f}%)",
                showarrow=False,
                xanchor='left',
                yanchor='middle',
                font=dict(size=12)
            ) for r, s in zip(sales_by_region['Category'], sales_by_region['Sales_Million'])
        1)
        # Set the chart title and axes labels
        fig.update_layout(
            title='Total Sales by Category',
            xaxis title='Sales (Million $)',
            yaxis title='Category'
        # Show the chart
        fig.show()
```

Total Sales by Category



Inference from the Total Sales Bar Plot By Category;

According to the plot

- Technology Product is having the maximum sales and total contribution of 36.4%.
- Office Product is having the minimum sales and total contribution of 31.3 %

Q3. Total Sales by categories across Region?

To answer this question, we can group the data by 'Region' and calculate the Total Sale breakdown by Category for each region:

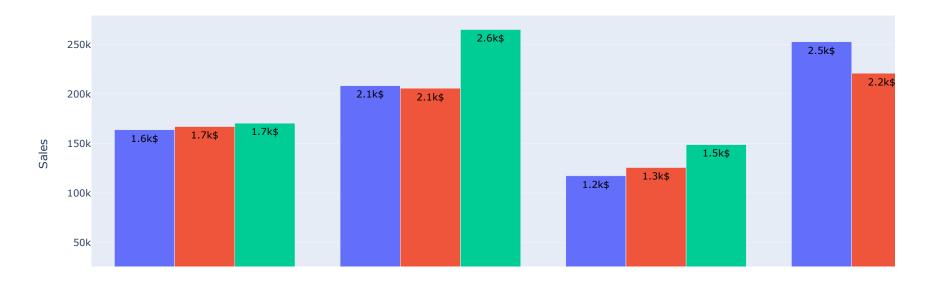
```
import plotly.graph_objects as go

# Group the data by region and category and calculate total sales
sales_by_region_category = df.groupby(['Region', 'Category'])['Sales'].sum().reset_index()

# Create a bar chart for region vs. sales anchored at category
fig = go.Figure()
```

```
# Add labels as separate bars
for i, category in enumerate(sales_by_region_category['Category'].unique()):
    category_data = sales_by_region_category[sales_by_region_category['Category'] == category]
    x_values = category_data['Region']
    y_values = category_data['Sales']
    labels = [f"{sales/100_000:.1f}k$" for sales in y_values]
    fig.add_trace(go.Bar(
       x=x_values,
       y=y_values,
       text=labels,
        name=category,
        textposition='auto',
        textfont=dict(color='black', size=12),
        showlegend= True,
        opacity=1
    ))
# Set the chart title and axes labels
fig.update_layout(
    title='Region vs. Sales Anchored at Category',
    xaxis_title='Region',
    yaxis_title='Sales'
# Show the chart
fig.show()
```

Region vs. Sales Anchored at Category



Q4. Total Profit by categories across Region?

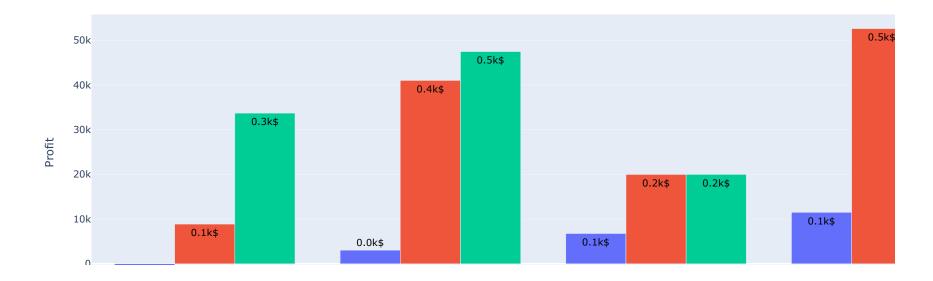
To answer this question, we can group the data by 'Region' and calculate the Total profit breakdown by Category for each region:

```
y=y_values,
    text=labels,
    name=category,
    textposition='auto',
    textfont=dict(color='black', size=12),
    showlegend= True,
    opacity=1
))

# Set the chart title and axes Labels
fig.update_layout(
    title='Region vs. Profit Anchored at Category',
    xaxis_title='Region',
    yaxis_title='Profit'
)

# Show the chart
fig.show()
```

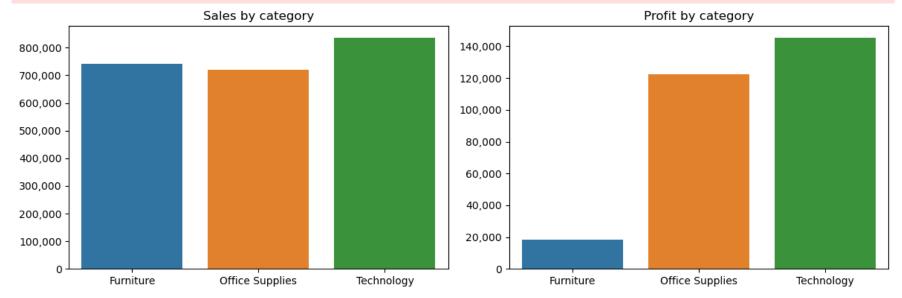
Region vs. Profit Anchored at Category



Q5. Which was the most Sales and Profit category, and which sold the most?

To answer this question, we can group the data by 'Region' and calculate the Total profit breakdown by Category for each region:

```
In [15]: grup_cat = df.groupby(['Category']).sum().reset_index()
         grup cat
         plt.figure(figsize=(12,4), tight_layout=True)
         plt.subplot(1,2,1)
         # Sales by category
         g3 = sns.barplot(x='Category', y='Sales', data=grup_cat)
         g3.set(xlabel=None, ylabel=None, title='Sales by category')
          current_values = plt.gca().get_yticks()
         plt.gca().set_yticklabels(['{:,.0f}'.format(x) for x in current_values])
         # Profit by category
         plt.subplot(1,2,2)
         g4 = sns.barplot(x='Category', y='Profit', data=grup_cat)
         g4.set(xlabel=None, ylabel=None, title='Profit by category')
          current_values = plt.gca().get_yticks()
         plt.gca().set_yticklabels(['{:,.0f}'.format(x) for x in current_values])
         # plt.savefig('fig4.png')
         plt.show()
         C:\Users\Mohan Sharma\AppData\Local\Temp\ipykernel_20640\2779832684.py:10: UserWarning:
         FixedFormatter should only be used together with FixedLocator
         C:\Users\Mohan Sharma\AppData\Local\Temp\ipykernel_20640\2779832684.py:16: UserWarning:
         FixedFormatter should only be used together with FixedLocator
```



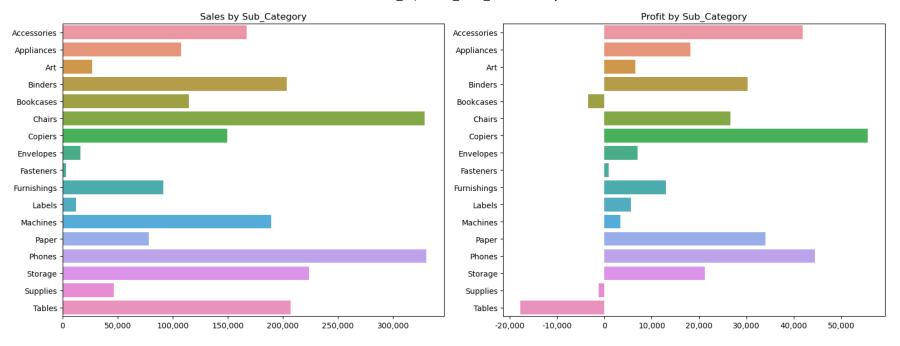
Inference from the above bar Plot;

Technology sold the most, and was the most profitable category. But furniture, despite having sold more than Office Supplies, it made less profit than it.

Q6. Which was the most profitable sub-category, and which sold the most?

To answer this question, we can group the data by 'Region' and calculate the Total profit breakdown by Category for each region:

```
In [16]: grup_subcat = df.groupby(['Sub_Category']).sum().reset_index()
         plt.figure(figsize=(16,6), tight_layout=True)
         plt.subplot(1,2,1)
         # Sales by sub-category
         g5 = sns.barplot(y='Sub_Category', x='Sales', data=grup_subcat, orient='h')
         g5.set(xlabel=None, ylabel=None, title='Sales by Sub Category')
          current_values = plt.gca().get_xticks()
         plt.gca().set_xticklabels(['{:,.0f}'.format(x) for x in current_values])
         # Profit by subcategoria
         plt.subplot(1,2,2)
         g6 = sns.barplot(y='Sub_Category', x='Profit', data=grup_subcat, orient='h')
         g6.set(xlabel=None, ylabel=None, title='Profit by Sub_Category')
          current_values = plt.gca().get_xticks()
         plt.gca().set xticklabels(['{:,.0f}'.format(x) for x in current values])
         plt.show()
         C:\Users\Mohan Sharma\AppData\Local\Temp\ipykernel_20640\1394607245.py:9: UserWarning:
         FixedFormatter should only be used together with FixedLocator
         C:\Users\Mohan Sharma\AppData\Local\Temp\ipykernel_20640\1394607245.py:15: UserWarning:
         FixedFormatter should only be used together with FixedLocator
```



Inference from the above bar Plot;

- 1. Looking at the graphs, we can see that the subcategories who sold the most, didn't make the highest profit.
- 2. Phones was the most sold sub-category, and Copiers was the most profitable sub-category.
- 3. Tables reached almost 800,000 in sales, but made more than 50,000 in negative profit.

Q7. What factors affect Sale the most?

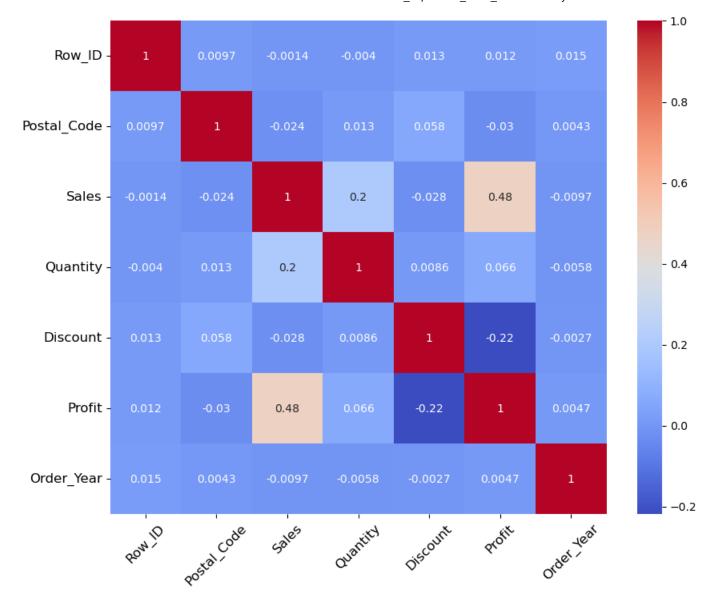
To answer this question, To answer this question, we can create a correlation matrix using the px.imshow() function in Plotly:

```
In [24]: # Using Seaborn Library
# Create heatmap

corr_matrix = df.corr()
fig, ax = plt.subplots(figsize=(10,8))
sns.heatmap(corr_matrix, annot=True, cmap='coolwarm', ax=ax)

# Customize plot
# ax.set_title("Correlation Matrix for US Retail Data", fontsize=16)
ax.tick_params(axis='x', labelsize=12, rotation=45)
ax.tick_params(axis='y', labelsize=12, rotation=0)

# Show the plot
plt.show()
```



Inference from the Corr Chart Price with other Cols

According to the correlation matrix, the variables don't really affect each other.

In []: