

```
In [1]: # Import Libraries:
import numpy as np
import pandas as pd
from matplotlib import pyplot as plt
from matplotlib import style
import seaborn as sns
```

```
In [2]: # Read the CSV file from the file C drive
df = pd.read_csv(r"C:\Users\HP\Downloads\Superstore data.csv")
df
```

Out[2]:

| | Row ID | Order ID | Order Date | Ship Date | Ship Mode | Customer ID | Customer Name | Segment | Country | |
|------|--------|----------------|------------|------------|----------------|-------------|------------------|-----------|---------------|-----|
| 0 | 1 | CA-2013-152156 | 09-11-2013 | 12-11-2013 | Second Class | CG-12520 | Claire Gute | Consumer | United States | Hi |
| 1 | 2 | CA-2013-152156 | 09-11-2013 | 12-11-2013 | Second Class | CG-12520 | Claire Gute | Consumer | United States | Hi |
| 2 | 3 | CA-2013-138688 | 13-06-2013 | 17-06-2013 | Second Class | DV-13045 | Darrin Van Huff | Corporate | United States | Los |
| 3 | 4 | US-2012-108966 | 11-10-2012 | 18-10-2012 | Standard Class | SO-20335 | Sean O'Donnell | Consumer | United States | Lé |
| 4 | 5 | US-2012-108966 | 11-10-2012 | 18-10-2012 | Standard Class | SO-20335 | Sean O'Donnell | Consumer | United States | Lé |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 9989 | 9990 | CA-2011-110422 | 22-01-2011 | 24-01-2011 | Second Class | TB-21400 | Tom Boeckenhauer | Consumer | United States | |
| 9990 | 9991 | CA-2014-121258 | 27-02-2014 | 04-03-2014 | Standard Class | DB-13060 | Dave Brooks | Consumer | United States | Co |
| 9991 | 9992 | CA-2014-121258 | 27-02-2014 | 04-03-2014 | Standard Class | DB-13060 | Dave Brooks | Consumer | United States | Co |
| 9992 | 9993 | CA-2014-121258 | 27-02-2014 | 04-03-2014 | Standard Class | DB-13060 | Dave Brooks | Consumer | United States | Co |
| 9993 | 9994 | CA-2014-119914 | 05-05-2014 | 10-05-2014 | Second Class | CC-12220 | Chris Cortes | Consumer | United States | We |

9994 rows × 20 columns



In [3]: *#Information about the each column and its data types*
df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9994 entries, 0 to 9993
Data columns (total 20 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Row ID                9994 non-null   int64
1   Order ID              9991 non-null   object
2   Order Date            9993 non-null   object
3   Ship Date              9993 non-null   object
4   Ship Mode              9991 non-null   object
5   Customer ID            9994 non-null   object
6   Customer Name          9993 non-null   object
7   Segment                9994 non-null   object
8   Country                9994 non-null   object
9   City                  9994 non-null   object
10  State                  9994 non-null   object
11  Postal Code            9994 non-null   int64
12  Region                 9994 non-null   object
13  Product ID             9994 non-null   object
14  Category                9994 non-null   object
15  Sub-Category           9994 non-null   object
16  Sales                  9989 non-null   float64
17  Quantity               9994 non-null   int64
18  Discount               9994 non-null   float64
19  Profit                 9994 non-null   float64
dtypes: float64(3), int64(3), object(14)
memory usage: 1.5+ MB
```

In [4]: *# It shows the number of number of null values in the each column*
null_counts = df.isnull().sum()
print("Rows with null values:")
null_counts

Rows with null values:

```
Out[4]: Row ID                0
Order ID                3
Order Date              1
Ship Date                1
Ship Mode                3
Customer ID              0
Customer Name            1
Segment                  0
Country                  0
City                     0
State                    0
Postal Code              0
Region                   0
Product ID               0
Category                 0
Sub-Category             0
Sales                    5
Quantity                 0
Discount                 0
Profit                   0
dtype: int64
```

```
In [5]: # Drop the null- values rows
df.dropna(inplace=True)
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 9983 entries, 0 to 9993
Data columns (total 20 columns):
 #   Column                Non-Null Count  Dtype  
---  -
 0   Row ID                9983 non-null  int64  
 1   Order ID              9983 non-null  object  
 2   Order Date            9983 non-null  object  
 3   Ship Date              9983 non-null  object  
 4   Ship Mode              9983 non-null  object  
 5   Customer ID           9983 non-null  object  
 6   Customer Name          9983 non-null  object  
 7   Segment                9983 non-null  object  
 8   Country                9983 non-null  object  
 9   City                   9983 non-null  object  
10   State                  9983 non-null  object  
11   Postal Code            9983 non-null  int64  
12   Region                 9983 non-null  object  
13   Product ID             9983 non-null  object  
14   Category               9983 non-null  object  
15   Sub-Category           9983 non-null  object  
16   Sales                   9983 non-null  float64 
17   Quantity               9983 non-null  int64  
18   Discount                9983 non-null  float64 
19   Profit                  9983 non-null  float64 
dtypes: float64(3), int64(3), object(14)
memory usage: 1.6+ MB
```

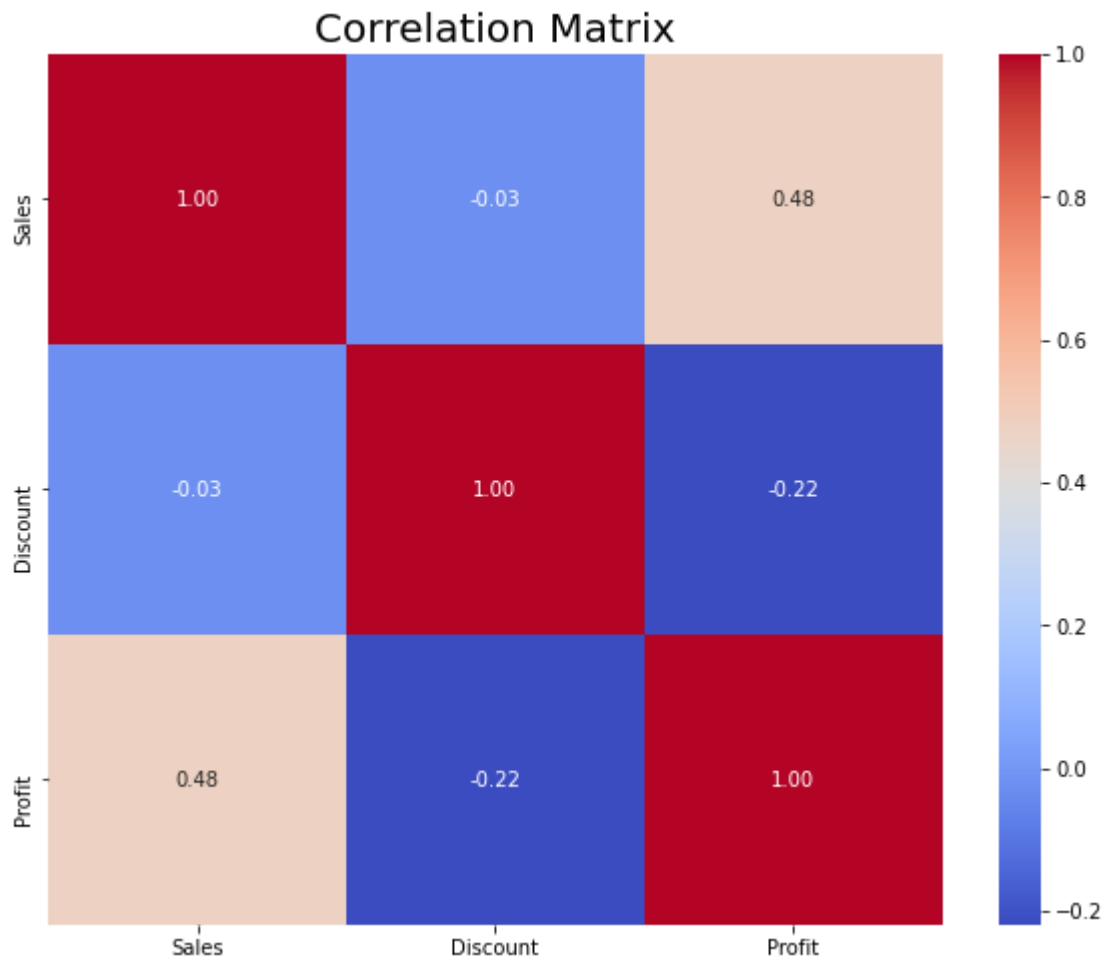
```
In [6]: #Profit margin :  
  
df['Profit margin'] = df['Profit'] / df['Sales']*100  
df
```

Out[6]:

| | Row ID | Order ID | Order Date | Ship Date | Ship Mode | Customer ID | Customer Name | Segment | Country | |
|------|--------|----------------|------------|------------|----------------|-------------|------------------|-----------|---------------|-----|
| 0 | 1 | CA-2013-152156 | 09-11-2013 | 12-11-2013 | Second Class | CG-12520 | Claire Gute | Consumer | United States | Hi |
| 1 | 2 | CA-2013-152156 | 09-11-2013 | 12-11-2013 | Second Class | CG-12520 | Claire Gute | Consumer | United States | Hi |
| 2 | 3 | CA-2013-138688 | 13-06-2013 | 17-06-2013 | Second Class | DV-13045 | Darrin Van Huff | Corporate | United States | Los |
| 3 | 4 | US-2012-108966 | 11-10-2012 | 18-10-2012 | Standard Class | SO-20335 | Sean O'Donnell | Consumer | United States | Le |
| 4 | 5 | US-2012-108966 | 11-10-2012 | 18-10-2012 | Standard Class | SO-20335 | Sean O'Donnell | Consumer | United States | Le |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 9989 | 9990 | CA-2011-110422 | 22-01-2011 | 24-01-2011 | Second Class | TB-21400 | Tom Boeckenhauer | Consumer | United States | |
| 9990 | 9991 | CA-2014-121258 | 27-02-2014 | 04-03-2014 | Standard Class | DB-13060 | Dave Brooks | Consumer | United States | Co |
| 9991 | 9992 | CA-2014-121258 | 27-02-2014 | 04-03-2014 | Standard Class | DB-13060 | Dave Brooks | Consumer | United States | Co |
| 9992 | 9993 | CA-2014-121258 | 27-02-2014 | 04-03-2014 | Standard Class | DB-13060 | Dave Brooks | Consumer | United States | Co |
| 9993 | 9994 | CA-2014-119914 | 05-05-2014 | 10-05-2014 | Second Class | CC-12220 | Chris Cortes | Consumer | United States | We |

9983 rows × 21 columns

```
In [7]: # Correlation of sales, discount and profit
correlation_matrix = df[['Sales', 'Discount', 'Profit']].corr()
plt.figure(figsize=(10, 8))
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt=".2f")
plt.title('Correlation Matrix', fontsize=20)
plt.show()
```



```
In [8]: # Top 10 Customers by Sales:
top_customers = df.groupby('Customer Name')['Sales'].sum().sort_values(ascending=False)
print('Top 10 Customers Based on Sales:')
print(top_customers)
```

Top 10 Customers Based on Sales:

| Customer Name | Sales |
|--------------------|-----------|
| Sean Miller | 25043.050 |
| Tamara Chand | 19052.218 |
| Raymond Buch | 15117.339 |
| Tom Ashbrook | 14595.620 |
| Adrian Barton | 14473.571 |
| Ken Lonsdale | 14175.229 |
| Sanjit Chand | 14142.334 |
| Hunter Lopez | 12873.298 |
| Sanjit Engle | 12209.438 |
| Christopher Conant | 12129.072 |

Name: Sales, dtype: float64

```
In [9]: # Top 10 Best Selling Products:
best_selling_products = df.groupby('Product ID')['Sales'].sum().sort_values
print('Top 10 Best Selling Products:')
print(best_selling_products)
```

Top 10 Best Selling Products:

Product ID

| | |
|-----------------|-----------|
| TEC-CO-10004722 | 61599.824 |
| OFF-BI-10003527 | 27453.384 |
| TEC-MA-10002412 | 22638.480 |
| FUR-CH-10002024 | 21870.576 |
| OFF-BI-10001359 | 19823.479 |
| OFF-BI-10000545 | 19024.500 |
| TEC-CO-10001449 | 18839.686 |
| TEC-MA-10001127 | 18374.895 |
| OFF-BI-10004995 | 17965.068 |
| OFF-SU-10000151 | 17030.312 |

Name: Sales, dtype: float64

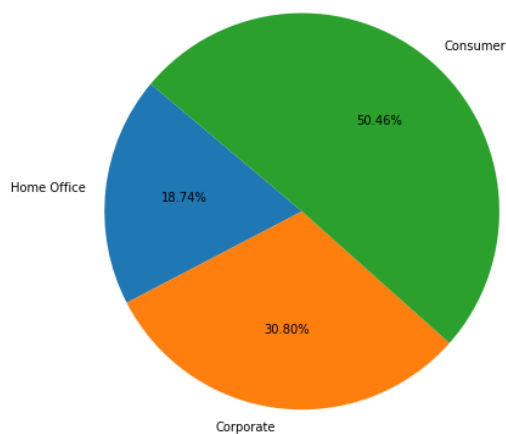
```
In [10]: #Sales and profit distribution by Segment

segment_wise_sales = df.groupby('Segment')['Sales'].sum().sort_values(ascending=True)
segment_wise_profit = df.groupby('Segment')['Profit'].sum().sort_values(ascending=True)
fig, axs = plt.subplots(1, 2, figsize=(16, 8))
axs[0].pie(segment_wise_sales, labels=segment_wise_sales.index, autopct='%1.1f%%')
axs[0].set_title('Segment-wise Sales Distribution', fontsize=16)

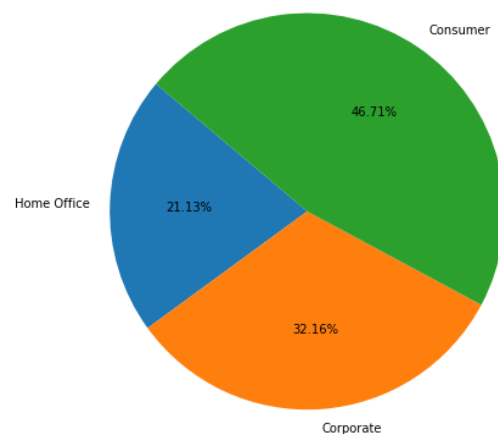
axs[1].pie(segment_wise_profit, labels=segment_wise_profit.index, autopct='%1.1f%%')
axs[1].set_title('Segment-wise Profit Distribution', fontsize=16)

plt.show()
```

Segment-wise Sales Distribution

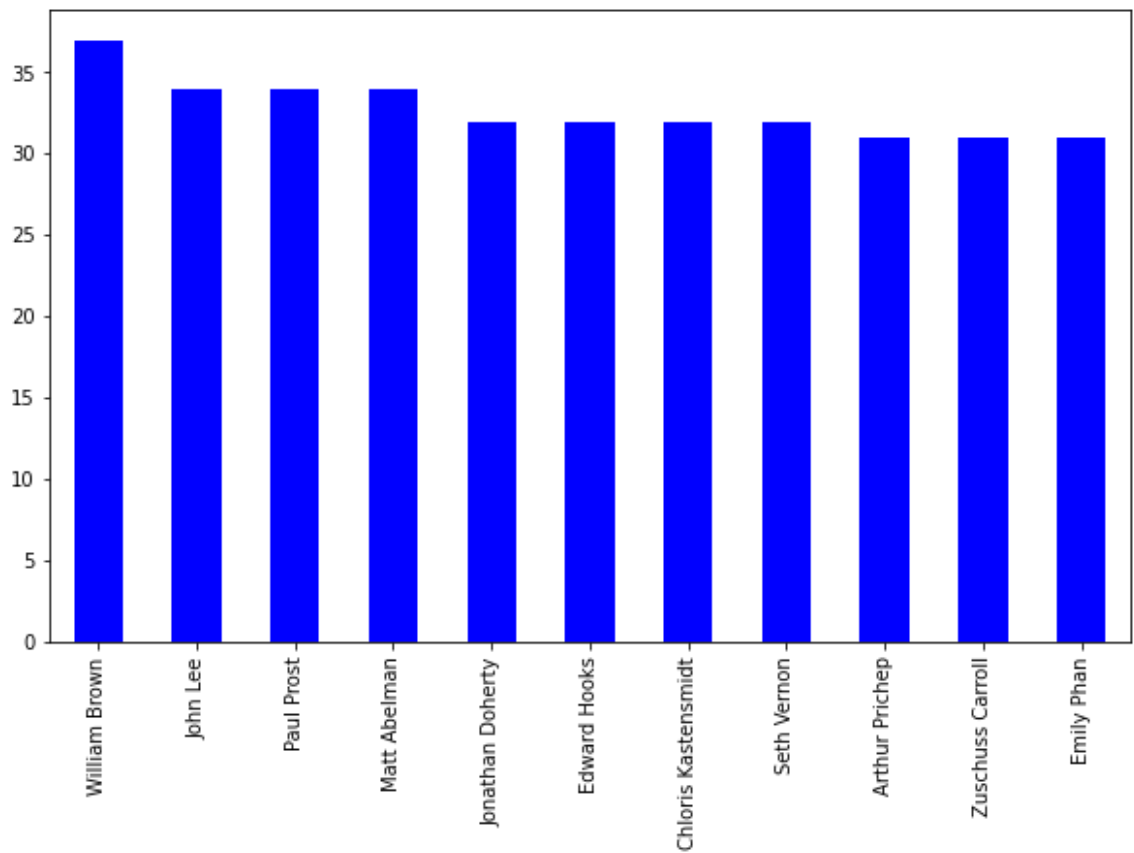


Segment-wise Profit Distribution



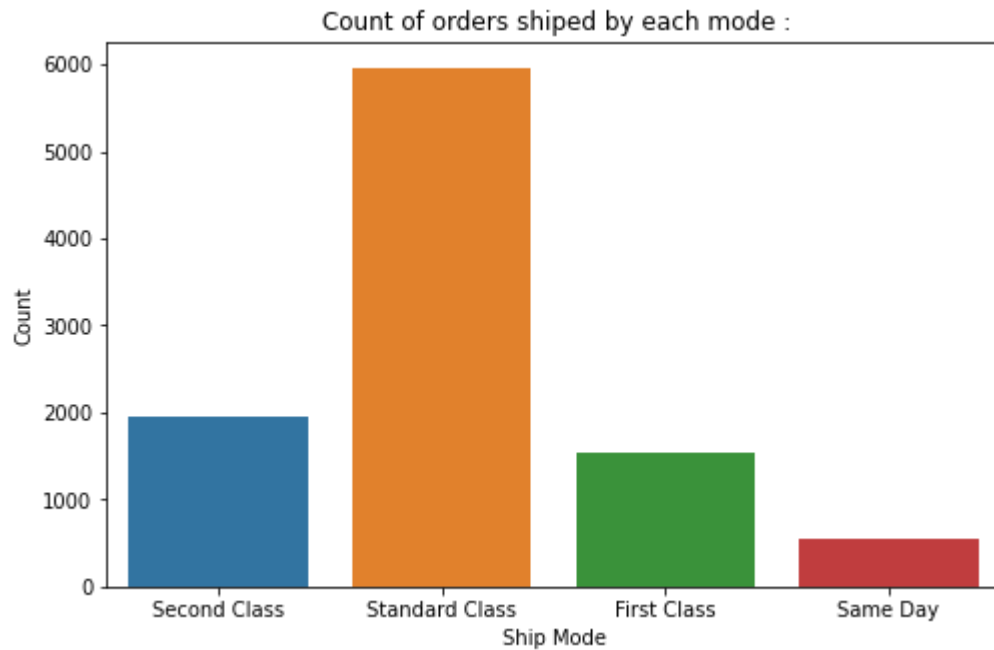
```
In [11]: # The frequency of purchases for each specific customer  
df['Customer Name'].value_counts().head(11).plot(kind='bar', figsize=(10, 6
```

Out[11]: <Axes: >



In [12]: *# Count of orders shipped by each mode :*

```
plt.figure(figsize=(8, 5))
sns.countplot(x='Ship Mode', data=df)
plt.title(' Count of orders shipped by each mode :')
plt.xlabel('Ship Mode')
plt.ylabel('Count')
plt.show()
```




```
In [13]: # Top 10 most profitable and 10 least profitable cities:
most_profitable_cities = df.groupby('City')['Profit'].sum().sort_values(asc
least_profitable_cities = df.groupby('City')['Profit'].sum().sort_values().l
print('Top 10 Most Profitable Cities:')
print(most_profitable_cities)
print('\n10 Least Profitable Cities:')
print(least_profitable_cities)
```

Top 10 Most Profitable Cities:

City

New York City 62026.0741

Los Angeles 30287.0919

Seattle 28858.1232

San Francisco 17489.4262

Detroit 13181.7908

Lafayette 10018.3876

Jackson 7581.6828

Atlanta 6993.6629

Minneapolis 6824.5846

San Diego 6377.1960

Name: Profit, dtype: float64

10 Least Profitable Cities:

City

Philadelphia -13837.7674

Houston -10153.5485

San Antonio -7299.0502

Lancaster -7239.0684

Chicago -6654.5688

Burlington -3622.8772

Dallas -2846.5257

Phoenix -2790.8832

Aurora -2691.7386

Jacksonville -2323.8350

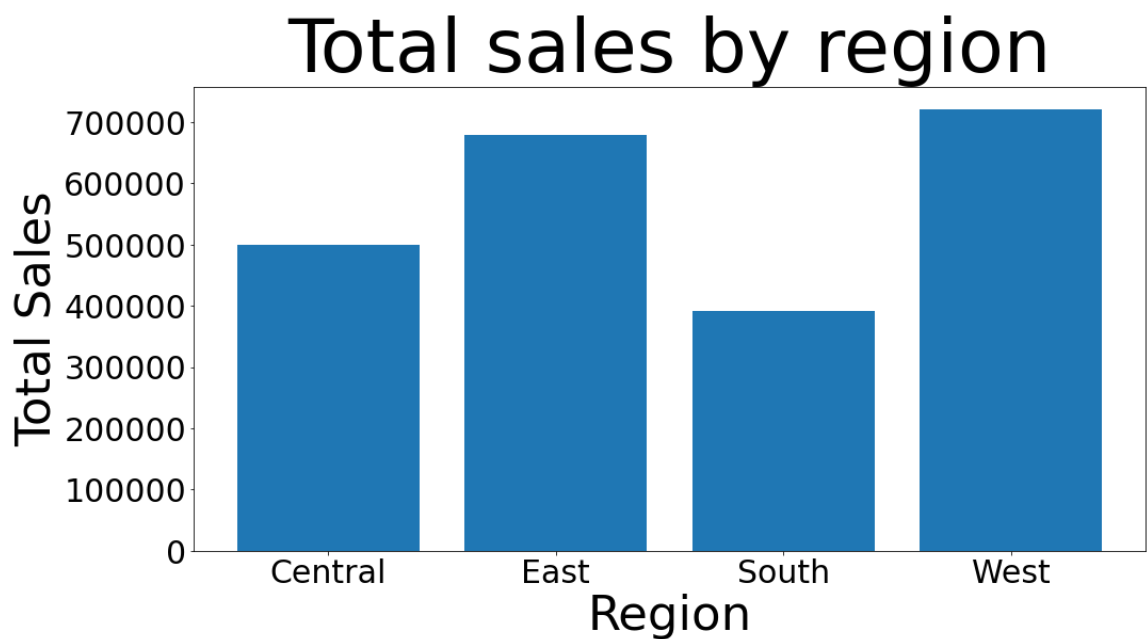
Name: Profit, dtype: float64

In [16]: *#Total sales by region :*

```
plt.figure(figsize=(16, 8))
d1 = df.groupby('Region', as_index = False)['Sales'].sum()
plt.bar(d1.Region,d1.Sales)

plt.xlabel('Region', fontsize=45)
plt.ylabel('Total Sales', fontsize=45)
plt.title('Total sales by region', fontsize=70)
plt.xticks(fontsize=30)
plt.yticks(fontsize=30)
```

Out[16]: (array([0., 100000., 200000., 300000., 400000., 500000., 600000.,
700000., 800000.]),
[Text(0, 0.0, '0'),
Text(0, 100000.0, '100000'),
Text(0, 200000.0, '200000'),
Text(0, 300000.0, '300000'),
Text(0, 400000.0, '400000'),
Text(0, 500000.0, '500000'),
Text(0, 600000.0, '600000'),
Text(0, 700000.0, '700000'),
Text(0, 800000.0, '800000')])



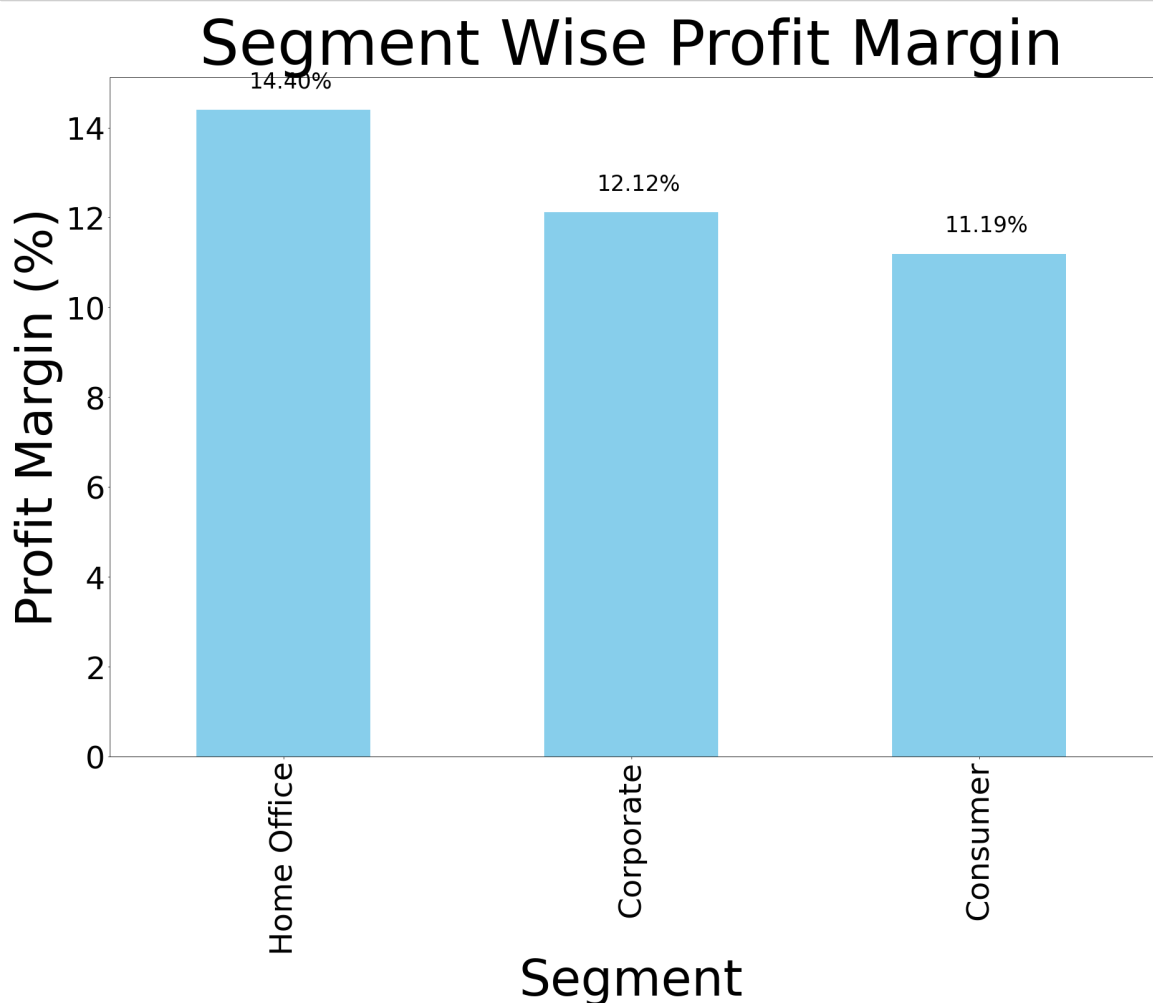
```
In [17]: # Segment wise profit margin :

