

EDA ON SALES ANALYSIS

```
# Import Necessary Python Libraries:
print('Import the essential libraries for data manipulation, analysis,
and visualization')
```

Import the essential libraries for data manipulation, analysis, and visualization

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt # visualizing data
%matplotlib inline
import seaborn as sns
```

Load the Dataset

```
df=pd.read_csv(r'C:\Users\Vipin\Downloads\End To End Power BI Project\
Python\E Commerce\supermarket_sales.csv')
df
```

	Invoice ID	Branch	City	Customer type	Gender	\
0	750-67-8428	A	Yangon	Member	Female	
1	226-31-3081	C	Naypyitaw	Normal	Female	
2	631-41-3108	A	Yangon	Normal	Male	
3	123-19-1176	A	Yangon	Member	Male	
4	373-73-7910	A	Yangon	Normal	Male	
..	
995	233-67-5758	C	Naypyitaw	Normal	Male	
996	303-96-2227	B	Mandalay	Normal	Female	
997	727-02-1313	A	Yangon	Member	Male	
998	347-56-2442	A	Yangon	Normal	Male	
999	849-09-3807	A	Yangon	Member	Female	

	Product line	Unit price	Quantity	Tax 5%	Total
0	Health and beauty	74.69	7	26.1415	548.9715
1	Electronic accessories	15.28	5	3.8200	80.2200

2	Home and lifestyle	46.33	7	16.2155	340.5255
3	Health and beauty	58.22	8	23.2880	489.0480
4	Sports and travel	86.31	7	30.2085	634.3785
..
995	Health and beauty	40.35	1	2.0175	42.3675
996	Home and lifestyle	97.38	10	48.6900	1022.4900
997	Food and beverages	31.84	1	1.5920	33.4320
998	Home and lifestyle	65.82	1	3.2910	69.1110
999	Fashion accessories	88.34	7	30.9190	649.2990
	Date	Time	Payment	cogs	gross margin percentage \
0	05-01-19	13:08	Ewallet	522.83	4.761905
1	08-03-19	10:29	Cash	76.40	4.761905
2	03-03-19	13:23	Credit card	324.31	4.761905
3	27-01-19	20:33	Ewallet	465.76	4.761905
4	08-02-19	10:37	Ewallet	604.17	4.761905
..
995	29-01-19	13:46	Ewallet	40.35	4.761905
996	02-03-19	17:16	Ewallet	973.80	4.761905
997	09-02-19	13:22	Cash	31.84	4.761905
998	22-02-19	15:33	Cash	65.82	4.761905
999	18-02-19	13:28	Cash	618.38	4.761905
	gross income	Rating			
0	26.1415	9.1			
1	3.8200	9.6			
2	16.2155	7.4			
3	23.2880	8.4			
4	30.2085	5.3			
..			
995	2.0175	6.2			
996	48.6900	4.4			
997	1.5920	7.7			
998	3.2910	4.1			
999	30.9190	6.6			
[1000 rows x 17 columns]					

Display Basic Information

```
# Display the first few rows of the dataframe
df.head()
```

	Invoice ID	Branch	City	Customer type	Gender	\
0	750-67-8428	A	Yangon	Member	Female	
1	226-31-3081	C	Naypyitaw	Normal	Female	
2	631-41-3108	A	Yangon	Normal	Male	
3	123-19-1176	A	Yangon	Member	Male	
4	373-73-7910	A	Yangon	Normal	Male	

	Product line	Unit price	Quantity	Tax 5%	Total
Date \					
0	Health and beauty	74.69	7	26.1415	548.9715
05-01-19					
1	Electronic accessories	15.28	5	3.8200	80.2200
08-03-19					
2	Home and lifestyle	46.33	7	16.2155	340.5255
03-03-19					
3	Health and beauty	58.22	8	23.2880	489.0480
27-01-19					
4	Sports and travel	86.31	7	30.2085	634.3785
08-02-19					

	Time	Payment	cogs	gross margin percentage	gross income
Rating					
0	13:08	Ewallet	522.83	4.761905	26.1415
9.1					
1	10:29	Cash	76.40	4.761905	3.8200
9.6					
2	13:23	Credit card	324.31	4.761905	16.2155
7.4					
3	20:33	Ewallet	465.76	4.761905	23.2880
8.4					
4	10:37	Ewallet	604.17	4.761905	30.2085
5.3					

```
# Display the last few rows of the dataframe
df.tail()
```

	Invoice ID	Branch	City	Customer type	Gender	
Product line \						
995	233-67-5758	C	Naypyitaw	Normal	Male	Health and beauty
996	303-96-2227	B	Mandalay	Normal	Female	Home and lifestyle
997	727-02-1313	A	Yangon	Member	Male	Food and beverages
998	347-56-2442	A	Yangon	Normal	Male	Home and

```
lifestyle
999 849-09-3807      A      Yangon      Member  Female  Fashion
accessories
```

	Unit price	Quantity	Tax 5%	Total	Date	Time
Payment \						
995	40.35	1	2.0175	42.3675	29-01-19	13:46
Ewallet						
996	97.38	10	48.6900	1022.4900	02-03-19	17:16
Ewallet						
997	31.84	1	1.5920	33.4320	09-02-19	13:22
Cash						
998	65.82	1	3.2910	69.1110	22-02-19	15:33
Cash						
999	88.34	7	30.9190	649.2990	18-02-19	13:28
Cash						

	cogs	gross margin percentage	gross income	Rating
995	40.35	4.761905	2.0175	6.2
996	973.80	4.761905	48.6900	4.4
997	31.84	4.761905	1.5920	7.7
998	65.82	4.761905	3.2910	4.1
999	618.38	4.761905	30.9190	6.6

```
#checking shape of Iris dataset
```

```
df.shape
```

```
(1000, 17)
```

```
#checking size of Iris dataset
```

```
df.size
```

```
17000
```

```
#checking what are the variables here
```

```
df.columns
```

```
Index(['Invoice ID', 'Branch', 'City', 'Customer type', 'Gender',
       'Product line', 'Unit price', 'Quantity', 'Tax 5%', 'Total',
       'Date',
       'Time', 'Payment', 'cogs', 'gross margin percentage', 'gross
income',
       'Rating'],
      dtype='object')
```

```
# Get a concise summary of the dataframe
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 1000 entries, 0 to 999
```

```
Data columns (total 17 columns):
```

#	Column	Non-Null Count		Dtype
----	-----	-----		-----
0	Invoice ID	1000	non-null	object
1	Branch	1000	non-null	object
2	City	1000	non-null	object
3	Customer type	1000	non-null	object
4	Gender	1000	non-null	object
5	Product line	1000	non-null	object
6	Unit price	1000	non-null	float64
7	Quantity	1000	non-null	int64
8	Tax 5%	1000	non-null	float64
9	Total	1000	non-null	float64
10	Date	1000	non-null	object
11	Time	1000	non-null	object
12	Payment	1000	non-null	object
13	cogs	1000	non-null	float64
14	gross margin percentage	1000	non-null	float64
15	gross income	1000	non-null	float64
16	Rating	1000	non-null	float64

dtypes: float64(7), int64(1), object(9)
memory usage: 132.9+ KB

Basic descriptive statistics
df.describe()

	Unit price	Quantity	Tax 5%	Total	cogs
\					
count	1000.000000	1000.000000	1000.000000	1000.000000	1000.000000
mean	55.672130	5.510000	15.379369	322.966749	307.58738
std	26.494628	2.923431	11.708825	245.885335	234.17651
min	10.080000	1.000000	0.508500	10.678500	10.17000
25%	32.875000	3.000000	5.924875	124.422375	118.49750
50%	55.230000	5.000000	12.088000	253.848000	241.76000
75%	77.935000	8.000000	22.445250	471.350250	448.90500
max	99.960000	10.000000	49.650000	1042.650000	993.00000

	gross margin percentage	gross income	Rating
count	1.000000e+03	1000.000000	1000.00000
mean	4.761905e+00	15.379369	6.97270
std	6.131498e-14	11.708825	1.71858
min	4.761905e+00	0.508500	4.00000
25%	4.761905e+00	5.924875	5.50000
50%	4.761905e+00	12.088000	7.00000

75%	4.761905e+00	22.445250	8.500000
max	4.761905e+00	49.650000	10.000000

Data Cleaning and Preprocessing

```
# Check for missing values
```

```
df.isnull().sum()
```

```
Invoice ID      0
Branch          0
City            0
Customer type   0
Gender          0
Product line    0
Unit price      0
Quantity        0
Tax 5%          0
Total           0
Date            0
Time            0
Payment         0
cogs            0
gross margin percentage  0
gross income    0
Rating          0
dtype: int64
```

```
# Check for duplicate values
```

```
df.duplicated()
```

```
0      False
1      False
2      False
3      False
4      False
```

```
...
```

```
995     False
996     False
997     False
998     False
999     False
```

```
Length: 1000, dtype: bool
```

```
# Check for duplicate values
```

```
df[df.duplicated()]
```

```
Empty DataFrame
Columns: [Invoice ID, Branch, City, Customer type, Gender, Product line, Unit price, Quantity, Tax 5%, Total, Date, Time, Payment, cogs, gross margin percentage, gross income, Rating]
Index: []
```

```
# drop duplicate values if available
df.drop_duplicates(inplace=True)
```

```
# check datatypes
df.dtypes
```

Invoice ID	object
Branch	object
City	object
Customer type	object
Gender	object
Product line	object
Unit price	float64
Quantity	int64
Tax 5%	float64
Total	float64
Date	object
Time	object
Payment	object
cogs	float64
gross margin percentage	float64
gross income	float64
Rating	float64

dtype: object

```
# Convert 'Date' column to datetime format
df['Date']=pd.to_datetime(df['Date'])
```

```
# Convert 'Time' column to datetime format (hours and minutes)
df['Time']=pd.to_datetime(df['Time'])
```

```
# check datatypes
df.dtypes
```

Invoice ID	object
Branch	object
City	object
Customer type	object
Gender	object
Product line	object
Unit price	float64
Quantity	int64
Tax 5%	float64
Total	float64
Date	datetime64[ns]

```
Time                datetime64[ns]
Payment             object
cogs                float64
gross margin percentage float64
gross income        float64
Rating              float64
dtype: object
```

```
# Extracting year, month, month_name and day from the 'Date' column
df['Year'] = df['Date'].dt.year
df['Month'] = df['Date'].dt.month
df['Day'] = df['Date'].dt.day
df['month_name'] = df['Date'].dt.month_name()
```

```
df.head(2)
```

	Invoice ID	Branch	City	Customer type	Gender \
0	750-67-8428	A	Yangon	Member	Female
1	226-31-3081	C	Naypyitaw	Normal	Female

	Product line	Unit price	Quantity	Tax 5%
Total ... \				
0	Health and beauty	74.69	7	26.1415
548.9715 ...				
1	Electronic accessories	15.28	5	3.8200
80.2200 ...				

	Time	Payment	cogs	gross margin percentage	gross income \
0	2024-09-27 13:08:00	Ewallet	522.83	4.761905	26.1415
1	2024-09-27 10:29:00	Cash	76.40	4.761905	3.8200

	Rating	Year	Month	Day	month_name
0	9.1	2019	5	1	May
1	9.6	2019	8	3	August

```
[2 rows x 21 columns]
```


Exploratory Data Analysis (EDA)

```
df.head(3)
```

	Invoice ID	Branch	City	Customer type	Gender	\
0	750-67-8428	A	Yangon	Member	Female	
1	226-31-3081	C	Naypyitaw	Normal	Female	
2	631-41-3108	A	Yangon	Normal	Male	

	Product line	Unit price	Quantity	Tax 5%
Total	...			
0	Health and beauty	74.69	7	26.1415
548.9715	...			
1	Electronic accessories	15.28	5	3.8200
80.2200	...			
2	Home and lifestyle	46.33	7	16.2155
340.5255	...			

	Time	Payment	cogs	gross margin percentage	\
0	2024-09-27 13:08:00	Ewallet	522.83	4.761905	
1	2024-09-27 10:29:00	Cash	76.40	4.761905	
2	2024-09-27 13:23:00	Credit card	324.31	4.761905	

	gross income	Rating	Year	Month	Day	month_name
0	26.1415	9.1	2019	5	1	May
1	3.8200	9.6	2019	8	3	August
2	16.2155	7.4	2019	3	3	March

```
[3 rows x 21 columns]
```

Distribution of Rating

```
plt.figure(figsize=(12, 6))
```

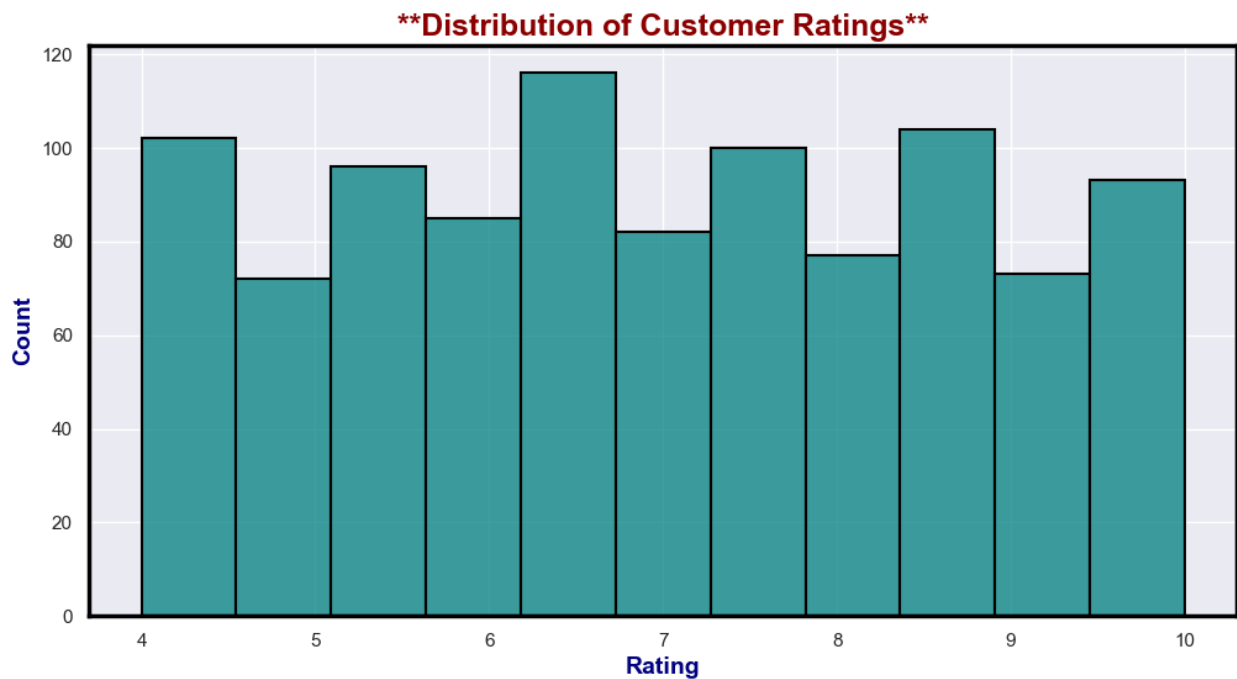
```
sns.histplot(data=df, x='Rating', kde=False, color='teal',  
edgecolor='black', linewidth=1.5)
```

```
plt.title('**Distribution of Customer Ratings**', fontsize=18,  
weight='bold', color='darkred')
```

```
plt.xlabel('Rating', fontsize=14, weight='bold', color='navy')
plt.ylabel('Count', fontsize=14, weight='bold', color='navy')
```

```
plt.gca().spines['top'].set_color('black')
plt.gca().spines['right'].set_color('black')
plt.gca().spines['left'].set_color('black')
plt.gca().spines['bottom'].set_color('black')
plt.gca().spines['top'].set_linewidth(2.5)
plt.gca().spines['right'].set_linewidth(2.5)
plt.gca().spines['left'].set_linewidth(2.5)
plt.gca().spines['bottom'].set_linewidth(2.5)
```

```
plt.show()
```



Distribution of Gross Income

```
plt.figure(figsize=(12, 6))
```

```
sns.histplot(data=df, x='gross income', kde=True, color='purple',
edgecolor='black', linewidth=1.5)
```

```
plt.title('Distribution of gross income', fontsize=18,
fontweight='bold', color='darkblue')

plt.xlabel('Gross Income', fontsize=14, fontweight='bold',
color='darkred')
plt.ylabel('Count', fontsize=14, fontweight='bold', color='darkred')

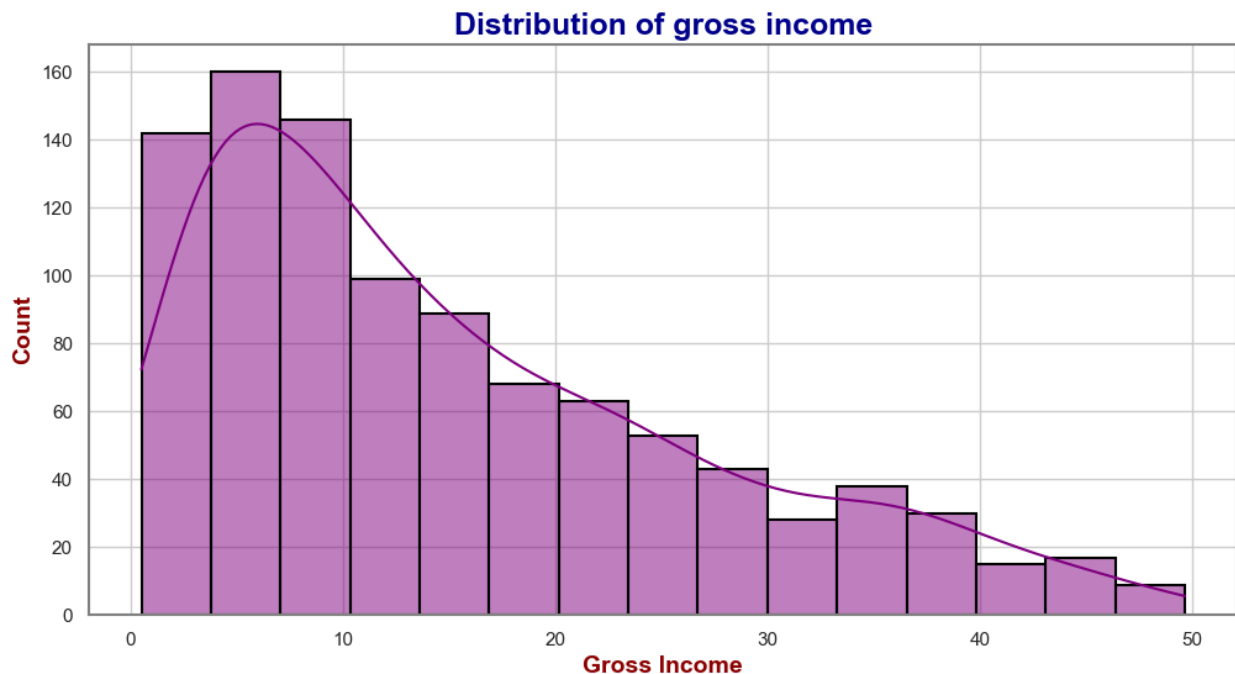
plt.gca().spines['top'].set_visible(True)
plt.gca().spines['top'].set_color('gray')
plt.gca().spines['top'].set_linewidth(1.5)

plt.gca().spines['right'].set_visible(True)
plt.gca().spines['right'].set_color('gray')
plt.gca().spines['right'].set_linewidth(1.5)

plt.gca().spines['left'].set_visible(True)
plt.gca().spines['left'].set_color('gray')
plt.gca().spines['left'].set_linewidth(1.5)

plt.gca().spines['bottom'].set_visible(True)
plt.gca().spines['bottom'].set_color('gray')
plt.gca().spines['bottom'].set_linewidth(1.5)

plt.show()
```



Correlation Heatmap

```
df.corr()
```

```
C:\Users\Vipin\AppData\Local\Temp\ipykernel_15636\1134722465.py:1:
FutureWarning: The default value of numeric_only in DataFrame.corr is
deprecated. In a future version, it will default to False. Select only
valid columns or specify the value of numeric_only to silence this
warning.
```

```
df.corr()
```

	Unit price	Quantity	Tax 5%	Total
cogs \				
Unit price	1.000000	0.010778	0.633962	0.633962
0.633962				
Quantity	0.010778	1.000000	0.705510	0.705510
0.705510				
Tax 5%	0.633962	0.705510	1.000000	1.000000
1.000000				
Total	0.633962	0.705510	1.000000	1.000000
1.000000				
cogs	0.633962	0.705510	1.000000	1.000000
1.000000				
gross margin percentage	NaN	NaN	NaN	NaN
NaN				
gross income	0.633962	0.705510	1.000000	1.000000
1.000000				
Rating	-0.008778	-0.015815	-0.036442	-0.036442
0.036442				
Year	NaN	NaN	NaN	NaN
NaN				
Month	-0.026155	0.036188	0.028576	0.028576
0.028576				
Day	0.053549	-0.048506	-0.012581	-0.012581
0.012581				

	gross margin percentage	gross income
Rating \		
Unit price	NaN	0.633962
0.008778		
Quantity	NaN	0.705510
0.015815		
Tax 5%	NaN	1.000000
0.036442		
Total	NaN	1.000000
0.036442		
cogs	NaN	1.000000
0.036442		
gross margin percentage	NaN	NaN
NaN		

gross income	NaN	1.000000 -
0.036442		
Rating	NaN	-0.036442
1.000000		
Year	NaN	NaN
NaN		
Month	NaN	0.028576
0.014373		
Day	NaN	-0.012581 -
0.013752		

	Year	Month	Day
Unit price	NaN	-0.026155	0.053549
Quantity	NaN	0.036188	-0.048506
Tax 5%	NaN	0.028576	-0.012581
Total	NaN	0.028576	-0.012581
cogs	NaN	0.028576	-0.012581
gross margin percentage	NaN	NaN	NaN
gross income	NaN	0.028576	-0.012581
Rating	NaN	0.014373	-0.013752
Year	NaN	NaN	NaN
Month	NaN	1.000000	-0.656430
Day	NaN	-0.656430	1.000000

```
plt.figure(figsize=(12, 8))
```

```
sns.heatmap(df.corr(), annot=True, cmap='viridis', linewidths=1.5,
linecolor='black', cbar=True,
             cbar_kws={"shrink": .8, "orientation": "vertical"},
square=True)
```

```
plt.title('Correlation Heatmap', fontsize=18, fontweight='bold',
color='darkblue', pad=20)
```

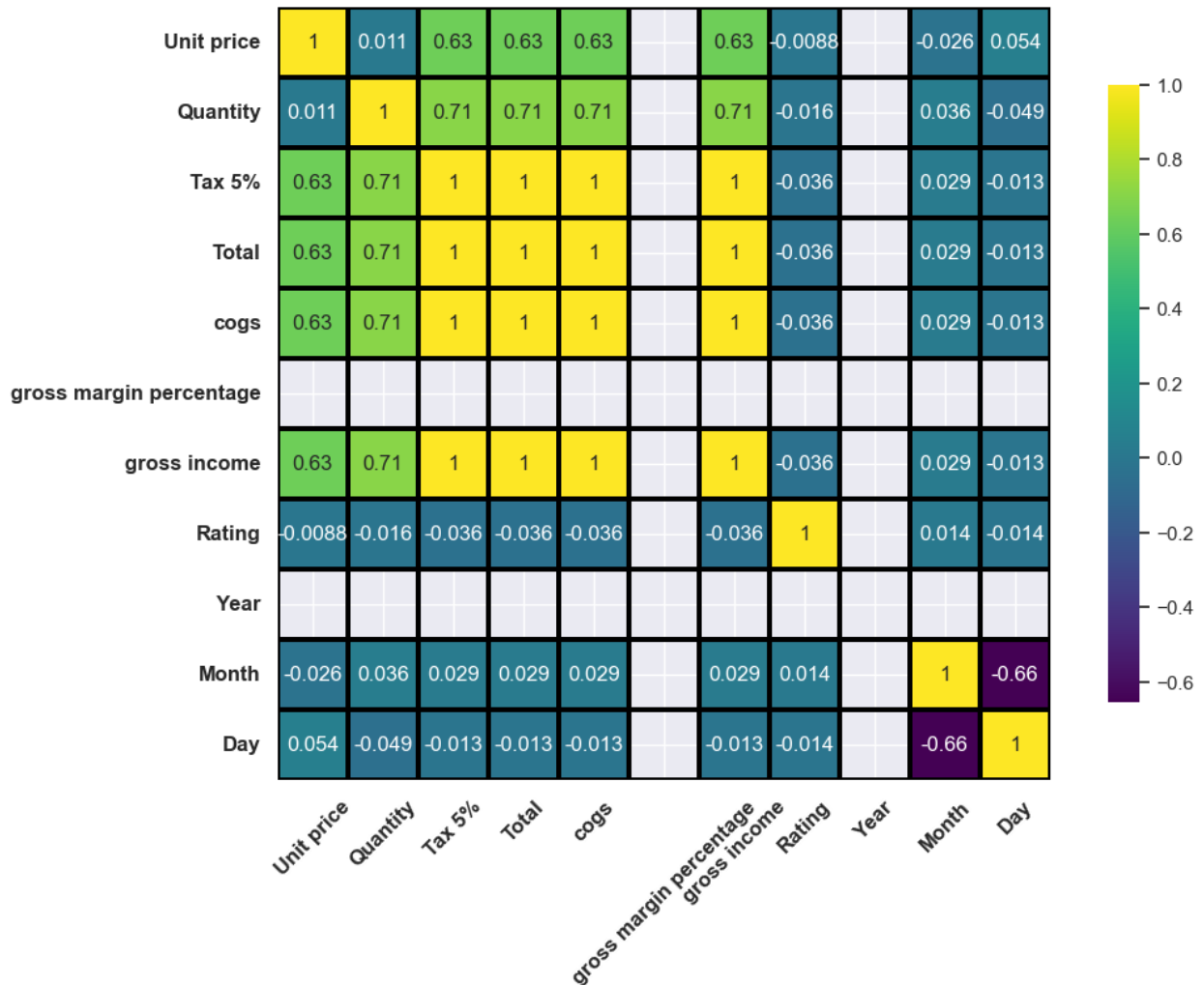
```
plt.xticks(fontsize=12, rotation=45, fontweight='bold')
plt.yticks(fontsize=12, rotation=0, fontweight='bold')
```

```
plt.show()
```

```
C:\Users\Vipin\AppData\Local\Temp\ipykernel_15636\881416650.py:4:
FutureWarning: The default value of numeric_only in DataFrame.corr is
deprecated. In a future version, it will default to False. Select only
valid columns or specify the value of numeric_only to silence this
warning.
```

```
sns.heatmap(df.corr(), annot=True, cmap='viridis', linewidths=1.5,
linecolor='black', cbar=True,
```

Correlation Heatmap



Payment by Gender

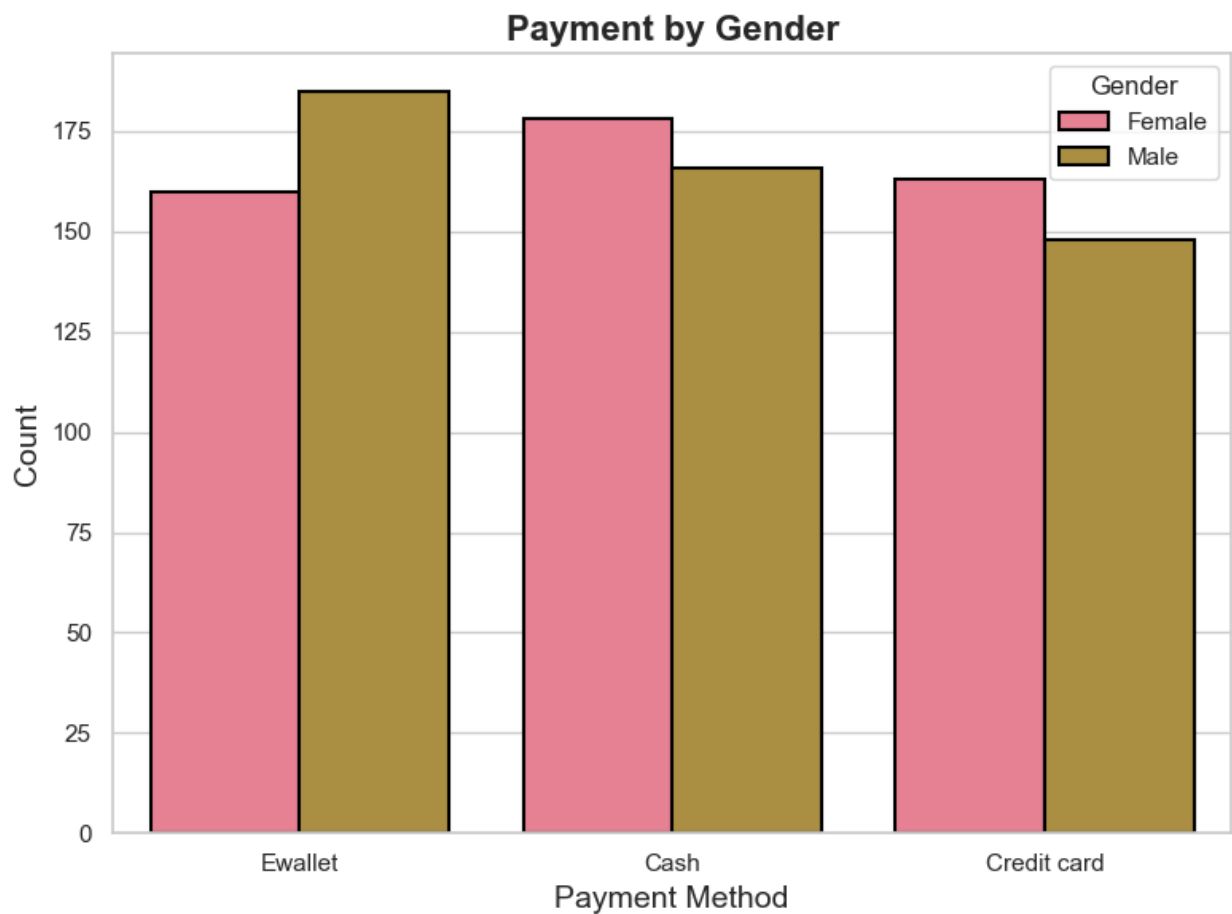
```
plt.figure(figsize=(8, 6))
sns.set_palette("husl")
ax = sns.countplot(x='Payment', data=df, hue='Gender',
edgecolor='black')

for patch in ax.patches:
    patch.set_edgecolor('black')
    patch.set_linewidth(1.5)
```

```
plt.title('Payment by Gender', fontsize=16, fontweight='bold')
plt.xlabel('Payment Method', fontsize=14)
plt.ylabel('Count', fontsize=14)

plt.legend(title='Gender', loc='upper right')

plt.tight_layout()
plt.show()
```



Gender by Customer

```
df['Gender'].value_counts()
```

```
Female    501
Male      499
Name: Gender, dtype: int64

textprops = {'fontsize': 12, 'color': 'g', 'weight': 'bold'}

plt.figure(figsize=(6, 5))

a = df['Gender'].value_counts()

wedgeprops = {'width': 0.6, 'edgecolor': 'black', 'linewidth': 2}

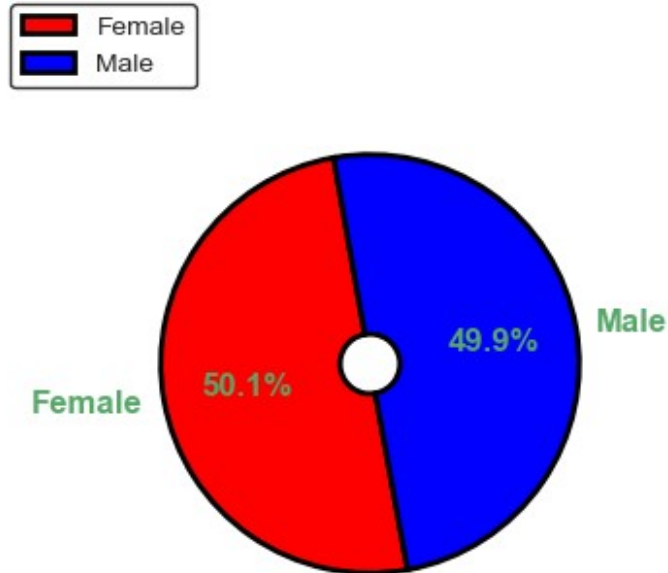
plt.pie(a, labels=a.index,
        colors=['red', 'blue'], startangle=100,
        autopct='%0.1f%%',
        radius=0.7,
        wedgeprops=wedgeprops,
        textprops=textprops)

plt.title('Gender by Customer', fontsize=14, weight='bold',
color='purple')

plt.legend(loc='upper left', edgecolor='black', fontsize=10)

plt.show()
```


Gender by Customer



Sales by Customer Type

```
textprops = {'fontsize': 12, 'color': 'k', 'weight': 'bold'}

plt.figure(figsize=(6, 5))

sales_Customer_type=df.groupby('Customer type')['Total'].sum()

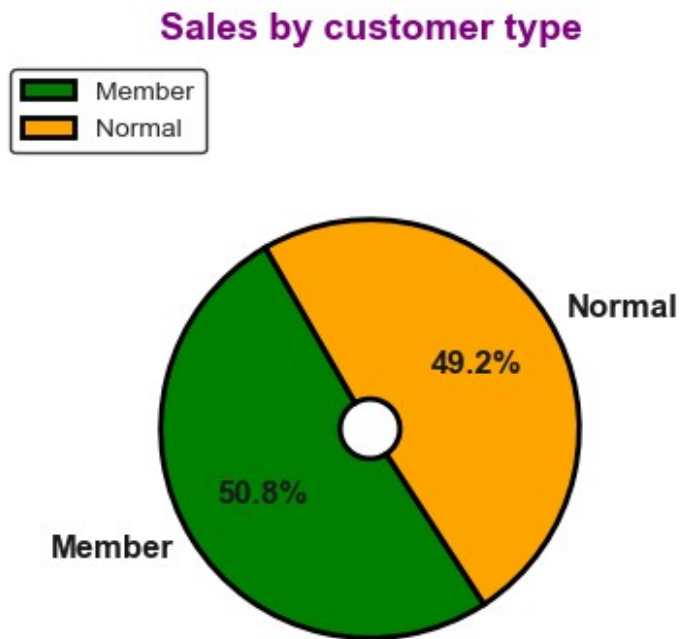
wedgeprops = {'width': 0.6, 'edgecolor': 'black', 'linewidth': 2}

plt.pie(sales_Customer_type, labels=sales_Customer_type.index,
        colors=['green', 'orange'], startangle=120,
        autopct='%0.1f%%',
        radius=0.7,
        wedgeprops=wedgeprops,
        textprops=textprops)
```

```
plt.title('Sales by customer type', fontsize=14, weight='bold',
color='purple')
```

```
plt.legend(loc='upper left',edgecolor='black', fontsize=10)
```

```
plt.show()
```



```
plt.figure(figsize=(8, 6))
```

```
textprops = {'fontsize': 12, 'color': 'k', 'weight': 'bold'}
titleprops = {'fontsize': 16, 'weight': 'bold', 'color': 'darkblue'}
```

```
a = df['Payment'].value_counts()
```

```
plt.pie(a,
```

```

labels=a.index,
colors=['green', 'purple', 'yellow', 'orange', 'blue'],
autopct='%0.1f%%',
textprops=textprops,
radius=0.8,
wedgeprops={'width': 0.5, 'edgecolor': 'black', 'linewidth':
2},
startangle=140)

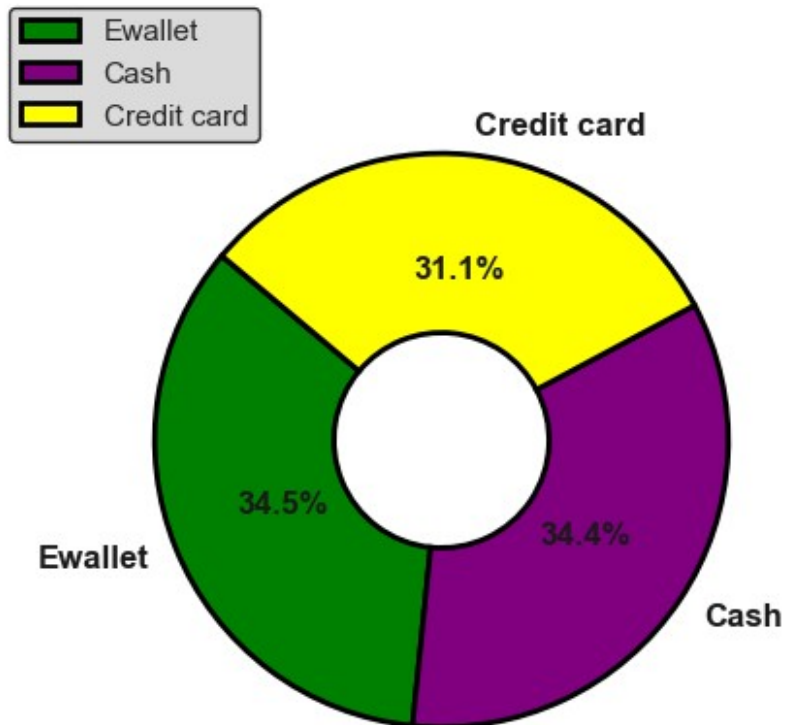
plt.title('Payment Method Distribution ', **titleprops)

plt.legend(loc='upper left', frameon=True, facecolor='lightgrey',
edgecolor='black')

plt.show()

```

Payment Method Distribution



Trend by Sales

```
a = df.groupby(['Date'], as_index=False)['Total'].sum()

a = a.sort_values('Date')

sns.set(style="whitegrid")

plt.figure(figsize=(12, 6))
sns.lineplot(x='Date', y='Total', data=a, marker='o', color='blue')

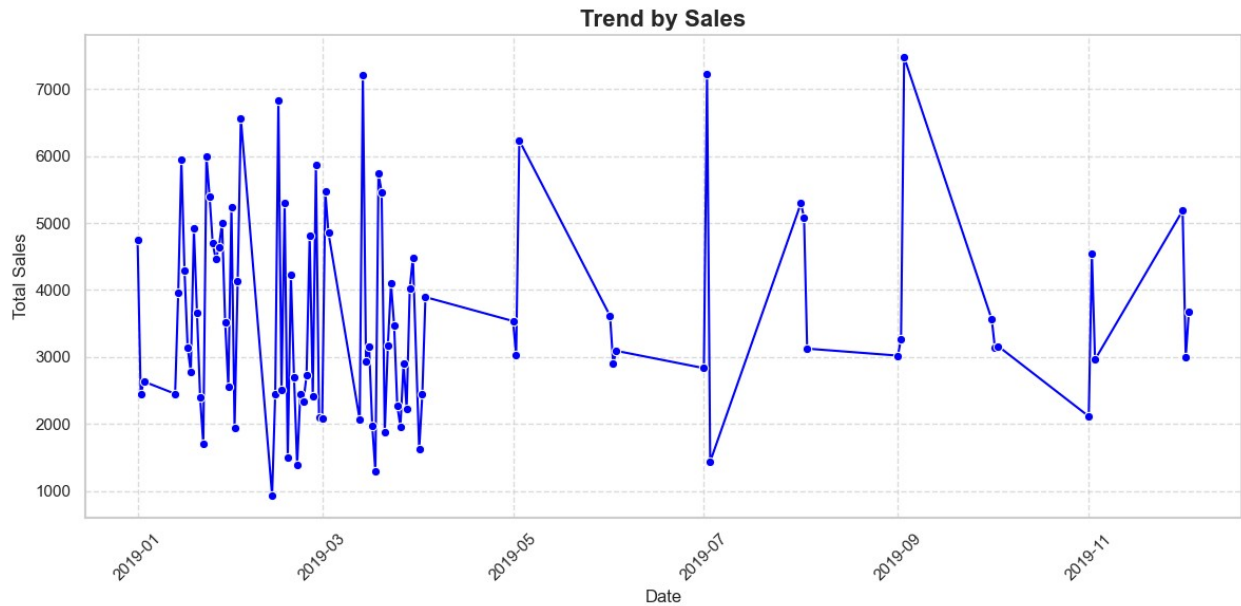
plt.gca().spines['top'].set_visible(True)
plt.gca().spines['right'].set_visible(True)
plt.gca().spines['left'].set_linewidth(1.5)
plt.gca().spines['bottom'].set_linewidth(1.5)

plt.xticks(rotation=45)

plt.title("Trend by Sales", fontsize=16, fontweight='bold')
plt.xlabel("Date", fontsize=12)
plt.ylabel("Total Sales", fontsize=12)

plt.grid(True, linestyle='--', alpha=0.7)

plt.tight_layout()
plt.show()
```



Sales by Branch

```
sales_branch = df.groupby(['Branch'], as_index=False)['Total'].sum()

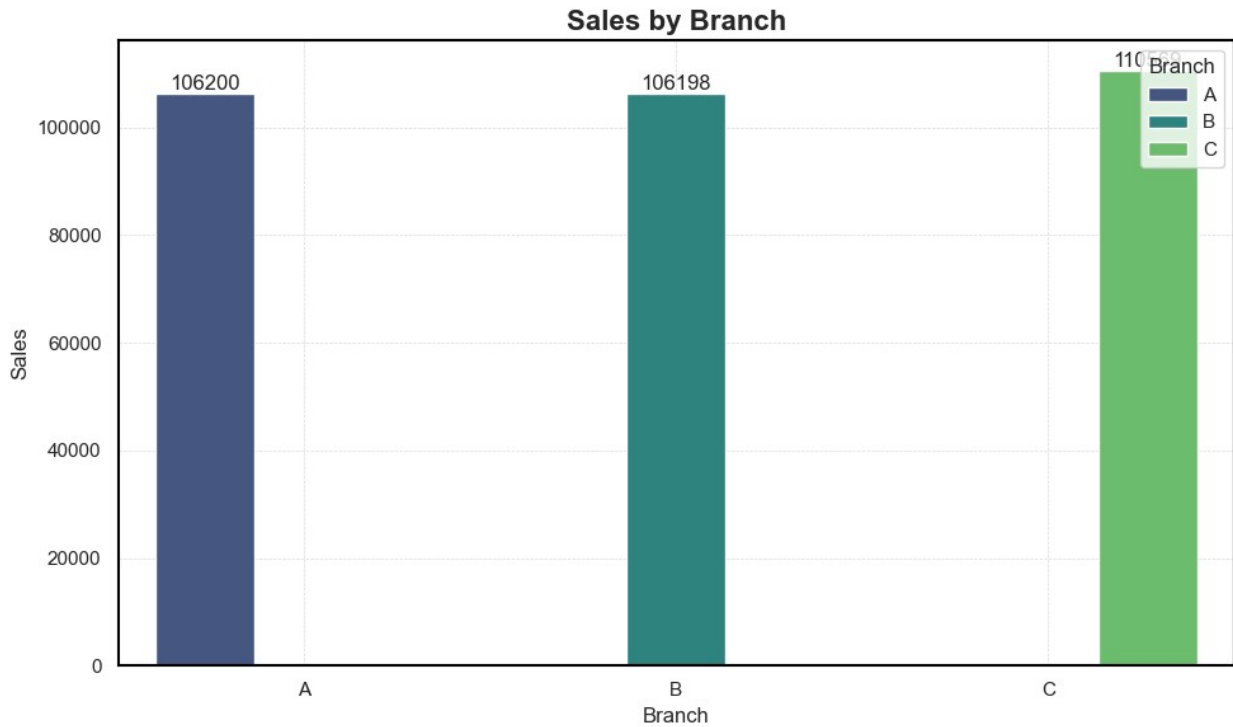
plt.figure(figsize=(10, 6))
ax = sns.barplot(x='Branch', y='Total', hue='Branch',
data=sales_branch, palette='viridis')

plt.ylabel('Sales', fontsize=12)
plt.title('Sales by Branch', fontsize=16, fontweight='bold')
plt.grid(True, linestyle='--', linewidth=0.5, alpha=0.7)

for spine in ax.spines.values():
    spine.set_edgecolor('black')
    spine.set_linewidth(1.5)

for bar in ax.containers:
    ax.bar_label(bar)

plt.tight_layout()
plt.show()
```



Sales by City

```
df['City'].unique()
array(['Yangon', 'Naypyitaw', 'Mandalay'], dtype=object)

city_by_total=df.groupby(['City'], as_index=False)
['Total'].sum().sort_values(by='Total', ascending=False)
city_by_total
```

	City	Total
1	Naypyitaw	110568.7065
2	Yangon	106200.3705
0	Mandalay	106197.6720

```
city_by_total = df.groupby(['City'], as_index=False)
['Total'].sum().sort_values(by='Total', ascending=False)

colors = ['orange']

ax = city_by_total.plot(kind='bar', x='City', color=colors,
figsize=(8, 6))
```

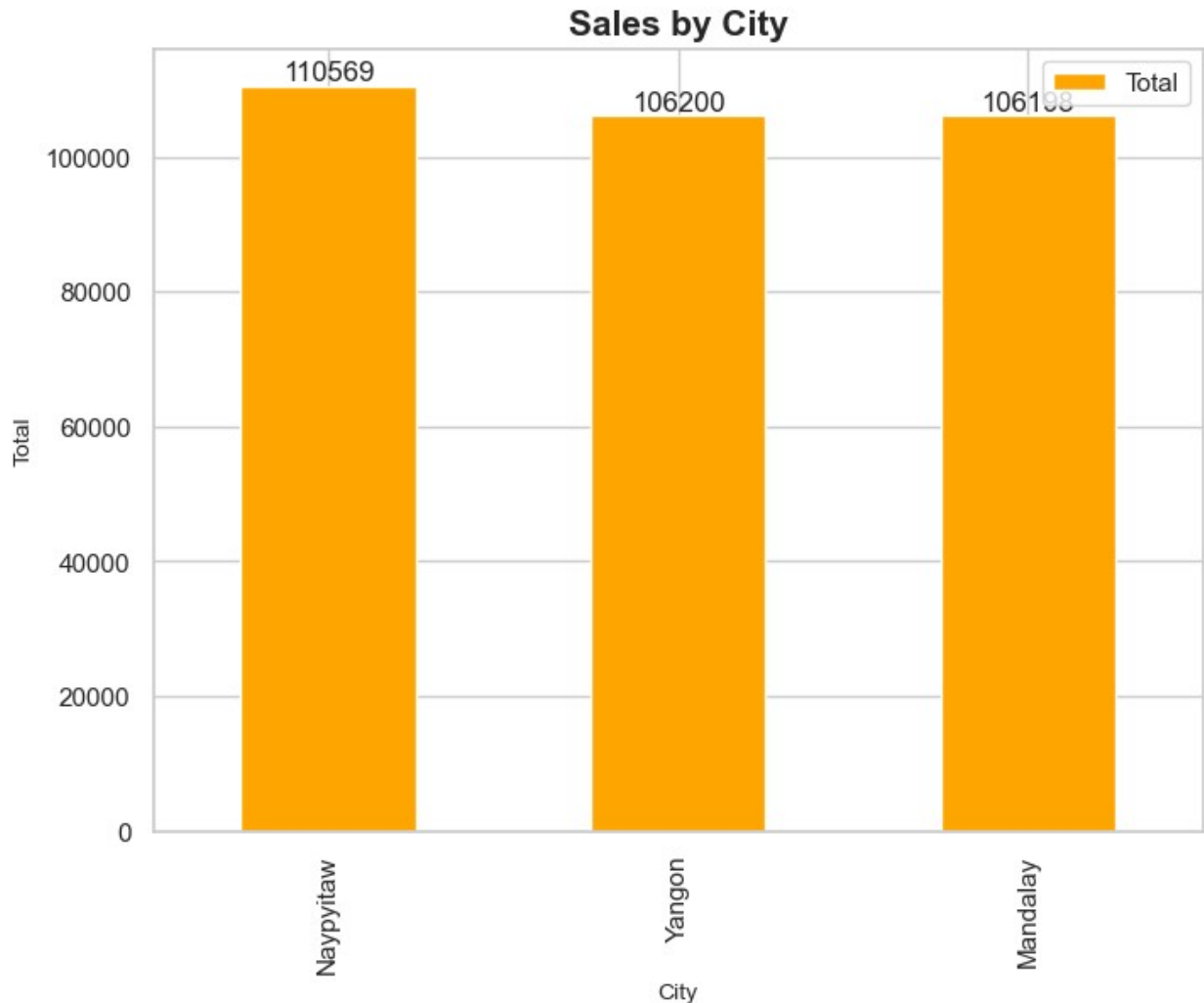
```

for bars in ax.containers:
    ax.bar_label(bars)

ax.set_xlabel('City', fontsize=10)
ax.set_ylabel('Total', fontsize=10)
ax.set_title('Sales by City', fontsize=15, fontweight='bold')

plt.show()

```



Average Product Rating by Product

```

avg_product_rating = df.groupby(['Product line'], as_index=False)

```

```

['Rating'].mean().sort_values(by='Rating', ascending=False)

colors = [ 'salmon', 'violet', 'peachpuff', 'lightcoral'] # Added
more colors

ax = avg_product_rating.plot(kind='barh', x='Product line',
color=colors, figsize=(10, 6), edgecolor='black')

for bars in ax.containers:
    ax.bar_label(bars, fmt='%.1f')

ax.set_xlabel('Average Rating', fontsize=12)
ax.set_ylabel('Product Line', fontsize=12)
ax.set_title('Average Product Rating by Product Line', fontsize=16,
fontweight='bold', color='navy')

ax.grid(axis='x', linestyle='--', alpha=0.7)

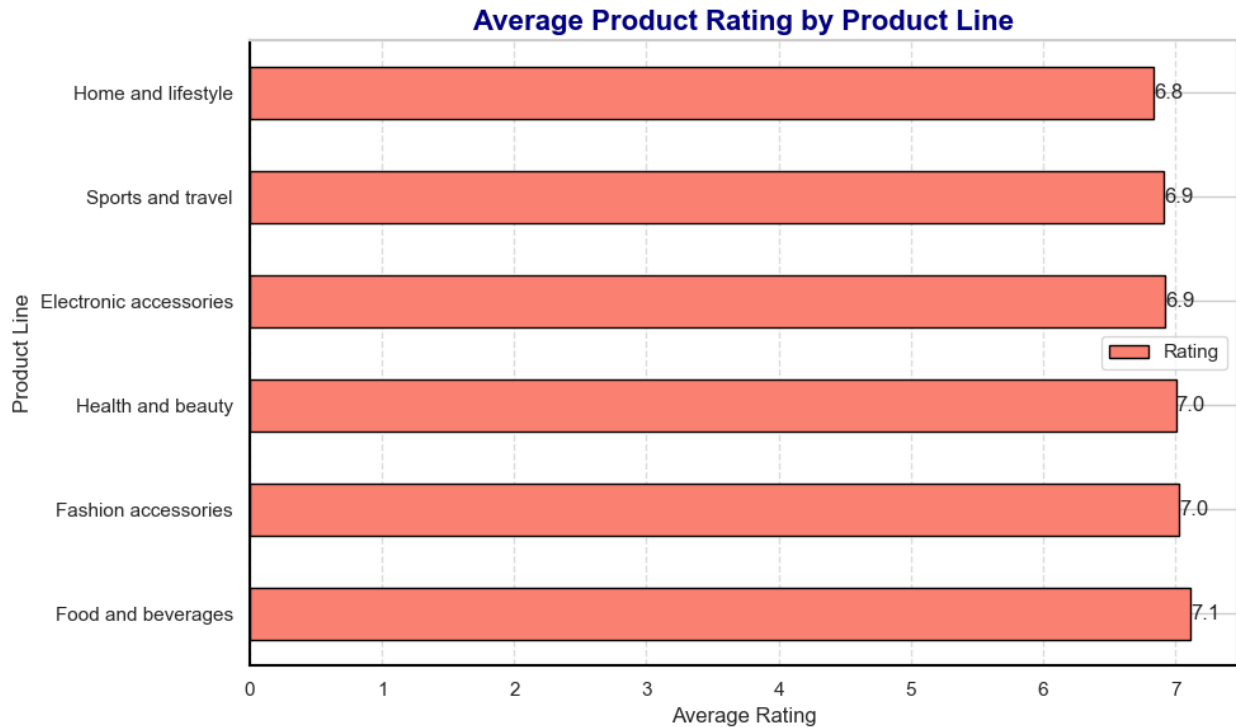
plt.gca().spines['top'].set_linewidth(1.5)
plt.gca().spines['right'].set_linewidth(1.5)
plt.gca().spines['left'].set_linewidth(1.5)
plt.gca().spines['bottom'].set_linewidth(1.5)

plt.gca().spines['left'].set_color('black')
plt.gca().spines['bottom'].set_color('black')

plt.tight_layout()

plt.show()

```

Total Rating vs Product

```
productline_by_rating = df.groupby(['Product line'], as_index=False)
['Rating'].sum().sort_values(by='Rating', ascending=False)
productline_by_rating
```

	Product line	Rating
1	Fashion accessories	1251.2
2	Food and beverages	1237.7
0	Electronic accessories	1177.2
5	Sports and travel	1148.1
4	Home and lifestyle	1094.0
3	Health and beauty	1064.5

```
sns.set(rc={'figure.figsize': (15, 5)})
```

```
a = df.groupby(['Product line'], as_index=False)
['Rating'].sum().sort_values(by='Rating', ascending=False)
```

```
ax = sns.barplot(x='Product line', y='Rating', data=a,
```

```

palette='viridis', edgecolor='black')

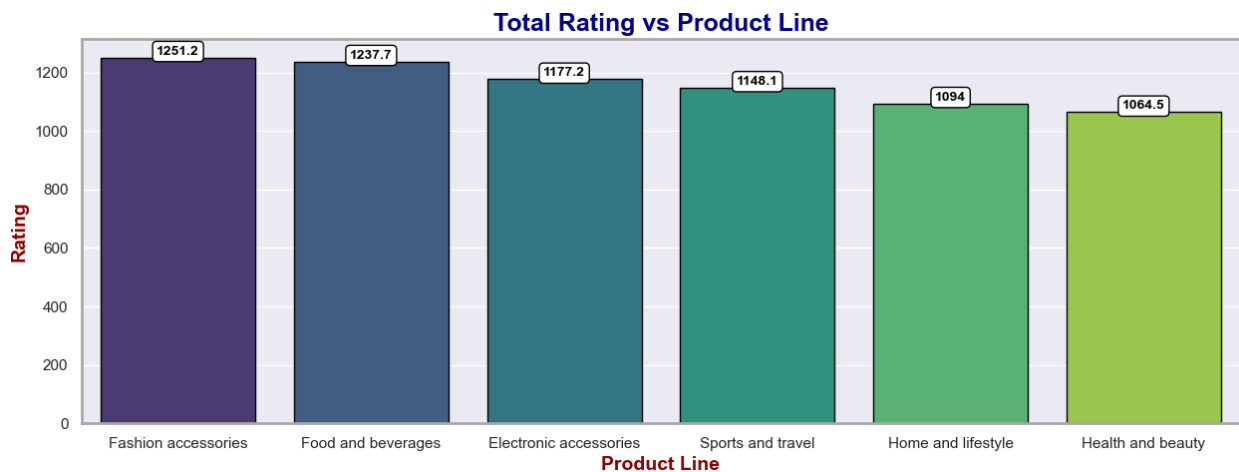
for i in ax.containers:
    ax.bar_label(i, label_type='edge', fontsize=10, color='black',
weight='bold', bbox=dict(facecolor='white', edgecolor='black',
boxstyle='round,pad=0.3'))

plt.title('Total Rating vs Product Line', fontsize=18,
fontweight='bold', color='darkblue')
plt.xlabel('Product Line', fontsize=14, fontweight='bold',
color='darkred')
plt.ylabel('Rating', fontsize=14, fontweight='bold', color='darkred')

plt.gca().spines['top'].set_linewidth(2)
plt.gca().spines['bottom'].set_linewidth(2)
plt.gca().spines['left'].set_linewidth(2)
plt.gca().spines['right'].set_linewidth(2)
plt.gca().spines['top'].set_color('darkgray')
plt.gca().spines['bottom'].set_color('darkgray')
plt.gca().spines['left'].set_color('darkgray')
plt.gca().spines['right'].set_color('darkgray')

plt.show()

```



Total Gross Income by Product

```
gross_income_by_products= df.groupby(['Product line'], as_index=False)
['gross income'].sum().sort_values(by='gross income', ascending=False)
gross_income_by_products
```

	Product line	gross income
2	Food and beverages	2673.5640
5	Sports and travel	2624.8965
0	Electronic accessories	2587.5015
1	Fashion accessories	2585.9950
4	Home and lifestyle	2564.8530
3	Health and beauty	2342.5590

```
a = df.groupby(['Product line'], as_index=False)['gross
income'].sum().sort_values(by='gross income', ascending=False)
```

```
plt.figure(figsize=(12, 7))
```

```
ax = a.plot(kind='bar', color=['#5DADE2', '#48C9B0', '#F4D03F',
'#E74C3C', '#AF7AC5', '#F39C12'],
x='Product line', width=0.6)
```

```
for container in ax.containers:
    ax.bar_label(container, label_type='center', fontsize=12)
```

```
plt.title('Total Gross Income by Product Line', fontsize=18,
fontweight='bold')
```

```
plt.xlabel('Product Line', fontsize=14)
```

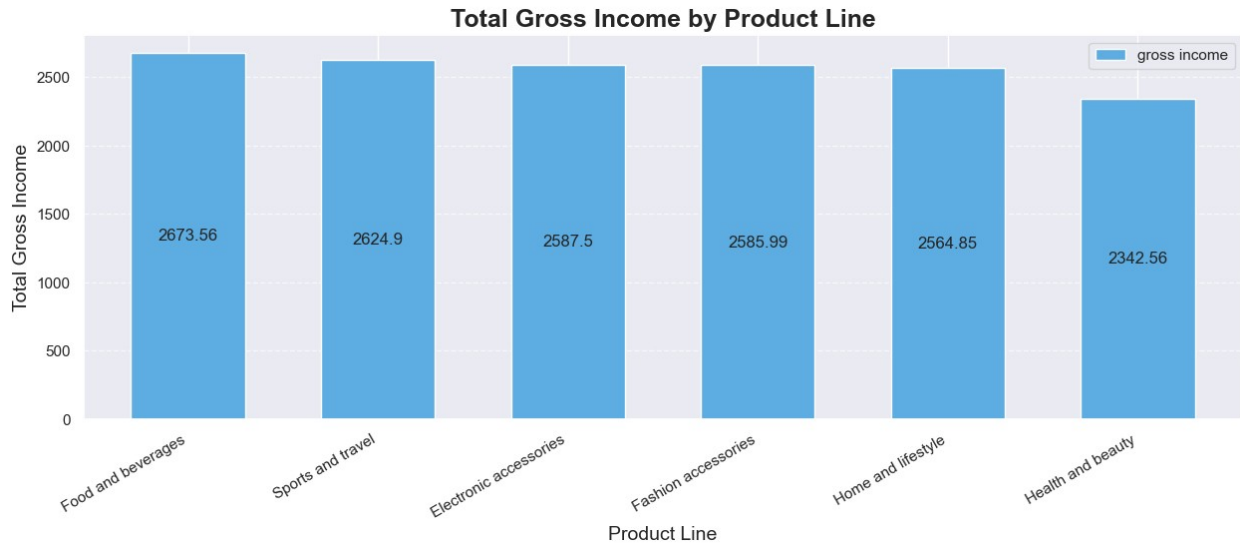
```
plt.ylabel('Total Gross Income', fontsize=14)
```

```
plt.xticks(rotation=30, ha='right')
```

```
plt.grid(axis='y', linestyle='--', alpha=0.7)
```

```
plt.show()
```

```
<Figure size 1200x700 with 0 Axes>
```



Trend by Month

```
month_order = ['January', 'February', 'March', 'April', 'May', 'June',
               'July', 'August', 'September', 'October', 'November',
               'December']
```

```
df['month_name'] = pd.Categorical(df['month_name'],
                                categories=month_order, ordered=True)
```

```
a = df.groupby(['month_name'], as_index=False)
    ['Total'].sum().sort_values(by='month_name')
```

```
a['Total'] = a['Total'].round(0).astype(int)
```

```
print(a)
```

	month_name	Total
0	January	86563
1	February	63170
2	March	72749
3	April	7958
4	May	12799
5	June	9612
6	July	11501

7	August	13504
8	September	13767
9	October	9865
10	November	9618
11	December	11861

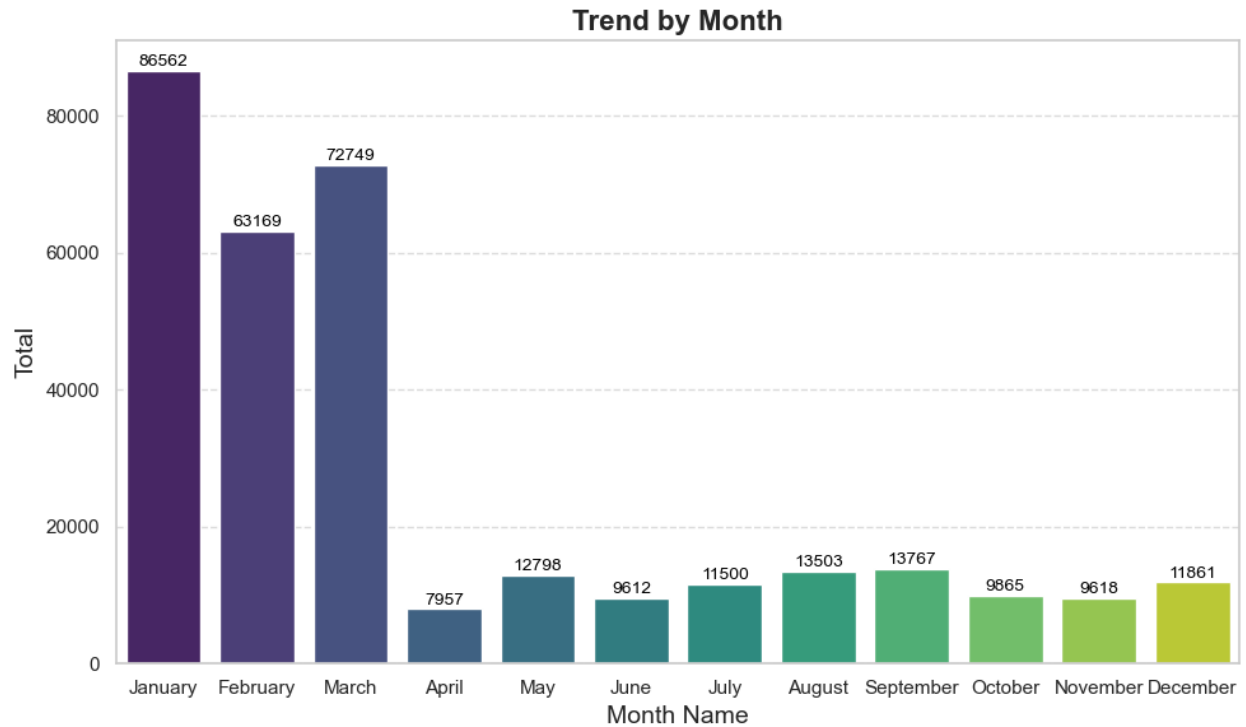
```
a = df.groupby(['month_name'], as_index=False)
['Total'].sum().sort_values(by='Total', ascending=False)
```

```
plt.figure(figsize=(10, 6))
bar_plot = sns.barplot(x='month_name', y='Total', data=a,
palette='viridis')
```

```
plt.title('Trend by Month', fontsize=16, fontweight='bold')
plt.xlabel('Month Name', fontsize=14)
plt.ylabel('Total', fontsize=14)
plt.grid(axis='y', linestyle='--', alpha=0.7)
```

```
for patch in bar_plot.patches:
    bar_plot.annotate(f'{int(patch.get_height())}', # Change here to
use int
                      (patch.get_x() + patch.get_width() / 2.,
                       patch.get_height()),
                      ha='center', va='center',
                      fontsize=10, color='black',
                      xytext=(0, 5),
                      textcoords='offset points')
```

```
plt.tight_layout()
plt.show()
```



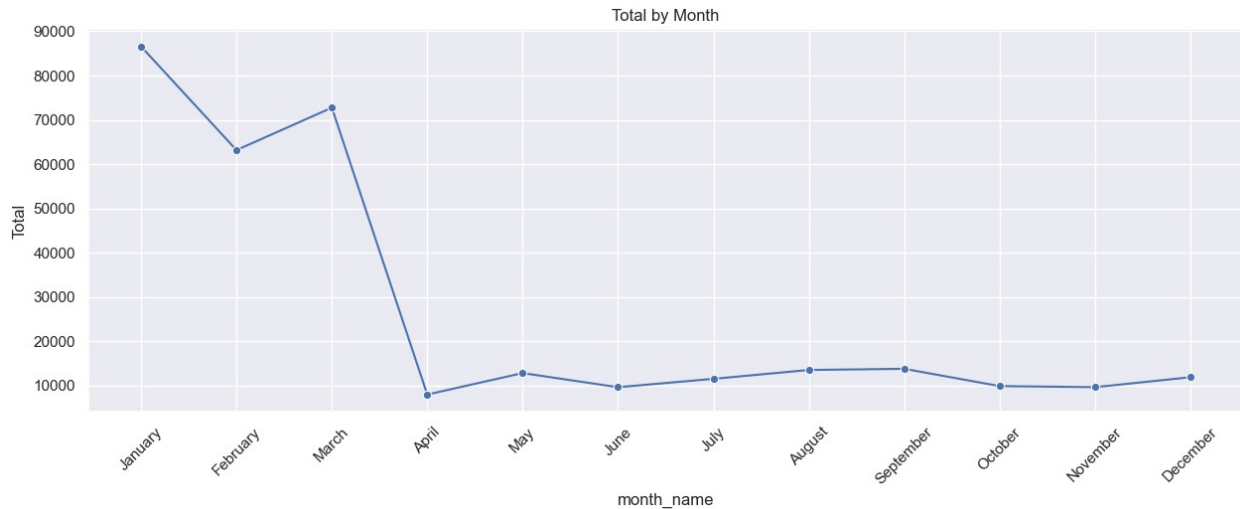
Trend by Month Using Line Graph

```
a = df.groupby(['month_name'], as_index=False)['Total'].sum()

month_order = ["January", "February", "March", "April", "May", "June",
               "July", "August", "September", "October", "November",
               "December"]

a['month_name'] = pd.Categorical(a['month_name'],
                                categories=month_order, ordered=True)
a = a.sort_values('month_name')

sns.lineplot(x='month_name', y='Total', data=a, marker='o')
plt.xticks(rotation=45)
plt.title("Trend by by Month")
plt.show()
```



Relationship between Quantity and Sales

```
plt.figure(figsize=(10, 5))

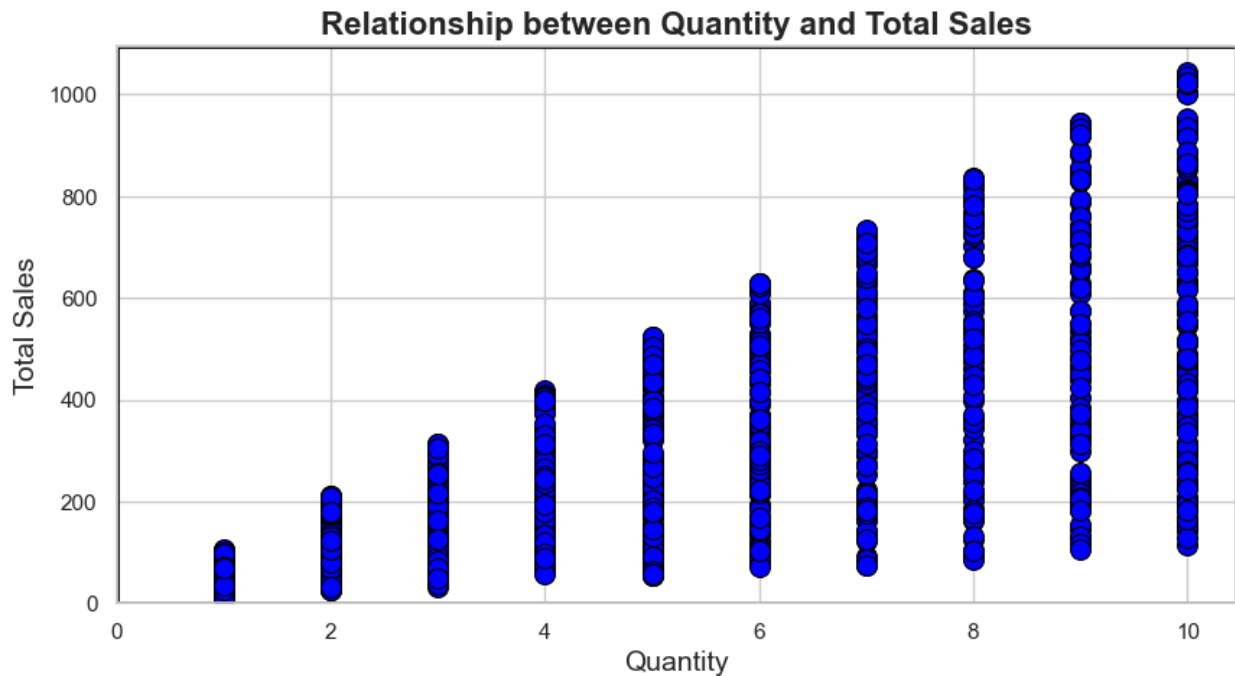
sns.scatterplot(data=df, x='Quantity', y='Total', color='blue',
edgecolor='black', s=100)

plt.title('Relationship between Quantity and Total Sales',
fontsize=16, fontweight='bold')

plt.xlabel('Quantity', fontsize=14)
plt.ylabel('Total Sales', fontsize=14)
plt.xlim(left=0)
plt.ylim(bottom=0)

plt.gca().add_patch(plt.Rectangle((0, 0), plt.xlim()[1], plt.ylim()
[1], fill=False, edgecolor='black', linewidth=2))

plt.show()
```



Sales by Payment Method and Gender

```
plt.figure(figsize=(10, 5))

sns.boxplot(data=df, x='Payment', y='Total', hue='Gender',
            palette='Set2')

plt.title('Total Sales by Payment Method and Gender', fontsize=16,
          fontweight='bold', color='darkblue')
plt.xlabel('Payment Method', fontsize=14)
plt.ylabel('Total Sales', fontsize=14)

plt.gca().spines['top'].set_linewidth(2)
plt.gca().spines['bottom'].set_linewidth(2)
plt.gca().spines['right'].set_linewidth(2)
plt.gca().spines['left'].set_linewidth(2)

plt.gca().spines['top'].set_color('gray')
plt.gca().spines['bottom'].set_color('gray')
plt.gca().spines['right'].set_color('gray')
```



```
plt.gca().spines['left'].set_color('gray')
```

```
plt.legend(title='Gender', fontsize=12)  
plt.grid(True, linestyle='--', alpha=0.7)  
plt.show()
```

