

```
In [1]: import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [2]: df=pd.read_csv('supermarket_sales - Sheet1.csv')
```

```
In [3]: df
```

Out[3]:

	Invoice ID	Branch	City	Customer type	Gender	Product line	Unit price	Quantity	Tax	5
0	750-67-8428	A	Yangon	Member	Female	Health and beauty	74.69	7	26.14	
1	226-31-3081	C	Naypyitaw	Normal	Female	Electronic accessories	15.28	5	3.82	
2	631-41-3108	A	Yangon	Normal	Male	Home and lifestyle	46.33	7	16.21	
3	123-19-1176	A	Yangon	Member	Male	Health and beauty	58.22	8	23.28	
4	373-73-7910	A	Yangon	Normal	Male	Sports and travel	86.31	7	30.20	
...
995	233-67-5758	C	Naypyitaw	Normal	Male	Health and beauty	40.35	1	2.01	
996	303-96-2227	B	Mandalay	Normal	Female	Home and lifestyle	97.38	10	48.69	
997	727-02-1313	A	Yangon	Member	Male	Food and beverages	31.84	1	1.59	
998	347-56-2442	A	Yangon	Normal	Male	Home and lifestyle	65.82	1	3.29	
999	849-09-3807	A	Yangon	Member	Female	Fashion accessories	88.34	7	30.91	

1000 rows × 17 columns



```
In [4]: df.columns
```

```
Out[4]: 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')
```

In [5]: `df.describe()`

Out[5]:

	Unit price	Quantity	Tax 5%	Total	cogs	gross margin percentage	
count	1000.000000	1000.000000	1000.000000	1000.000000	1000.000000	1000.000000	10
mean	55.672130	5.510000	15.379369	322.966749	307.58738	4.761905	
std	26.494628	2.923431	11.708825	245.885335	234.17651	0.000000	
min	10.080000	1.000000	0.508500	10.678500	10.17000	4.761905	
25%	32.875000	3.000000	5.924875	124.422375	118.49750	4.761905	
50%	55.230000	5.000000	12.088000	253.848000	241.76000	4.761905	
75%	77.935000	8.000000	22.445250	471.350250	448.90500	4.761905	
max	99.960000	10.000000	49.650000	1042.650000	993.00000	4.761905	

◀ ▶

In [6]: `df.dtypes`

```
Out[6]: 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
```

In [7]: `# chck for nul values
df.isnull().sum()`

```
Out[7]: 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
```

```
In [8]: #check a duplicate value  
duplicated_rows=df[df.duplicated()]  
print("duplicate rows based on all columns:")  
print(duplicated_rows)
```

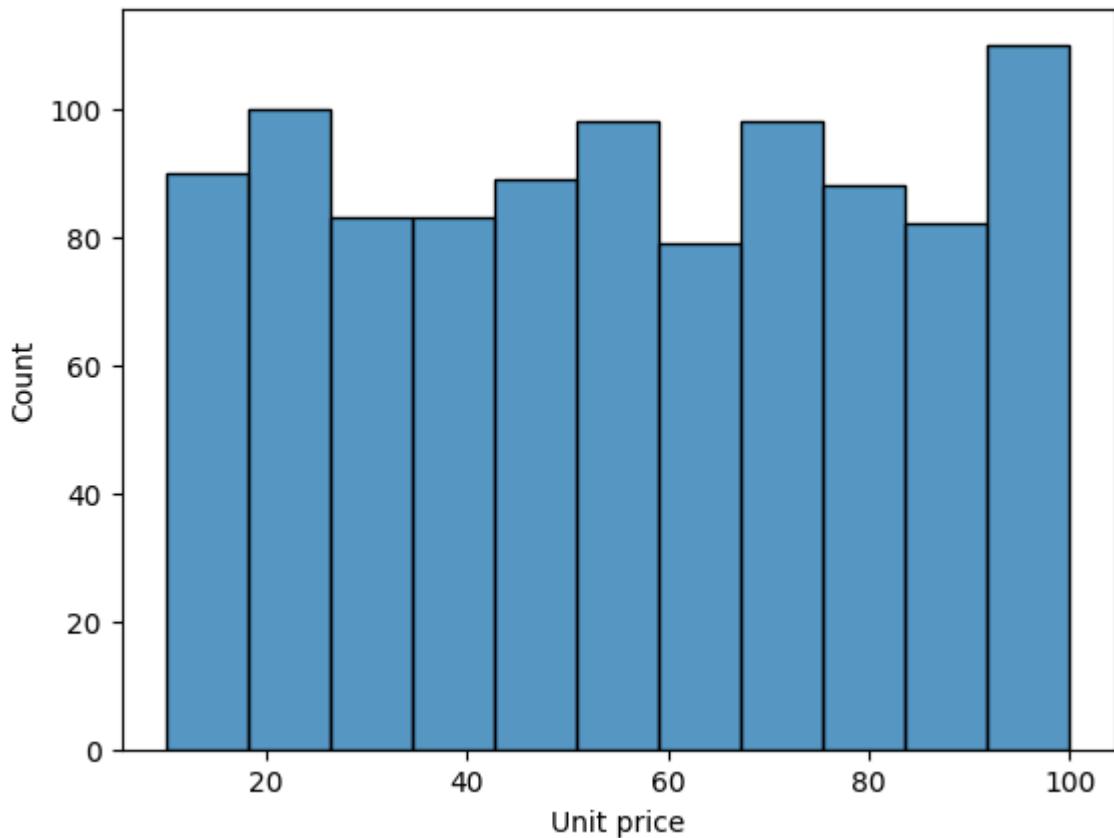
```
duplicate rows based on all columns:  
Empty DataFrame  
Columns: [Invoice ID, Branch, City, Customer type, Gender, Product line, Unit pri  
ce, Quantity, Tax 5%, Total, Date, Time, Payment, cogs, gross margin percentage,  
gross income, Rating]  
Index: []
```

```
In [9]: total_sum_duplicated_valuse=df.duplicated().sum()  
print(total_sum_duplicated_valuse)
```

```
0
```

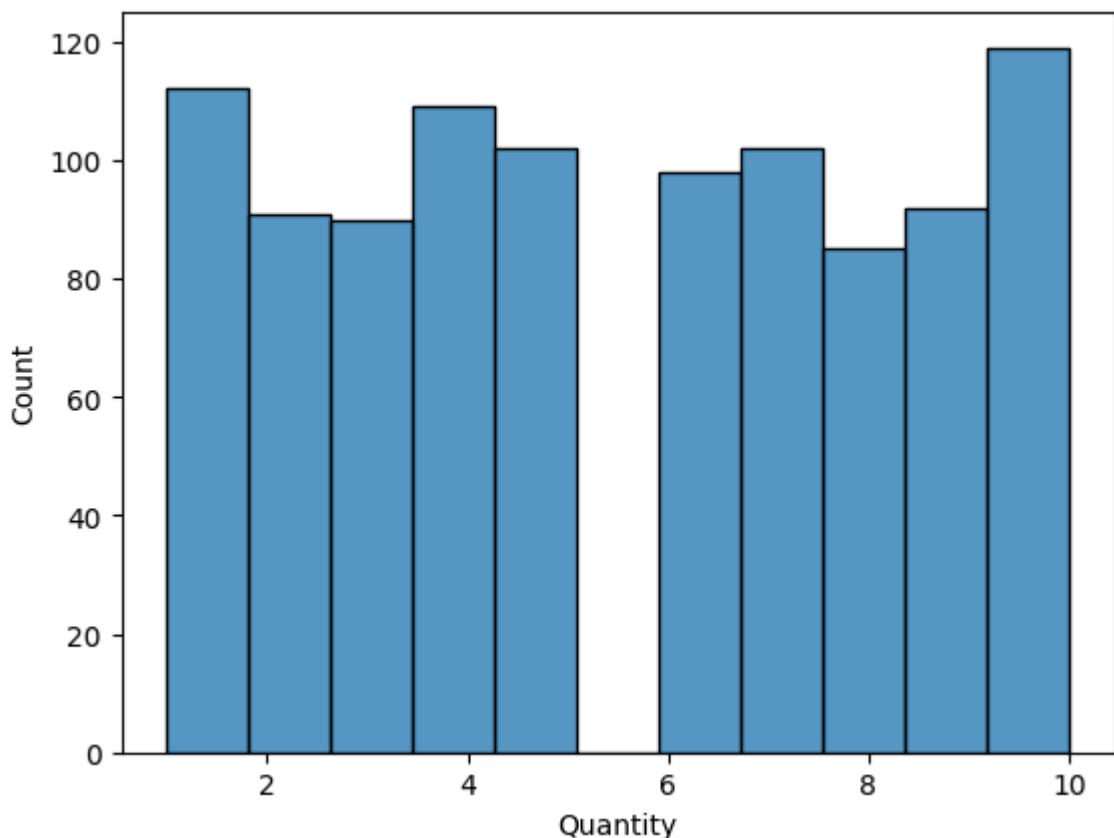
```
In [10]: #univariat analyst  
sns.histplot(x='Unit price',data=df)
```

```
Out[10]: <Axes: xlabel='Unit price', ylabel='Count'>
```



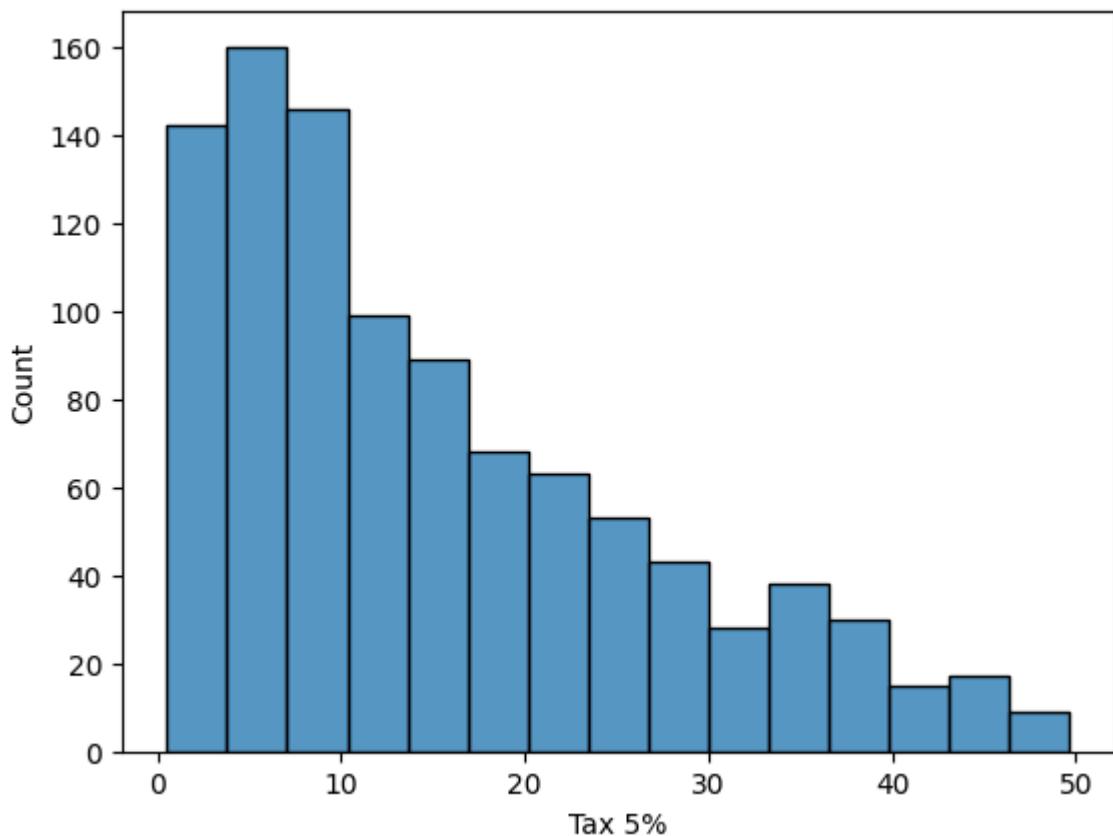
```
In [11]: #Quantity  
sns.histplot(x='Quantity', data=df)
```

```
Out[11]: <Axes: xlabel='Quantity', ylabel='Count'>
```



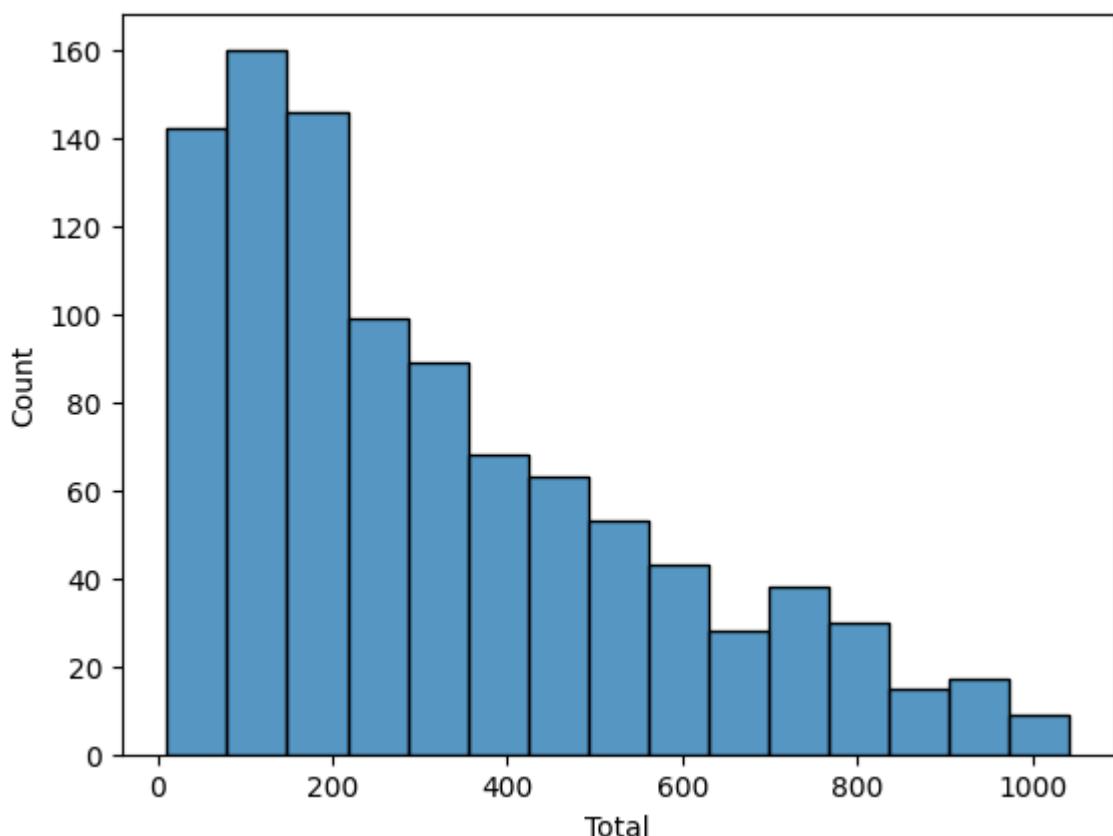
```
In [12]: sns.histplot(x='Tax %', data=df)
```

```
Out[12]: <Axes: xlabel='Tax 5%', ylabel='Count'>
```



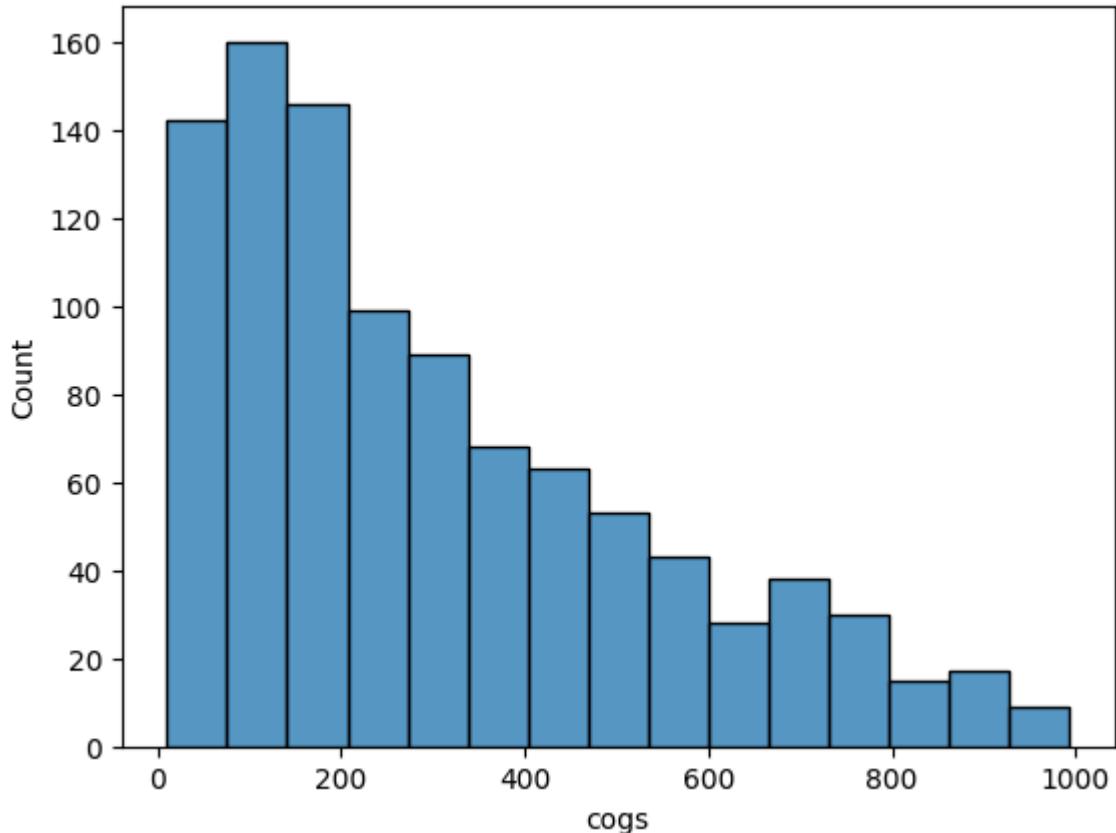
```
In [13]: sns.histplot(x='Total', data=df)
```

```
Out[13]: <Axes: xlabel='Total', ylabel='Count'>
```



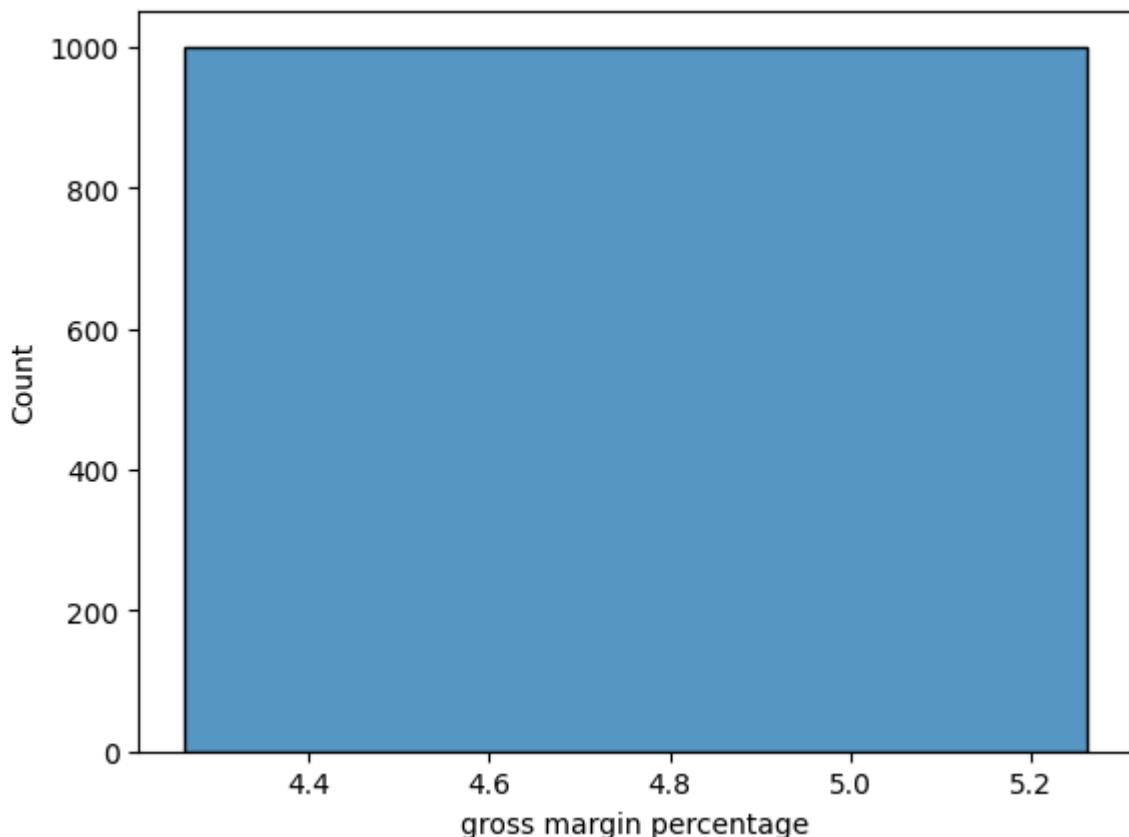
```
In [14]: sns.histplot(x='cogs', data=df)
```

```
Out[14]: <Axes: xlabel='cogs', ylabel='Count'>
```



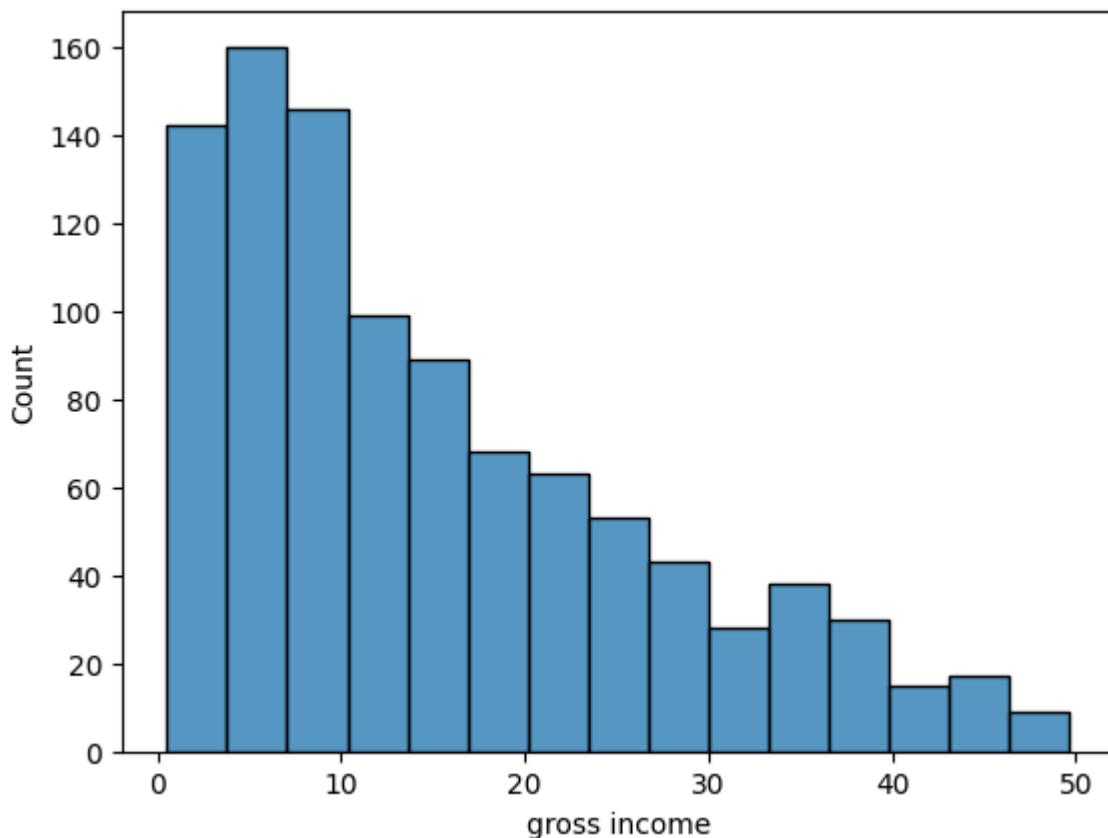
```
In [15]: sns.histplot(x='gross margin percentage', data=df)
```

```
Out[15]: <Axes: xlabel='gross margin percentage', ylabel='Count'>
```



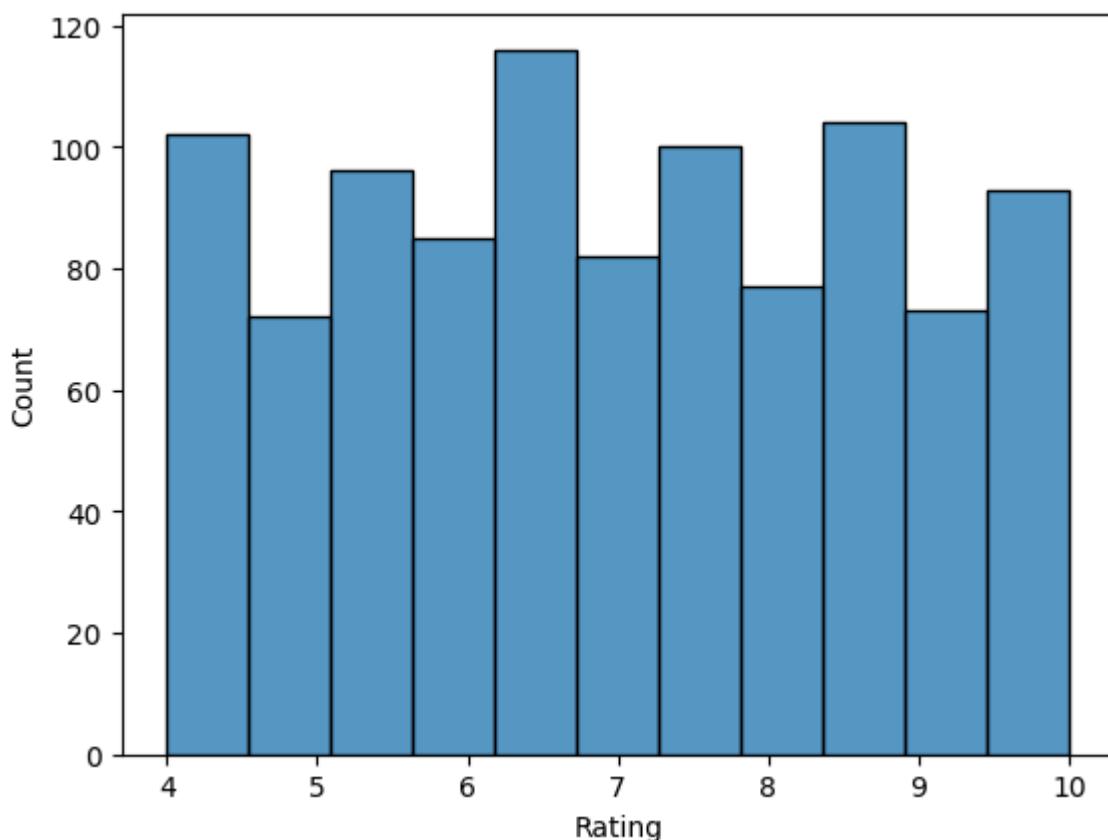
```
In [16]: sns.histplot(x='gross income', data=df)
```

```
Out[16]: <Axes: xlabel='gross income', ylabel='Count'>
```



```
In [17]: sns.histplot(x='Rating', data=df)
```

```
Out[17]: <Axes: xlabel='Rating', ylabel='Count'>
```

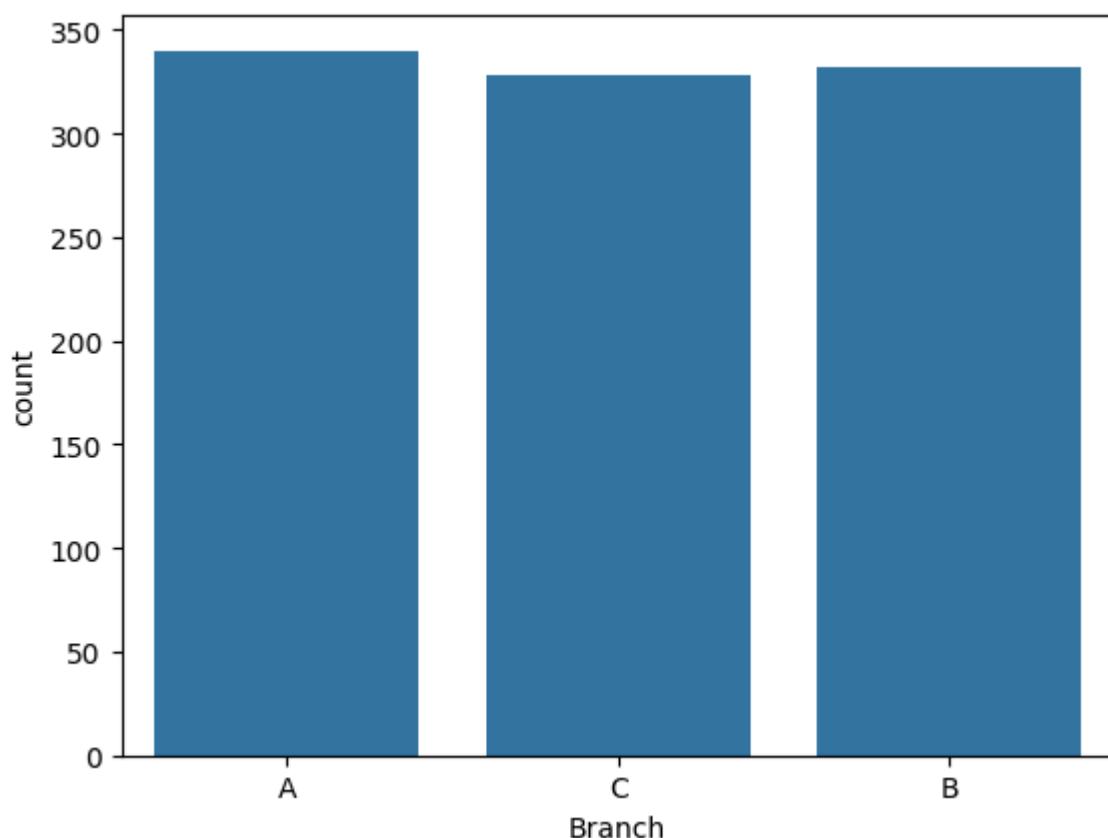


```
In [18]: df.columns
```

```
Out[18]: 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')
```

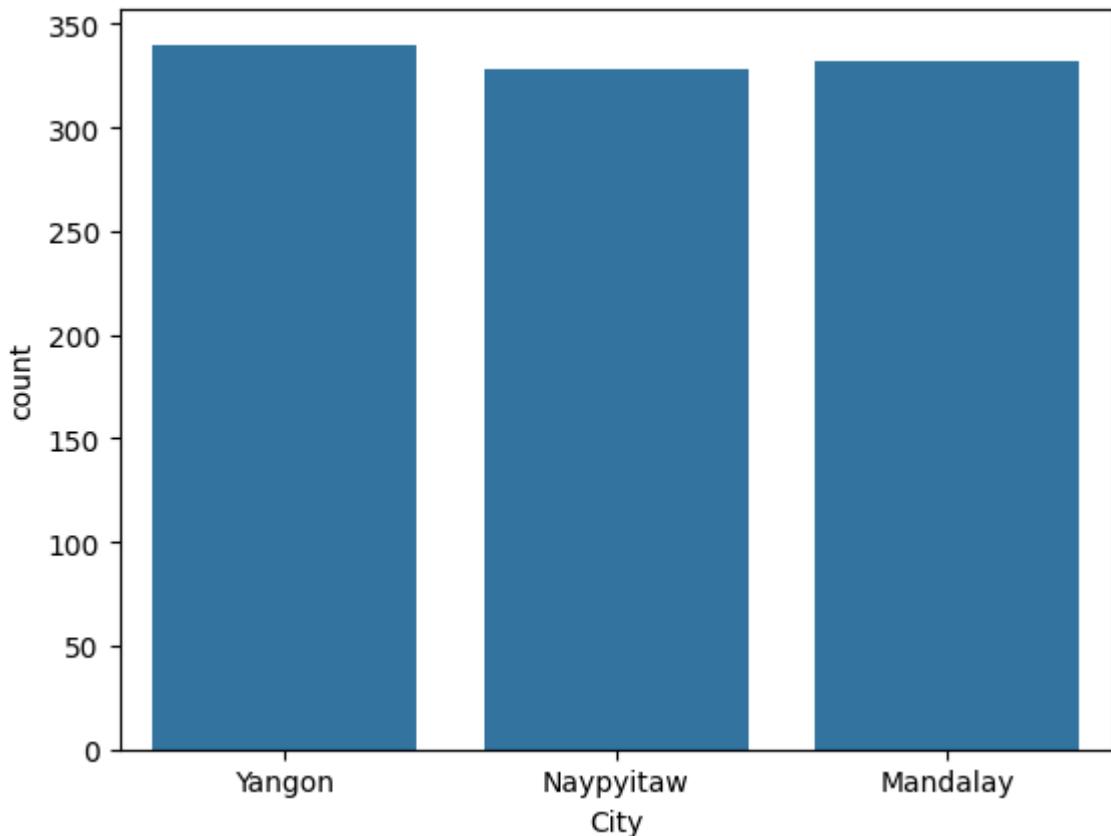
```
In [19]: #count each category
#branch, city,
sns.countplot(x='Branch',data=df)
df['Branch'].value_counts()
```

```
Out[19]: Branch
A    340
B    332
C    328
Name: count, dtype: int64
```



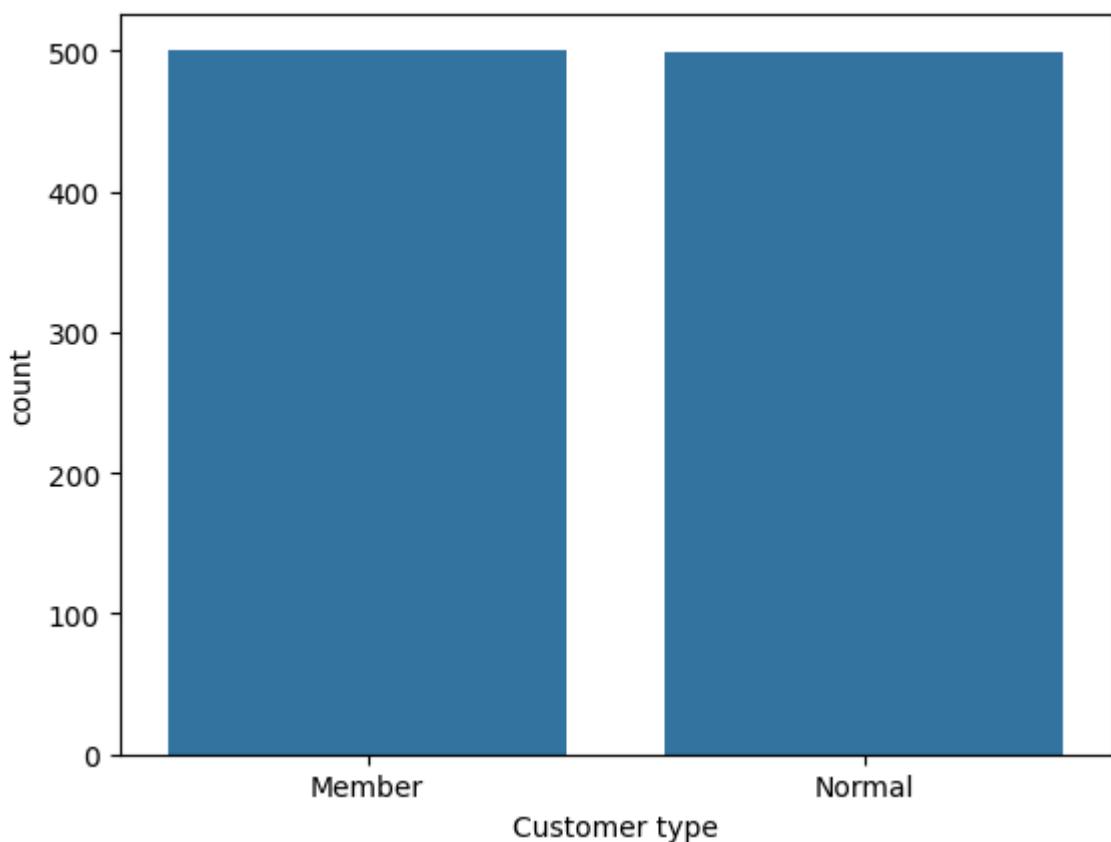
```
In [20]: sns.countplot(x='City',data=df)
df['City'].value_counts()
```

```
Out[20]: City
Yangon     340
Mandalay   332
Naypyitaw  328
Name: count, dtype: int64
```



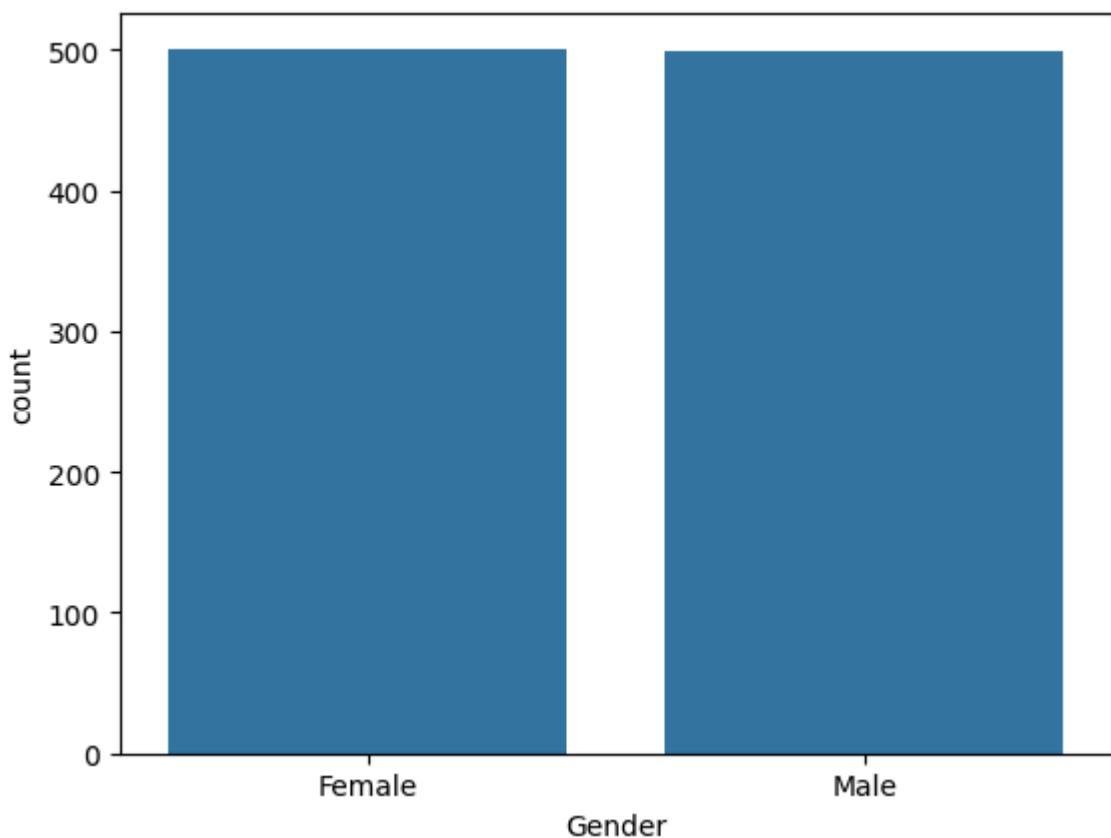
```
In [21]: sns.countplot(x='Customer type', data=df)
df['Customer type'].value_counts()
```

```
Out[21]: Customer type
Member      501
Normal     499
Name: count, dtype: int64
```

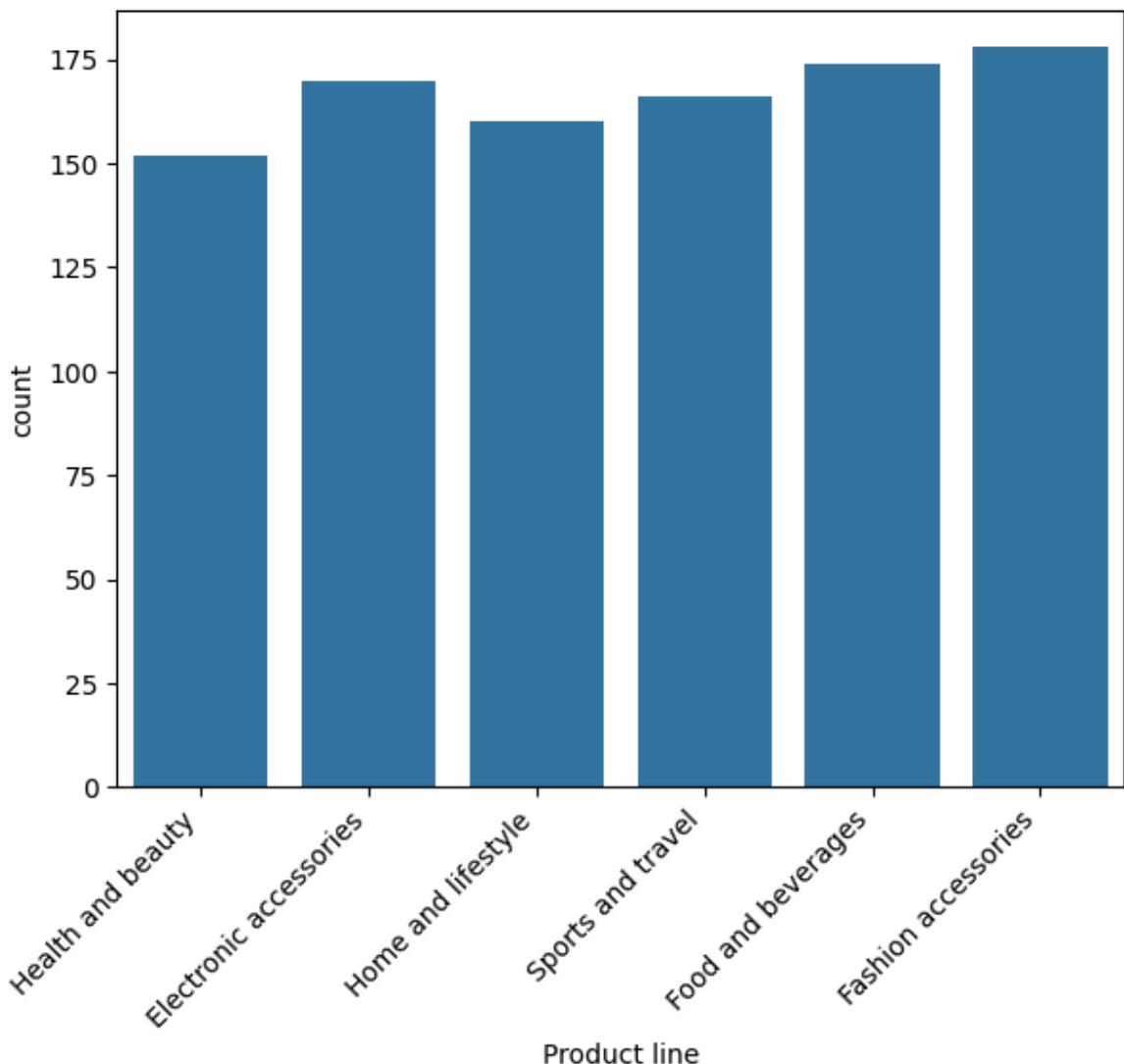


```
In [22]: sns.countplot(x='Gender',data=df)
df['Gender'].value_counts()
```

```
Out[22]: Gender
Female    501
Male      499
Name: count, dtype: int64
```

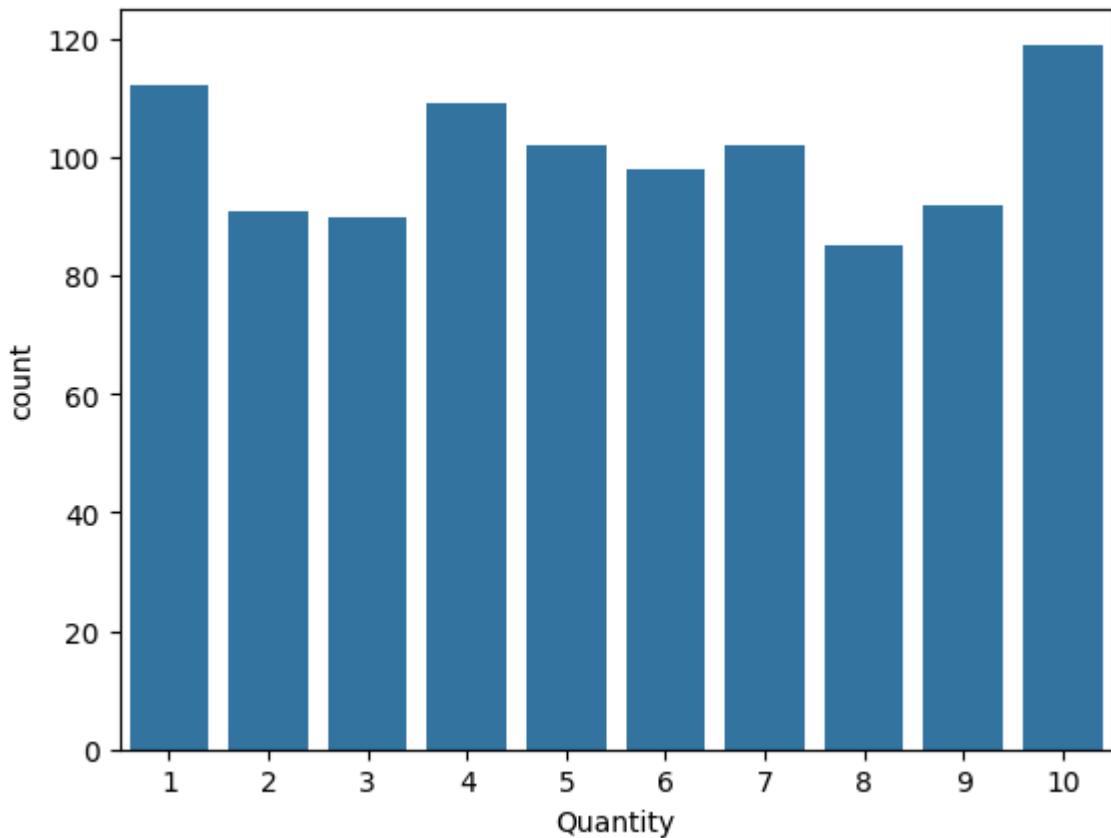


```
In [23]: sns.countplot(x='Product line',data=df)
plt.tight_layout()
plt.xticks(rotation=45,ha='right')
df['Product line'].value_counts()
plt.show()
```



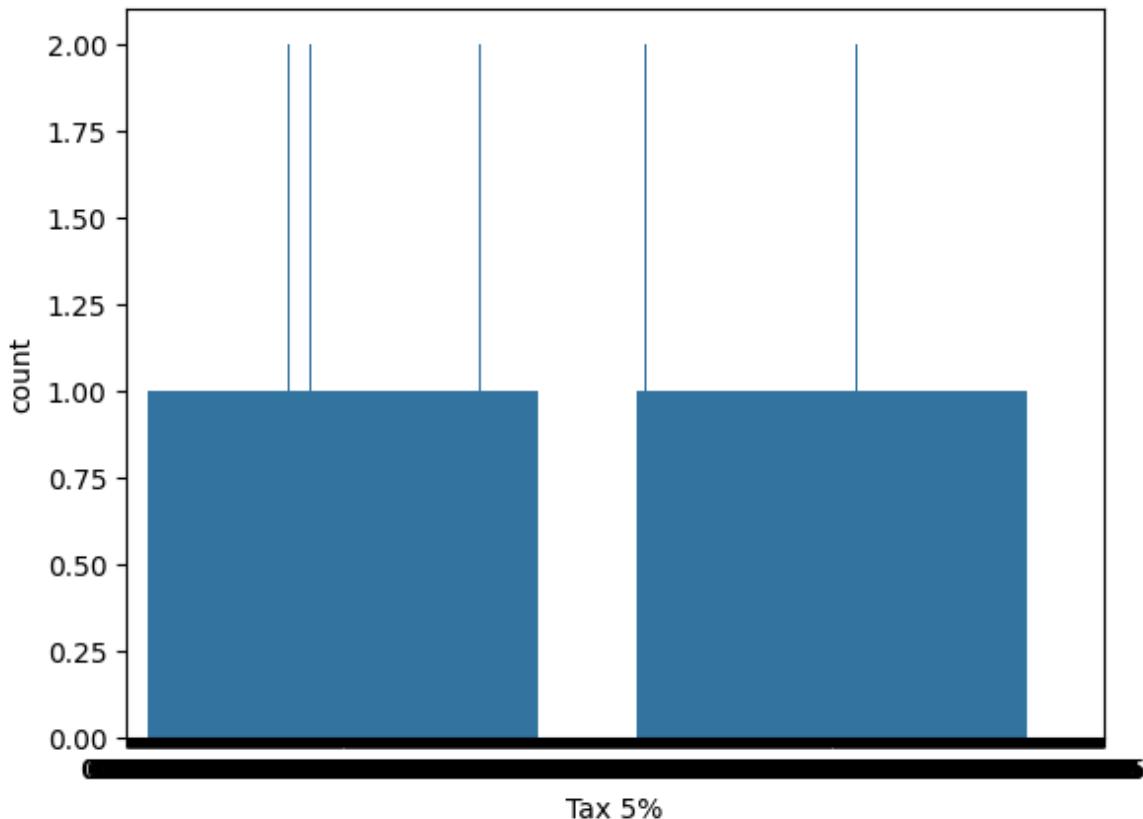
```
In [24]: sns.countplot(x='Quantity', data=df)
```

```
Out[24]: <Axes: xlabel='Quantity', ylabel='count'>
```



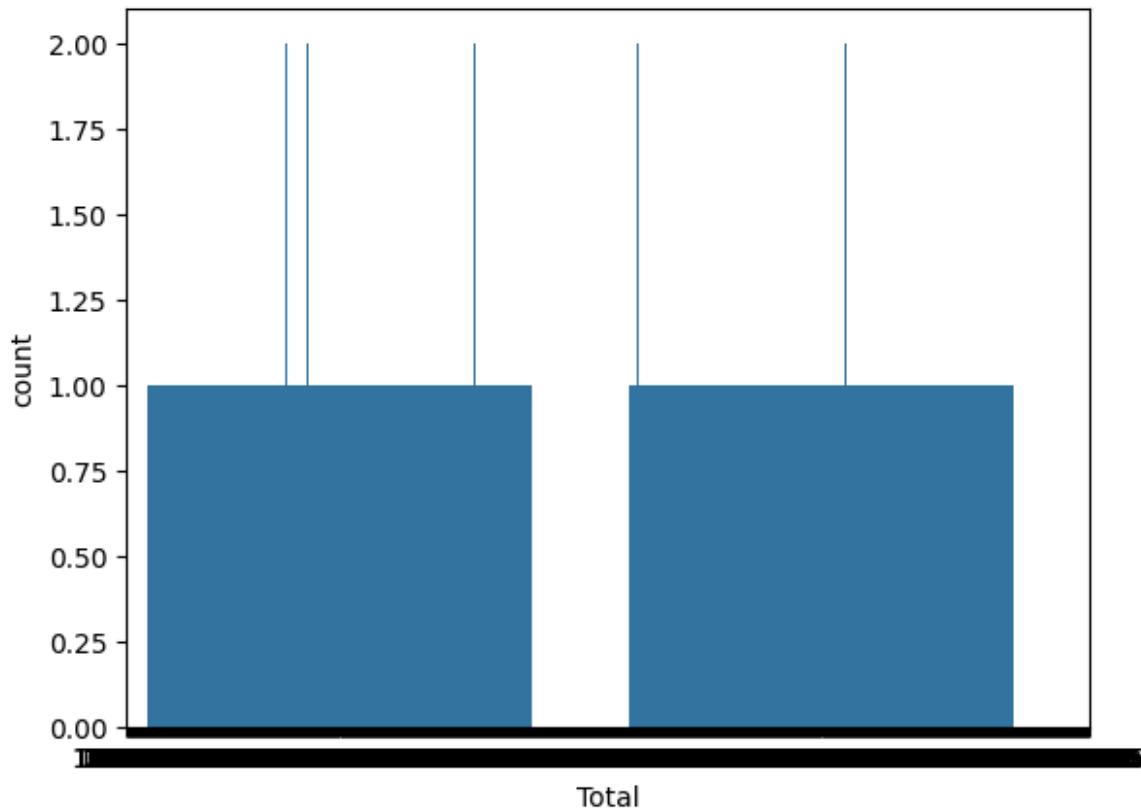
```
In [25]: sns.countplot(x='Tax 5%', data=df)
```

```
Out[25]: <Axes: xlabel='Tax 5%', ylabel='count'>
```



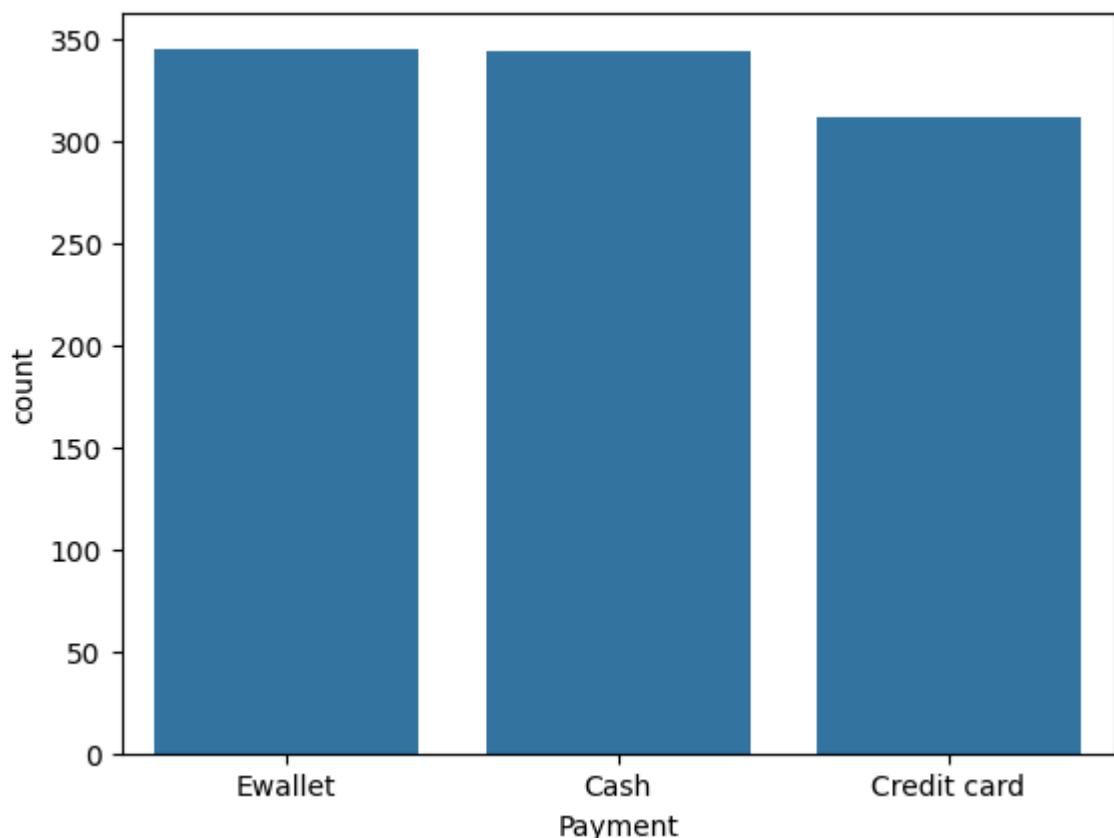
```
In [26]: sns.countplot(x='Total', data=df)
```

```
Out[26]: <Axes: xlabel='Total', ylabel='count'>
```



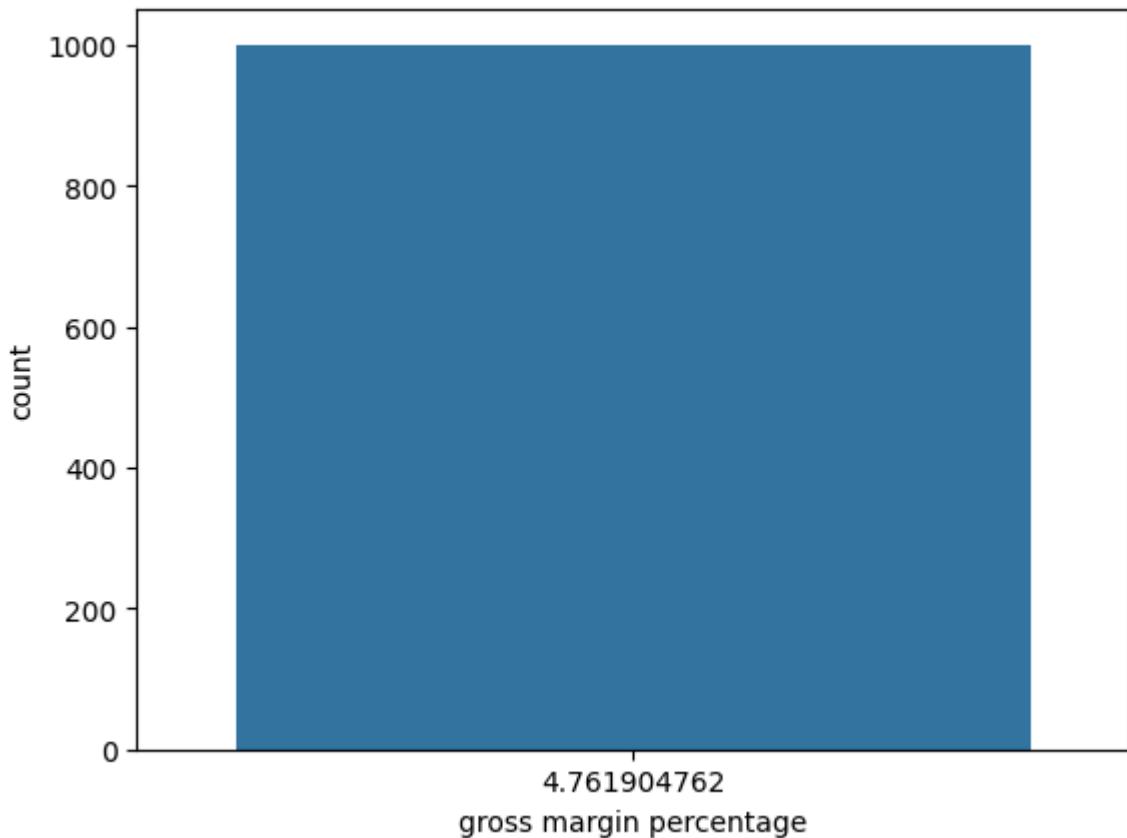
```
In [27]: sns.countplot(x='Payment', data=df)
```

```
Out[27]: <Axes: xlabel='Payment', ylabel='count'>
```



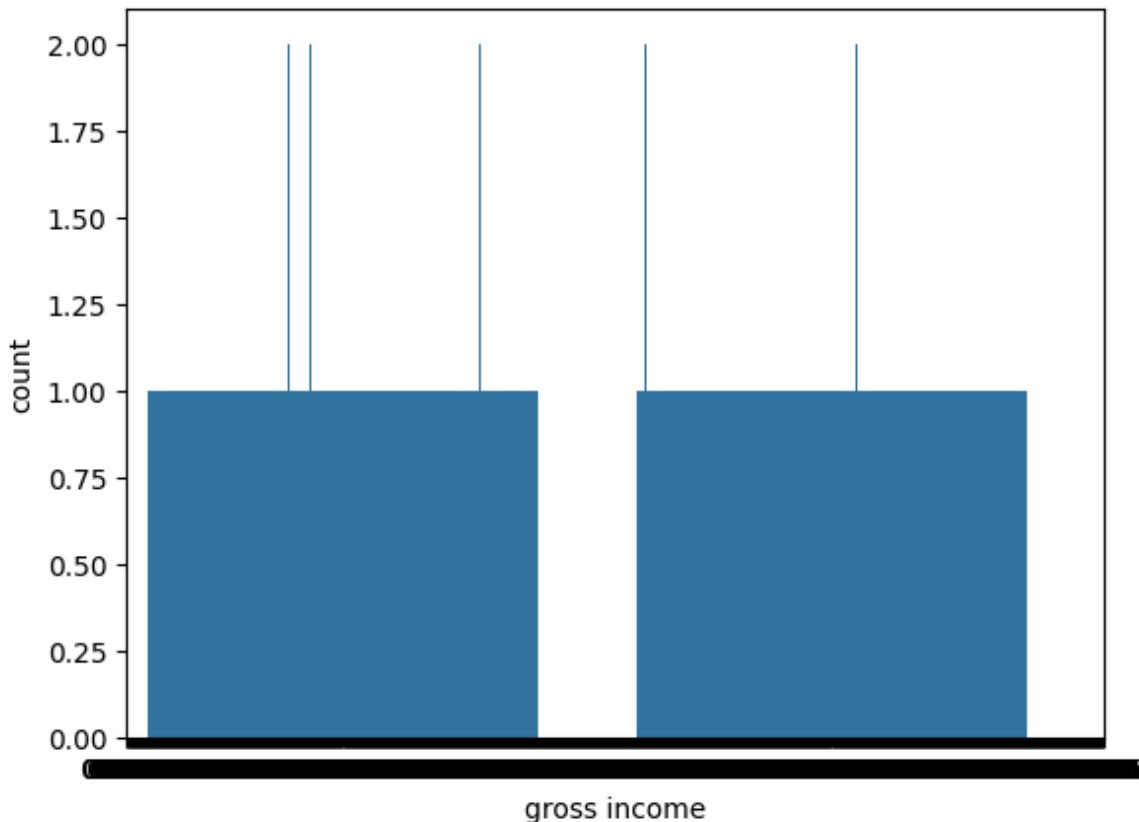
```
In [28]: sns.countplot(x='gross margin percentage', data=df)
```

```
Out[28]: <Axes: xlabel='gross margin percentage', ylabel='count'>
```



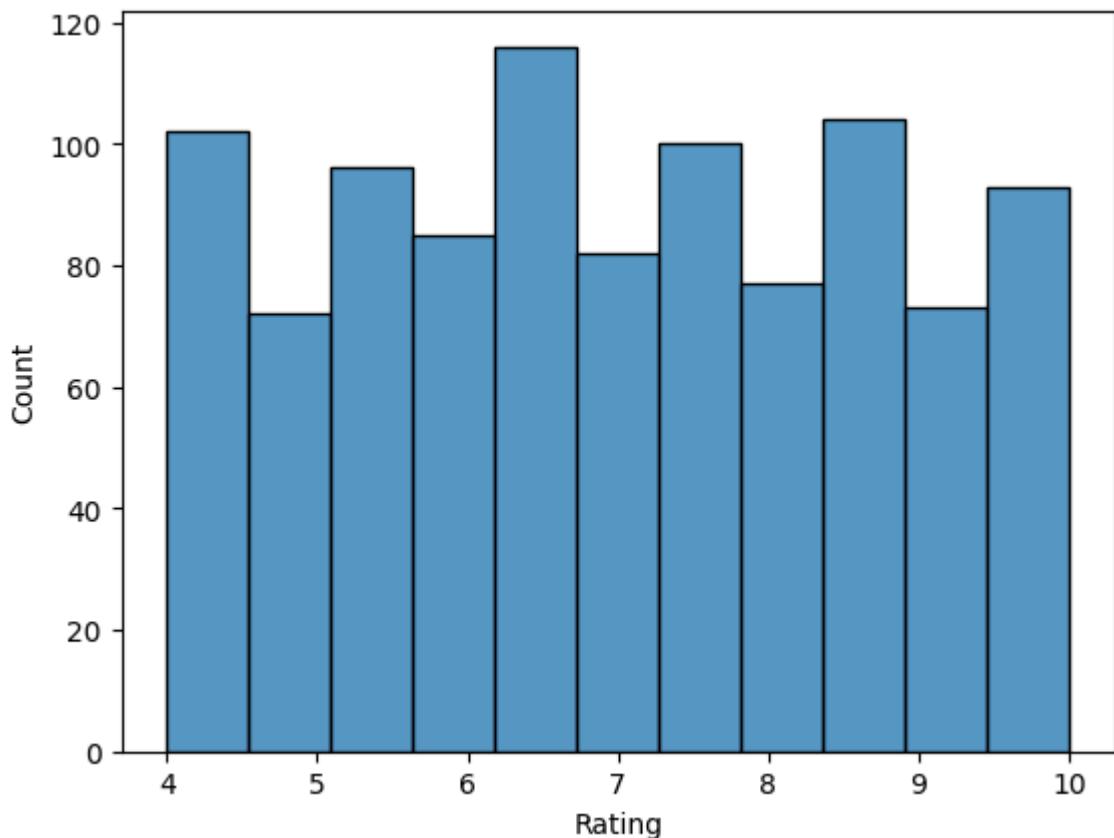
```
In [29]: sns.countplot(x='gross income', data=df)
```

```
Out[29]: <Axes: xlabel='gross income', ylabel='count'>
```



```
In [30]: sns.histplot(x='Rating', data=df)
```

```
Out[30]: <Axes: xlabel='Rating', ylabel='Count'>
```



```
In [1]: #converting data column into data
df['Date']=pd.to_datetime(df['Date'])
```

NameError
Cell In[1], line 2
1 #converting data column into data
----> 2 df['Date']=pd.to_datetime(df['Date'])

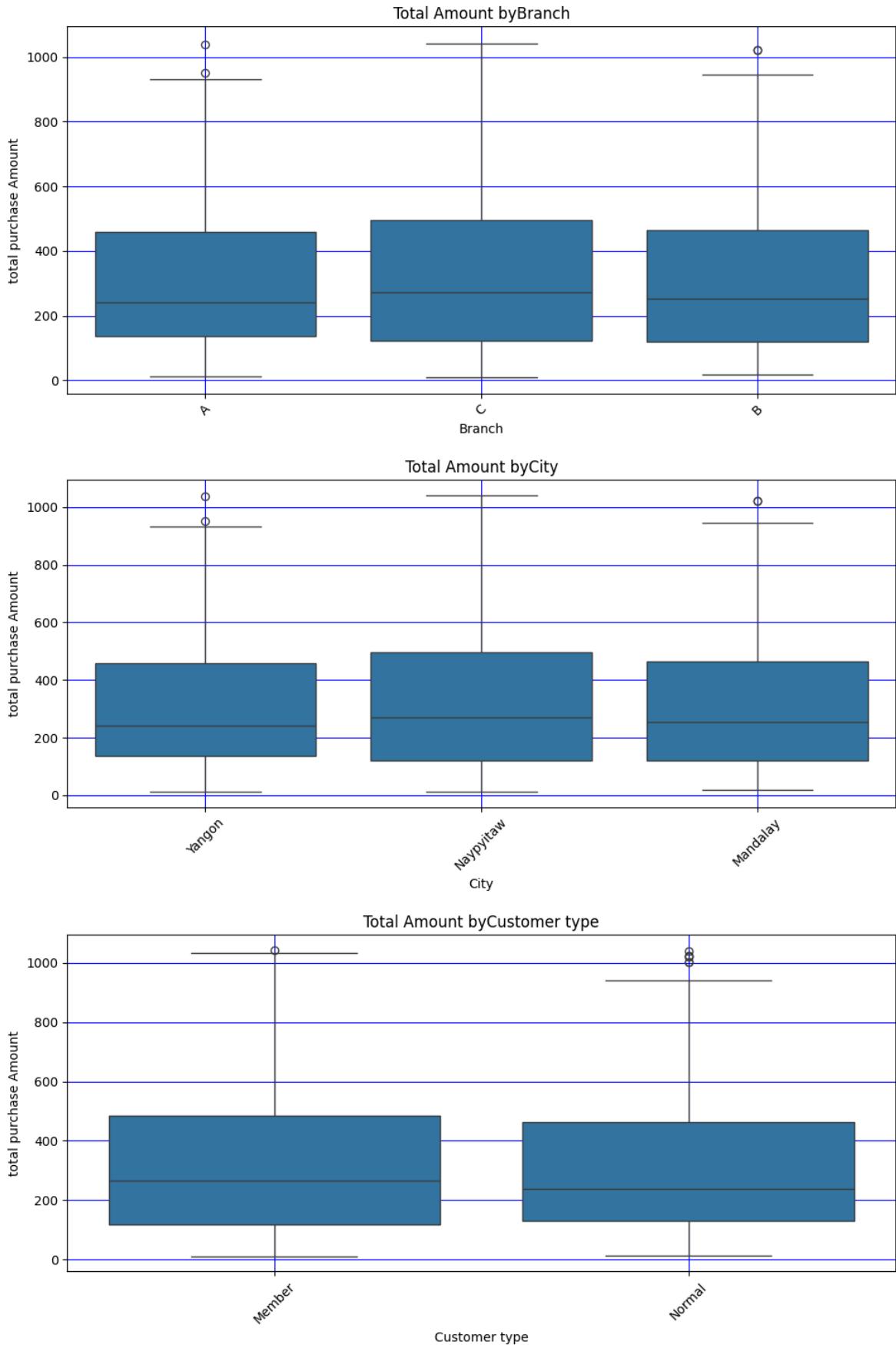
NameError: name 'pd' is not defined

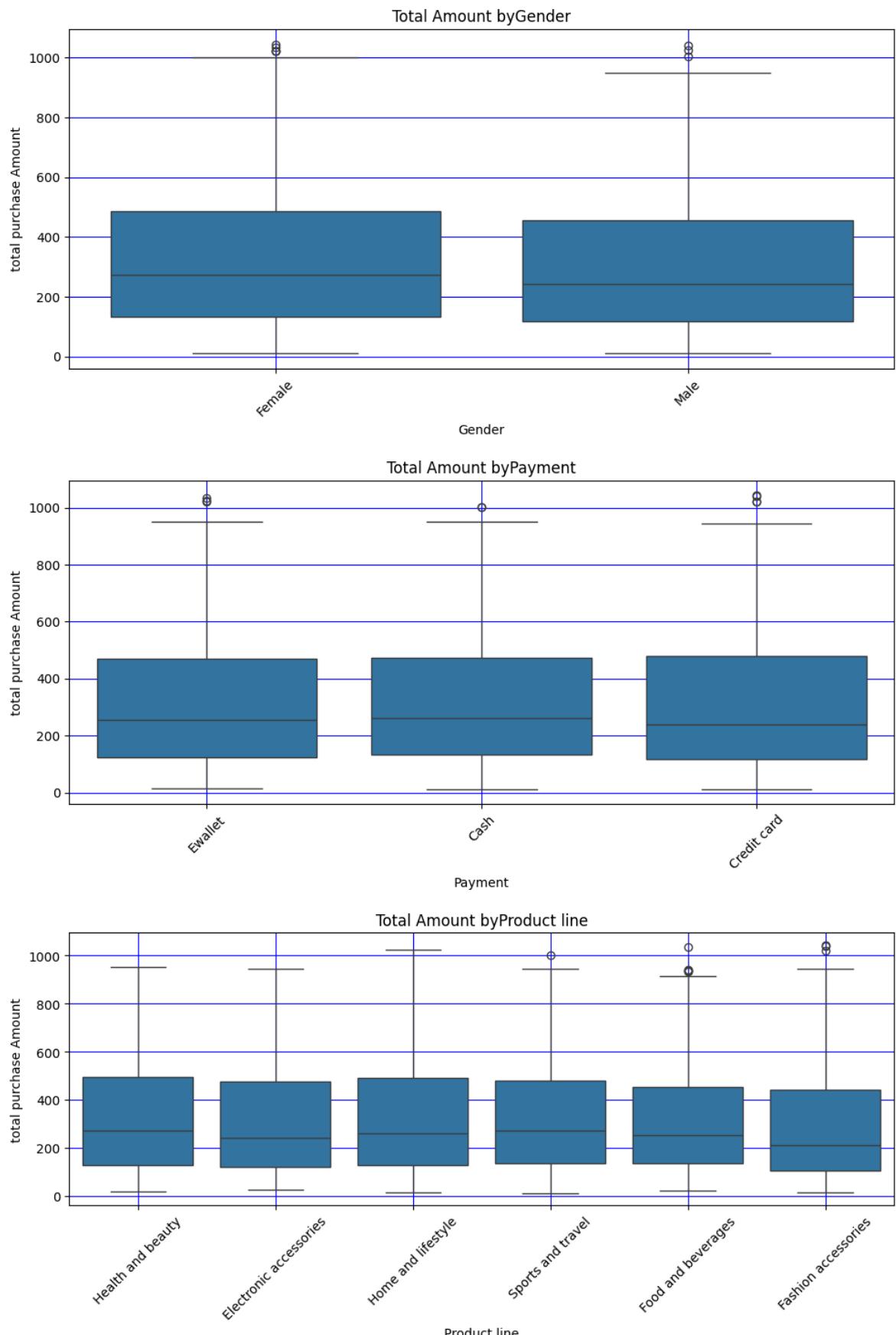
```
In [ ]: df['Date'].isnull().sum()
```

```
In [ ]: df['Date'].fillna(method='ffill',inplace=True)
```

```
In [34]: #list of categorical columns for bivariate analysis
categorical_cols=['Branch', 'City', 'Customer type', 'Gender','Payment',
                  'Product line']
print(categorical_cols)
#creating box plot with loop for bivariate analysis
#total vs categorical
# plot boxplots for 'total' vs categorical variables
for col in categorical_cols:
    plt.figure(figsize=(10,5))
    sns.boxplot(data=df,x=col,y='Total')
    plt.title(f'Total Amount by{col}')
    plt.xlabel(col)
    plt.ylabel('total purchase Amount')
    plt.xticks(rotation=45)
    plt.tight_layout()
    plt.grid(c='blue')
    plt.show()
```

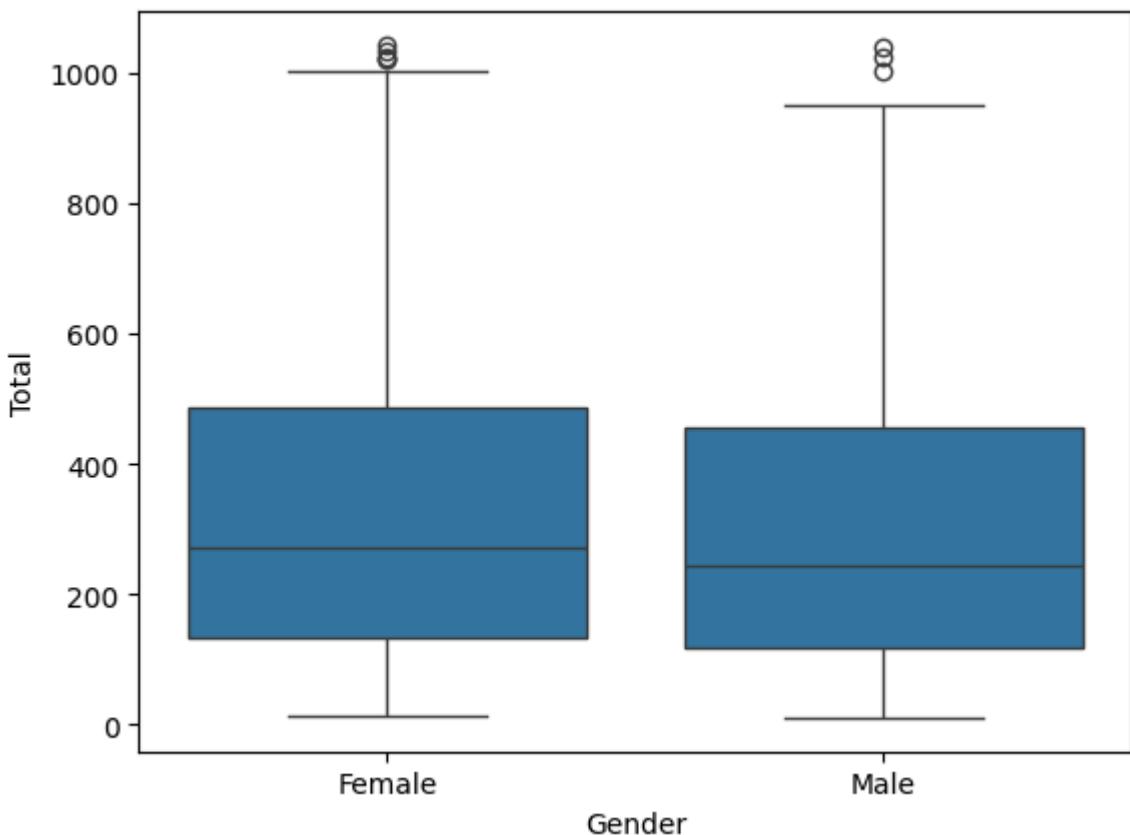
['Branch', 'City', 'Customer type', 'Gender', 'Payment', 'Product line']





```
In [35]: #box plot total,pl g,m,p
sns.boxplot(x='Gender',y='Total',data=df)
```

```
Out[35]: <Axes: xlabel='Gender', ylabel='Total'>
```



```
In [36]: #Identify numerical columns  
df=pd.read_csv('supermarket_sales - Sheet1.csv')  
df
```

Out[36]:

	Invoice ID	Branch	City	Customer type	Gender	Product line	Unit price	Quantity	Tax 5
0	750-67-8428	A	Yangon	Member	Female	Health and beauty	74.69	7	26.14
1	226-31-3081	C	Naypyitaw	Normal	Female	Electronic accessories	15.28	5	3.82
2	631-41-3108	A	Yangon	Normal	Male	Home and lifestyle	46.33	7	16.21
3	123-19-1176	A	Yangon	Member	Male	Health and beauty	58.22	8	23.28
4	373-73-7910	A	Yangon	Normal	Male	Sports and travel	86.31	7	30.20
...
995	233-67-5758	C	Naypyitaw	Normal	Male	Health and beauty	40.35	1	2.01
996	303-96-2227	B	Mandalay	Normal	Female	Home and lifestyle	97.38	10	48.69
997	727-02-1313	A	Yangon	Member	Male	Food and beverages	31.84	1	1.59
998	347-56-2442	A	Yangon	Normal	Male	Home and lifestyle	65.82	1	3.29
999	849-09-3807	A	Yangon	Member	Female	Fashion accessories	88.34	7	30.91

1000 rows × 17 columns



In [41]:

```

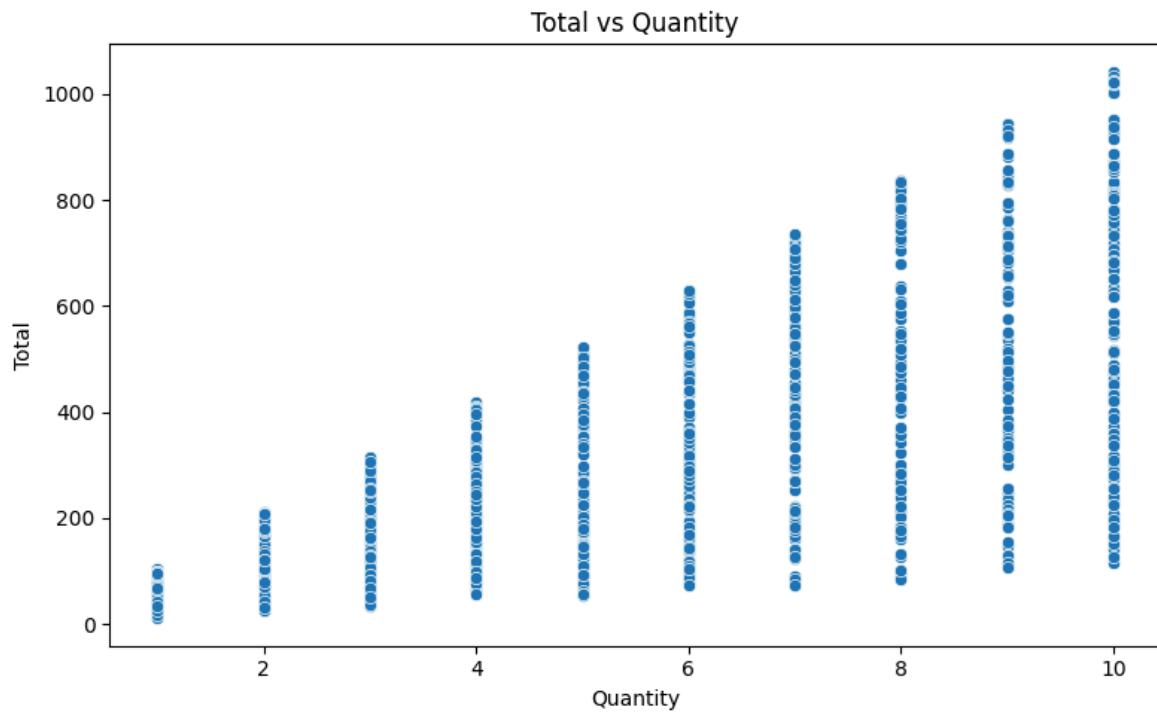
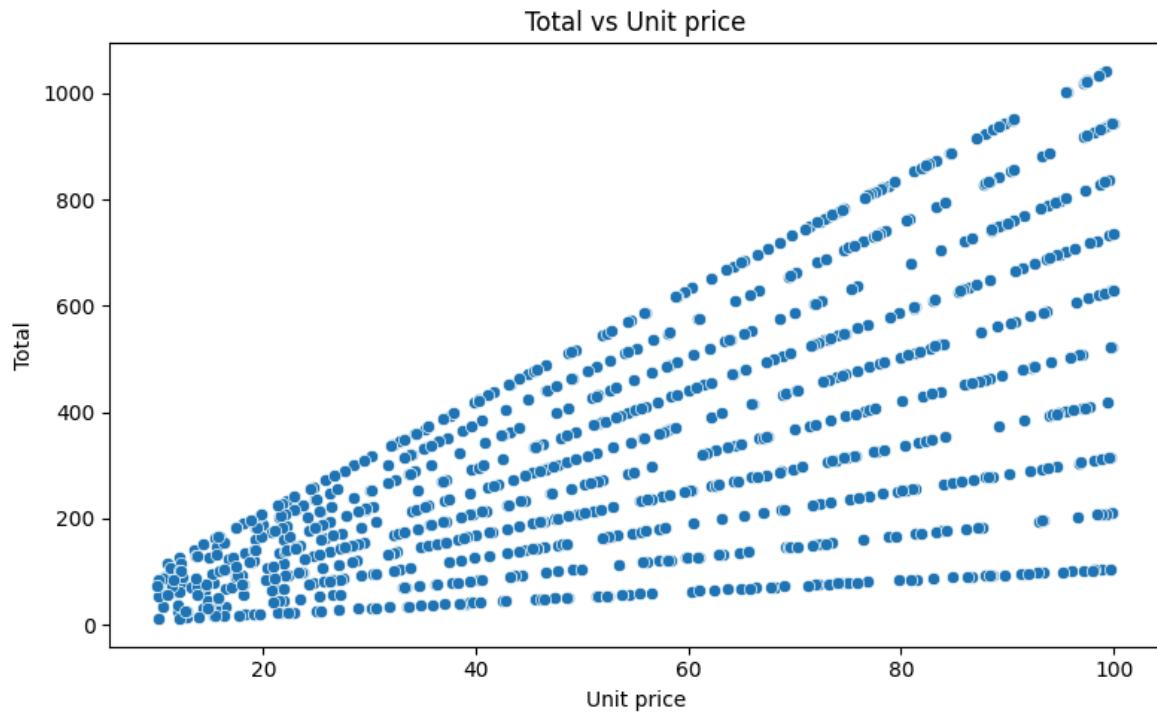
numerical_cols=df.select_dtypes(include=['float64','int64']).columns
print(numerical_cols)
scatter_pairs=[

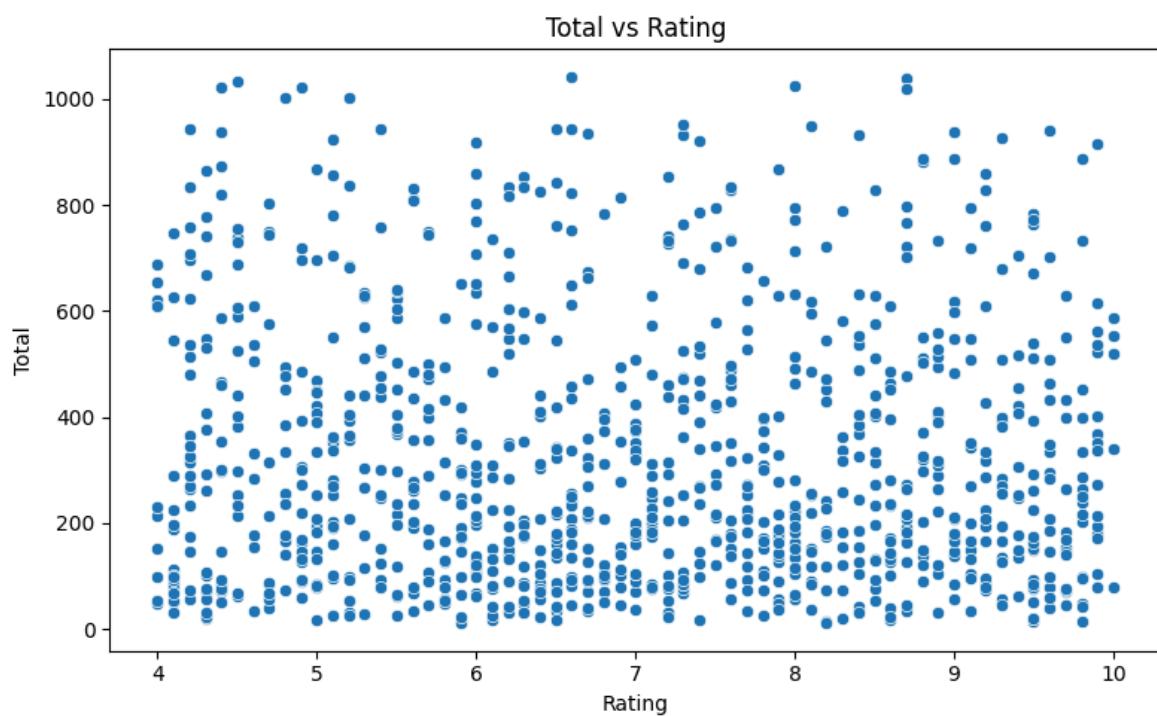
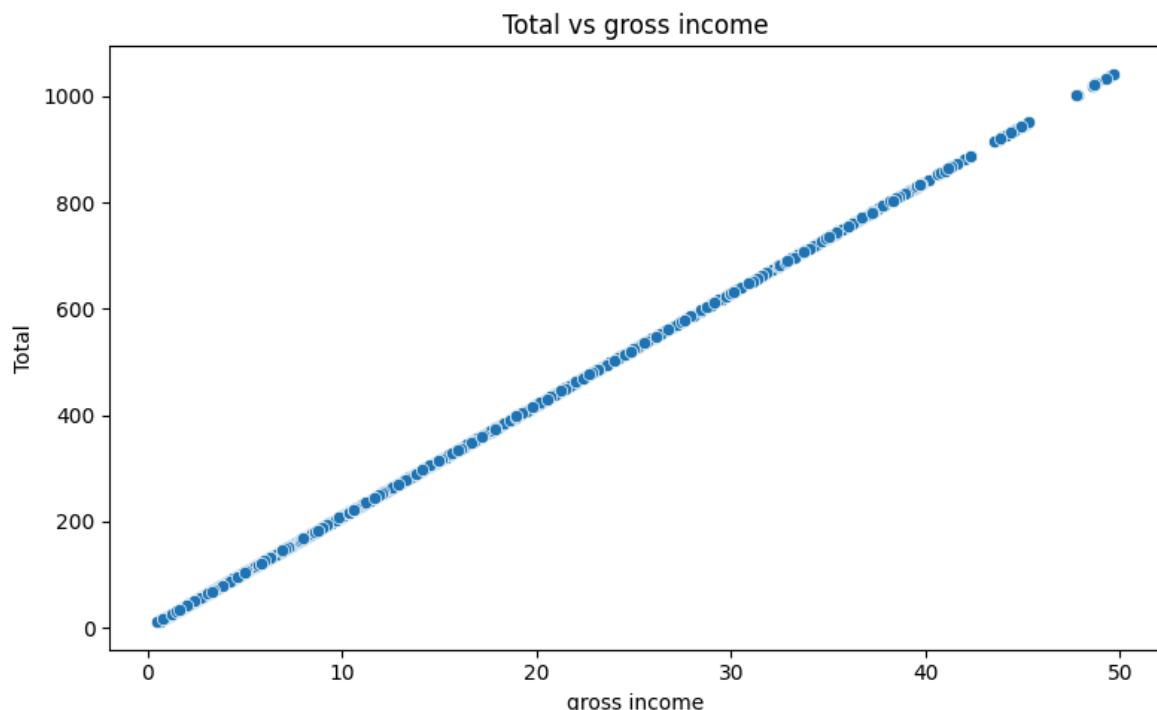
    ('Unit price','Total'),
    ('Quantity','Total'),
    ('gross income','Total'),
    ('Rating','Total')
]
print(scatter_pairs)
for x,y in scatter_pairs:
    plt.figure(figsize=(8,5))
    sns.scatterplot(data=df,x=x,y=y)

```

```
plt.title(f'{y} vs {x}')
plt.tight_layout()

Index(['Unit price', 'Quantity', 'Tax 5%', 'Total', 'cogs',
       'gross margin percentage', 'gross income', 'Rating'],
      dtype='object')
[('Unit price', 'Total'), ('Quantity', 'Total'), ('gross income', 'Total'), ('Rating', 'Total')]
```





```
In [42]: #code for time-based Trend Analysis  
#convert 'Data' and 'Time' To datetime objects  
import pandas as pd  
df['Date']=pd.to_datetime(df['Date'],errors='coerce')  
df['Time']=pd.to_datetime(df['Time'],format='%H:%M').dt.time
```

```
In [43]: df.dtypes
```

```
Out[43]: 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             object
Payment          object
cogs             float64
gross margin percentage float64
gross income     float64
Rating           float64
dtype: object
```

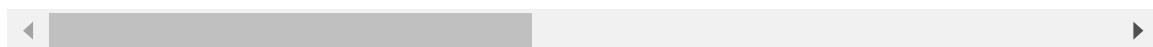
```
In [44]: #df['Date'].head(50)
#create additional Time-based feature
df['Day']=df['Date'].dt.date
df['Month']=df['Date'].dt.to_period('M')
df['Weekday']=df['Date'].dt.day_name()
```

```
In [48]: df
```

Out[48]:

	Invoice ID	Branch	City	Customer type	Gender	Product line	Unit price	Quantity	Tax 5
0	750-67-8428	A	Yangon	Member	Female	Health and beauty	74.69	7	26.14
1	226-31-3081	C	Naypyitaw	Normal	Female	Electronic accessories	15.28	5	3.82
2	631-41-3108	A	Yangon	Normal	Male	Home and lifestyle	46.33	7	16.21
3	123-19-1176	A	Yangon	Member	Male	Health and beauty	58.22	8	23.28
4	373-73-7910	A	Yangon	Normal	Male	Sports and travel	86.31	7	30.20
...
995	233-67-5758	C	Naypyitaw	Normal	Male	Health and beauty	40.35	1	2.01
996	303-96-2227	B	Mandalay	Normal	Female	Home and lifestyle	97.38	10	48.69
997	727-02-1313	A	Yangon	Member	Male	Food and beverages	31.84	1	1.59
998	347-56-2442	A	Yangon	Normal	Male	Home and lifestyle	65.82	1	3.29
999	849-09-3807	A	Yangon	Member	Female	Fashion accessories	88.34	7	30.91

1000 rows × 21 columns



In [51]:

```
#covert 'Time' to hour only
df['Hour']=pd.to_datetime(df['Time'],format='%H,%M,%S',errors='coerce').apply(lambda x:x.hour)
```

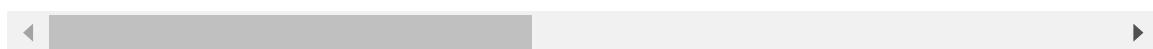
In [52]:

df

Out[52]:

	Invoice ID	Branch	City	Customer type	Gender	Product line	Unit price	Quantity	Tax	5
0	750-67-8428	A	Yangon	Member	Female	Health and beauty	74.69	7	26.14	
1	226-31-3081	C	Naypyitaw	Normal	Female	Electronic accessories	15.28	5	3.82	
2	631-41-3108	A	Yangon	Normal	Male	Home and lifestyle	46.33	7	16.21	
3	123-19-1176	A	Yangon	Member	Male	Health and beauty	58.22	8	23.28	
4	373-73-7910	A	Yangon	Normal	Male	Sports and travel	86.31	7	30.20	
...
995	233-67-5758	C	Naypyitaw	Normal	Male	Health and beauty	40.35	1	2.01	
996	303-96-2227	B	Mandalay	Normal	Female	Home and lifestyle	97.38	10	48.69	
997	727-02-1313	A	Yangon	Member	Male	Food and beverages	31.84	1	1.59	
998	347-56-2442	A	Yangon	Normal	Male	Home and lifestyle	65.82	1	3.29	
999	849-09-3807	A	Yangon	Member	Female	Fashion accessories	88.34	7	30.91	

1000 rows × 21 columns



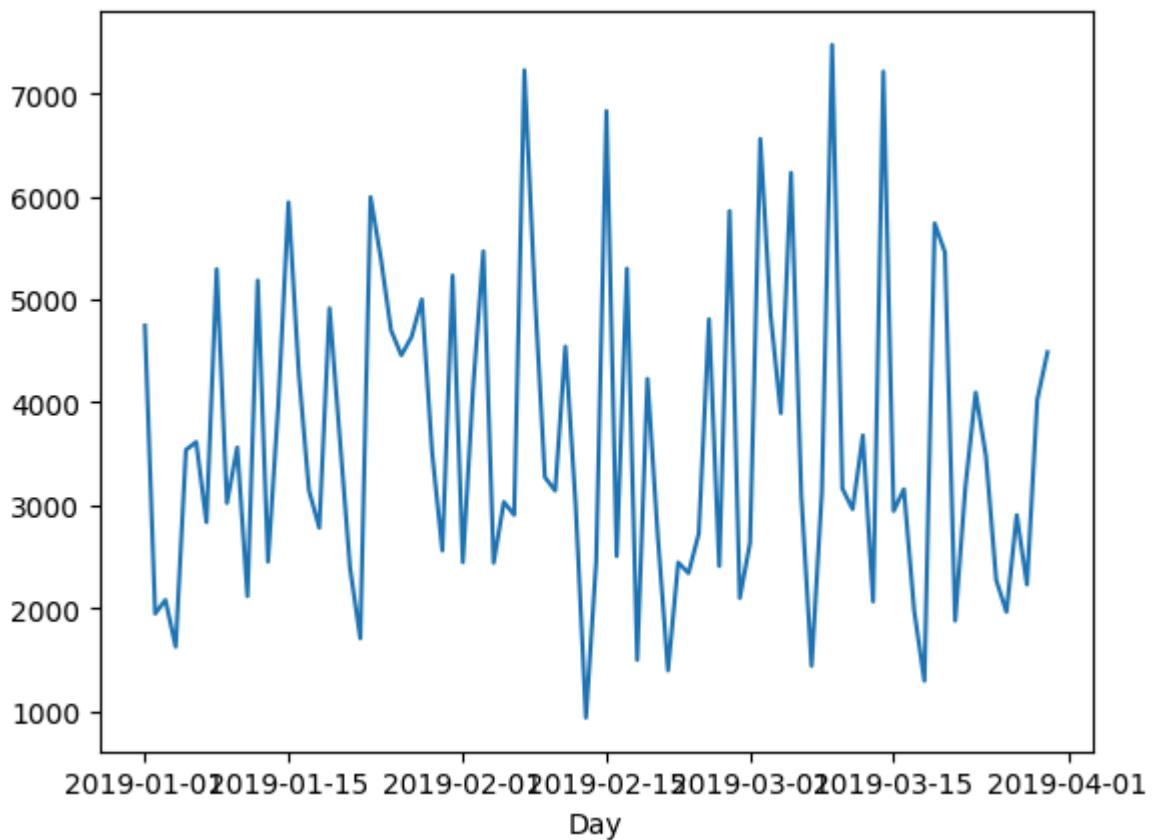
In [53]:

```
#date and month wise total sales
Total_sum=df.groupby('Day')[['Total']].sum()
print(Total_sum)
#month wise
Total_monthwise=df.groupby('Month')[['Total']].sum()
print(Total_monthwise)
#day wise Total
Day_wise_salse=df.groupby('Weekday')[['Total']].sum()
print(Day_wise_salse)
```

```
Day
2019-01-01    4745.1810
2019-01-02    1945.5030
2019-01-03    2078.1285
2019-01-04    1623.6885
2019-01-05    3536.6835
...
2019-03-26    1962.5130
2019-03-27    2902.8195
2019-03-28    2229.4020
2019-03-29    4023.2430
2019-03-30    4487.0595
Name: Total, Length: 89, dtype: float64
Month
2019-01    116291.868
2019-02    97219.374
2019-03    109455.507
Freq: M, Name: Total, dtype: float64
Weekday
Friday      43926.3405
Monday      37899.0780
Saturday    56120.8095
Sunday      44457.8925
Thursday    45349.2480
Tuesday     51482.2455
Wednesday   43731.1350
Name: Total, dtype: float64
```

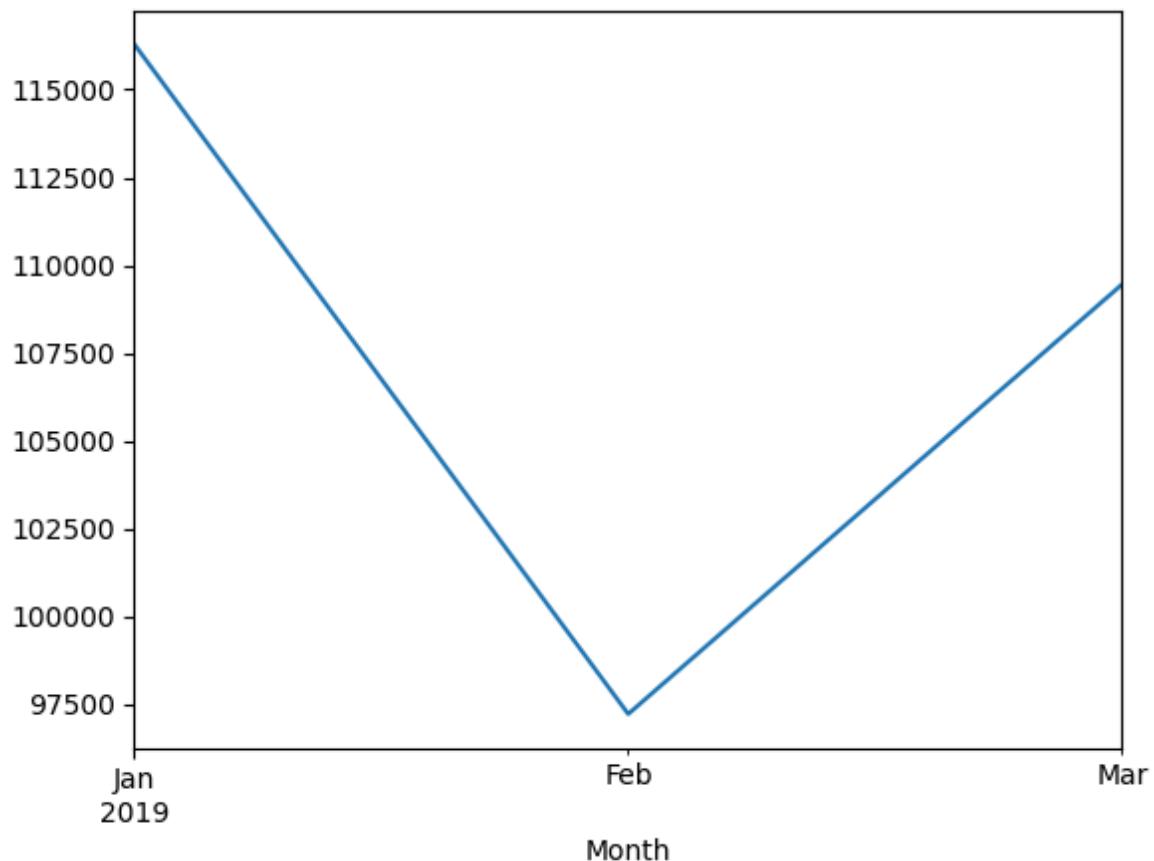
```
In [54]: Total_sum.plot()
```

```
Out[54]: <Axes: xlabel='Day'>
```



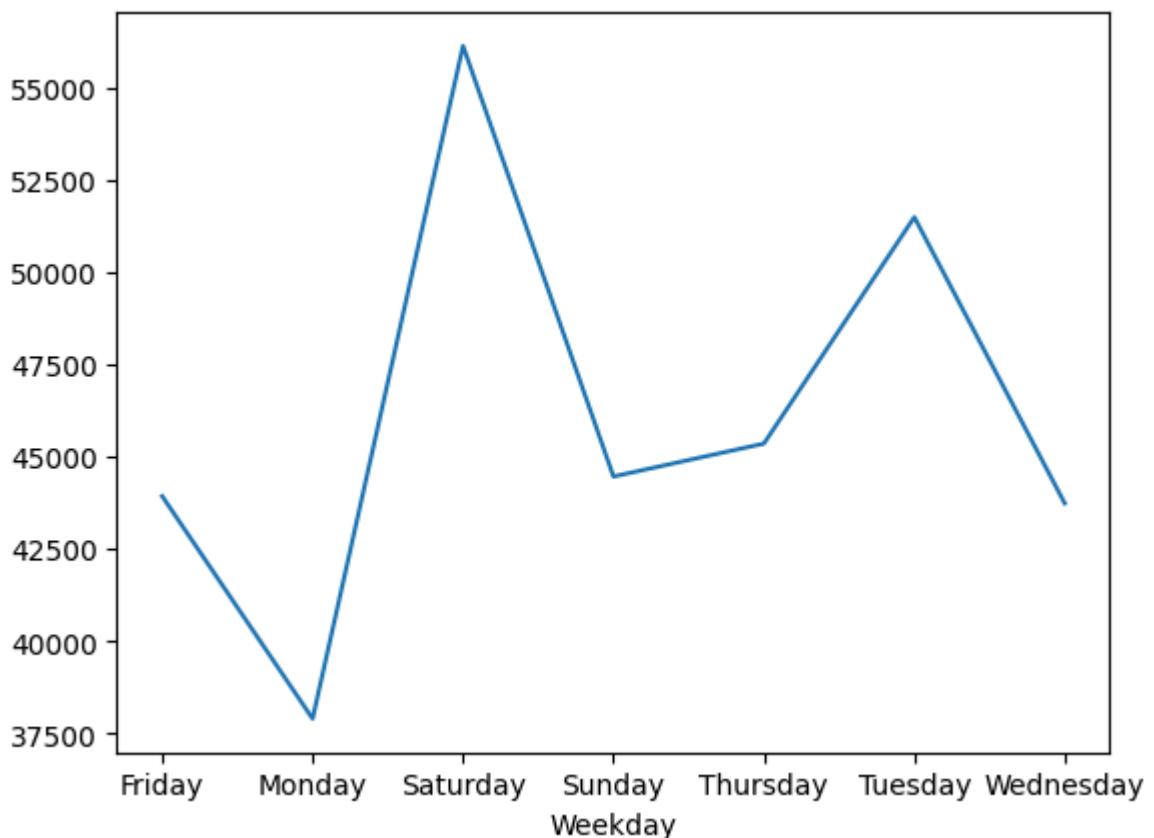
```
In [55]: Total_monthwise.plot()
```

```
Out[55]: <Axes: xlabel='Month'>
```



```
In [56]: Day_wise_salse.plot()
```

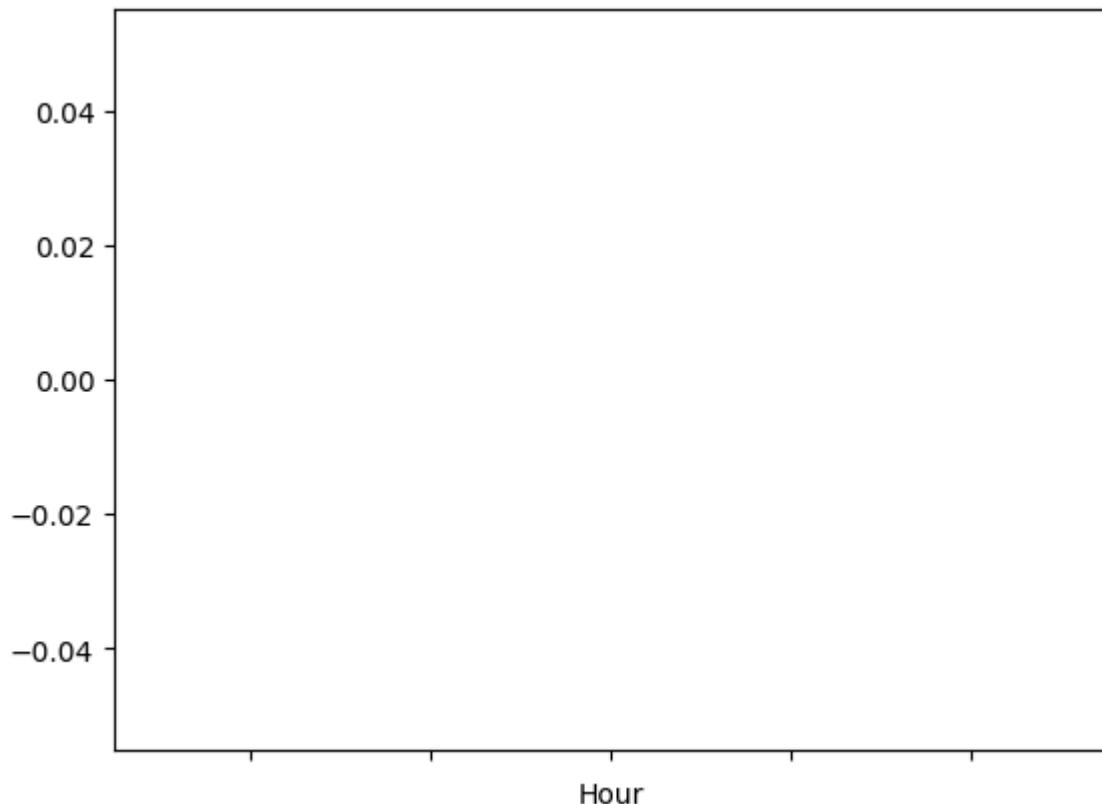
```
Out[56]: <Axes: xlabel='Weekday'>
```



```
In [57]: hourly_sales=df.groupby('Hour')['Total'].sum()  
print(hourly_sales)  
hourly_sales.plot()
```

```
Series([], Name: Total, dtype: float64)
```

```
Out[57]: <Axes: xlabel='Hour'>
```



```
In [ ]:
```