df['Profit Margin'] = (df['Profit'] / df['Sales']) * 100

segment_profit_margin = df.groupby('Segment')['Profit Margin'].mean().sort_
plt.figure(figsize=(30,20))
bars = segment_profit_margin.plot(kind='bar', color='skyblue')
plt.xlabel('Segment', fontsize=80)
plt.ylabel('Profit Margin (%)', fontsize=80)
plt.title('Segment Wise Profit Margin', fontsize=100)
plt.xticks(fontsize=50)
plt.yticks(fontsize=50)

for bar in bars.patches:
    plt.text(bar.get_x() + bar.get_width() / 2 - 0.1, bar.get_height() + 0.1,
             f'{bar.get_height():.2f}%', fontsize=35,color='black')

plt.show()
```



In [18]: *#Sub-category wise sales and its percentage:*

```
sub_category_wise_sales = df.groupby('Sub-Category')['Sales'].sum().sort_values(ascending=True)
print('Sub-category wise Sales :')
print()
print(sub_category_wise_sales)

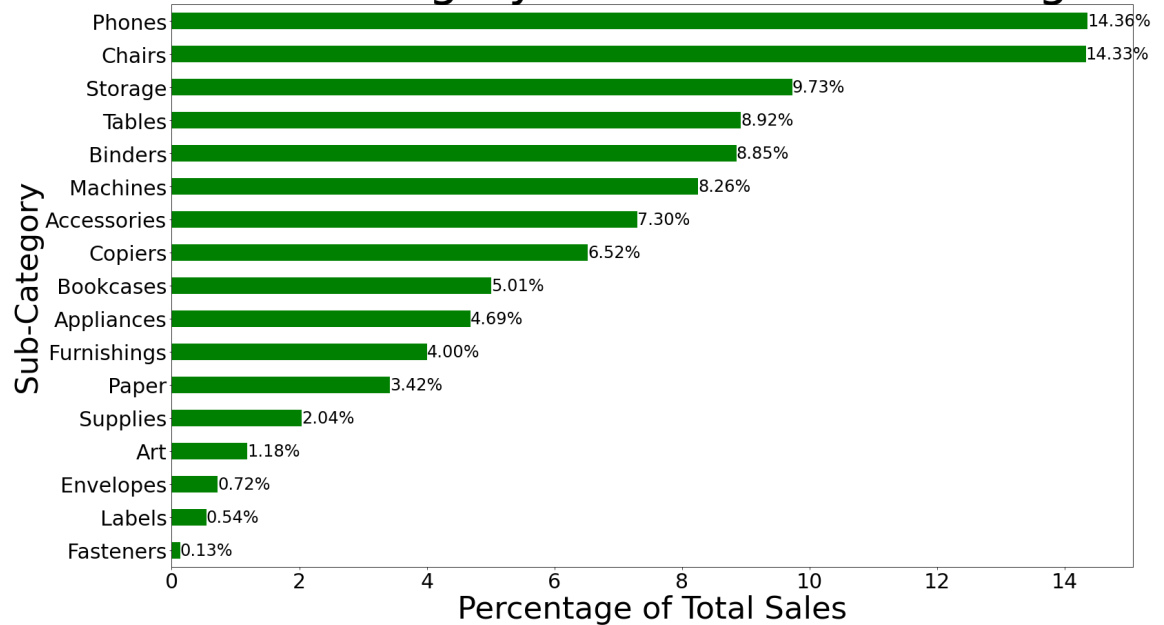
total_sales = df['Sales'].sum()
sub_category_wise_sales_percentage = (df.groupby('Sub-Category')['Sales'].sum() / total_sales) * 100
plt.figure(figsize=(25, 15))
ax = sub_category_wise_sales_percentage.sort_values(ascending=True).plot(kind='bar')
for rect in ax.patches:
    width = rect.get_width()
    plt.text(width, rect.get_y() + rect.get_height() / 2, f'{width:.2f}%', ha='center')
plt.xlabel('Percentage of Total Sales', fontsize=45)
plt.ylabel('Sub-Category', fontsize=45)
plt.title('Sub-Category Wise Sales Percentage', fontsize=70)
plt.xticks(fontsize=30)
plt.yticks(fontsize=30)
plt.show()
```

Sub-category wise Sales :

| Sub-Category | Sales |
|--------------|-------------|
| Phones | 329067.6380 |
| Chairs | 328449.1030 |
| Storage | 222951.1680 |
| Tables | 204471.8180 |
| Binders | 202953.4450 |
| Machines | 189238.6310 |
| Accessories | 167380.3180 |
| Copiers | 149528.0300 |
| Bookcases | 114879.9963 |
| Appliances | 107463.3510 |
| Furnishings | 91663.2040 |
| Paper | 78463.6540 |
| Supplies | 46673.5380 |
| Art | 27118.7920 |
| Envelopes | 16476.4020 |
| Labels | 12486.3120 |
| Fasteners | 3024.2800 |

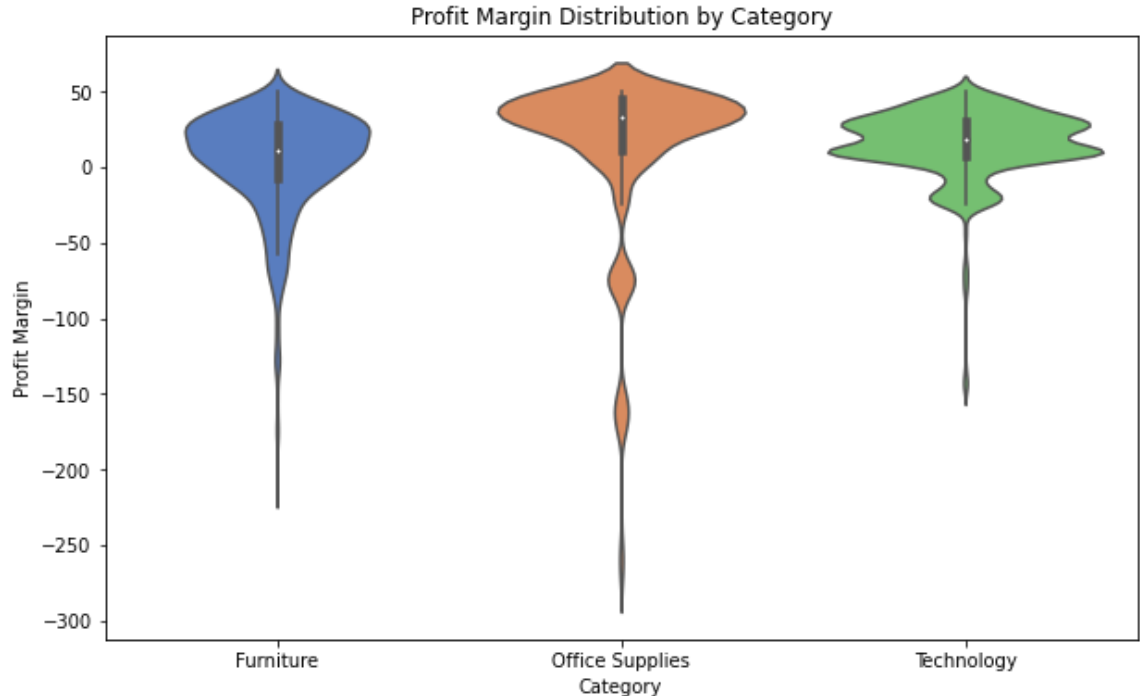
Name: Sales, dtype: float64

Sub-Category Wise Sales Percentage



In [19]: *#Profit Margin Distribution by Category :*

```
plt.figure(figsize=(10, 6))
sns.violinplot(x='Category', y='Profit margin', data=df, palette='muted')
plt.title('Profit Margin Distribution by Category')
plt.xlabel('Category')
plt.ylabel('Profit Margin')
plt.show()
```



In []:

