

Sales_project

July 4, 2025

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

```
[5]: df=pd.read_csv('/Users/akula/Desktop/Diwali Sales Data.csv', encoding=
↳ 'unicode_escape')
```

```
[6]: df.shape
```

```
[6]: (11251, 15)
```

the table data consists of 11251 rows and 15 columns

```
[7]: df.head(10)
```

```
[7]:   User_ID  Cust_name  Product_ID  Gender  Age  Group  Age  Marital_Status  \
0  1002903  Sanskriti  P00125942      F    26-35  28           0
1  1000732    Kartik  P00110942      F    26-35  35           1
2  1001990    Bindu  P00118542      F    26-35  35           1
3  1001425    Sudevi  P00237842      M     0-17  16           0
4  1000588     Joni  P00057942      M    26-35  28           1
5  1000588     Joni  P00057942      M    26-35  28           1
6  1001132     Balk  P00018042      F    18-25  25           1
7  1002092  Shivangi  P00273442      F     55+  61           0
8  1003224    Kushal  P00205642      M    26-35  35           0
9  1003650    Ginny  P00031142      F    26-35  26           1
```

```
   State      Zone  Occupation  Product_Category  Orders  \
0  Maharashtra  Western  Healthcare           Auto         1
1  Andhra Pradesh  Southern           Govt           Auto         3
2  Uttar Pradesh  Central  Automobile           Auto         3
3  Karnataka  Southern  Construction           Auto         2
4  Gujarat  Western  Food Processing           Auto         2
5  Himachal Pradesh  Northern  Food Processing           Auto         1
6  Uttar Pradesh  Central           Lawyer           Auto         4
7  Maharashtra  Western  IT Sector           Auto         1
8  Uttar Pradesh  Central           Govt           Auto         2
```

9	Andhra Pradesh	Southern	Media	Auto	4
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	Amount	Status	unnamed1
0	23952.00	NaN	NaN
1	23934.00	NaN	NaN
2	23924.00	NaN	NaN
3	23912.00	NaN	NaN
4	23877.00	NaN	NaN
5	23877.00	NaN	NaN
6	23841.00	NaN	NaN
7	NaN	NaN	NaN
8	23809.00	NaN	NaN
9	23799.99	NaN	NaN

```
[8]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 11251 entries, 0 to 11250
Data columns (total 15 columns):
#   Column                Non-Null Count  Dtype
---  -
0   User_ID               11251 non-null  int64
1   Cust_name             11251 non-null  object
2   Product_ID           11251 non-null  object
3   Gender               11251 non-null  object
4   Age Group            11251 non-null  object
5   Age                  11251 non-null  int64
6   Marital_Status       11251 non-null  int64
7   State               11251 non-null  object
8   Zone                11251 non-null  object
9   Occupation           11251 non-null  object
10  Product_Category     11251 non-null  object
11  Orders              11251 non-null  int64
12  Amount              11239 non-null  float64
13  Status              0 non-null      float64
14  unnamed1            0 non-null      float64
dtypes: float64(3), int64(4), object(8)
memory usage: 1.3+ MB
```

in above data two columns having null values and one column is with wrong data type that is amount should be in int but it is in float we have to change it and remove two null rows

```
[12]: df.shape
```

```
[12]: (11251, 13)
```

after removing the null columns from the table we have 11251 rows and 13 columns

```
[16]: df.isnull().sum()
```

```
[16]: User_ID      0
      Cust_name    0
      Product_ID   0
      Gender       0
      Age Group    0
      Age          0
      Marital_Status 0
      State        0
      Zone         0
      Occupation   0
      Product_Category 0
      Orders       0
      Amount      12
      dtype: int64
```

```
[17]: df.dropna(inplace=True)
```

```
[19]: df.shape
```

```
[19]: (11239, 13)
```

after removing all null values we have 11239 rows and 13 columns

```
[22]: df['Amount'] =df['Amount'].astype('int')
```

```
[24]: df['Amount'].dtypes
```

```
[24]: dtype('int64')
```

```
[27]: df.rename(columns= {'Marital_Status':'Shaadi'})
```

```
[27]:
```

	User_ID	Cust_name	Product_ID	Gender	Age	Group	Age	Shaadi	\
0	1002903	Sanskriti	P00125942	F	26-35	28	0		
1	1000732	Kartik	P00110942	F	26-35	35	1		
2	1001990	Bindu	P00118542	F	26-35	35	1		
3	1001425	Sudevi	P00237842	M	0-17	16	0		
4	1000588	Joni	P00057942	M	26-35	28	1		
...		
11246	1000695	Manning	P00296942	M	18-25	19	1		
11247	1004089	Reichenbach	P00171342	M	26-35	33	0		
11248	1001209	Oshin	P00201342	F	36-45	40	0		
11249	1004023	Noonan	P00059442	M	36-45	37	0		
11250	1002744	Brumley	P00281742	F	18-25	19	0		

	State	Zone	Occupation	Product_Category	Orders	\
0	Maharashtra	Western	Healthcare	Auto	1	
1	Andhra Pradesh	Southern	Govt	Auto	3	
2	Uttar Pradesh	Central	Automobile	Auto	3	

3	Karnataka	Southern	Construction	Auto	2
4	Gujarat	Western	Food Processing	Auto	2
...
11246	Maharashtra	Western	Chemical	Office	4
11247	Haryana	Northern	Healthcare	Veterinary	3
11248	Madhya Pradesh	Central	Textile	Office	4
11249	Karnataka	Southern	Agriculture	Office	3
11250	Maharashtra	Western	Healthcare	Office	3

	Amount
0	23952
1	23934
2	23924
3	23912
4	23877
...	...
11246	370
11247	367
11248	213
11249	206
11250	188

[11239 rows x 13 columns]

```
[28]: df.columns
```

```
[28]: Index(['User_ID', 'Cust_name', 'Product_ID', 'Gender', 'Age Group', 'Age',
          'Marital_Status', 'State', 'Zone', 'Occupation', 'Product_Category',
          'Orders', 'Amount'],
          dtype='object')
```

```
[41]: Continous=['Age']
discrete_count=['Orders', 'Amount']
discrete_categorical=['Cust_name', 'Gender', 'Age_
↳Group', 'Marital_Status', 'State', 'Zone', 'Occupation', 'Product_Category', 'Product_ID']
```

```
[42]: df[Continous].describe()
```

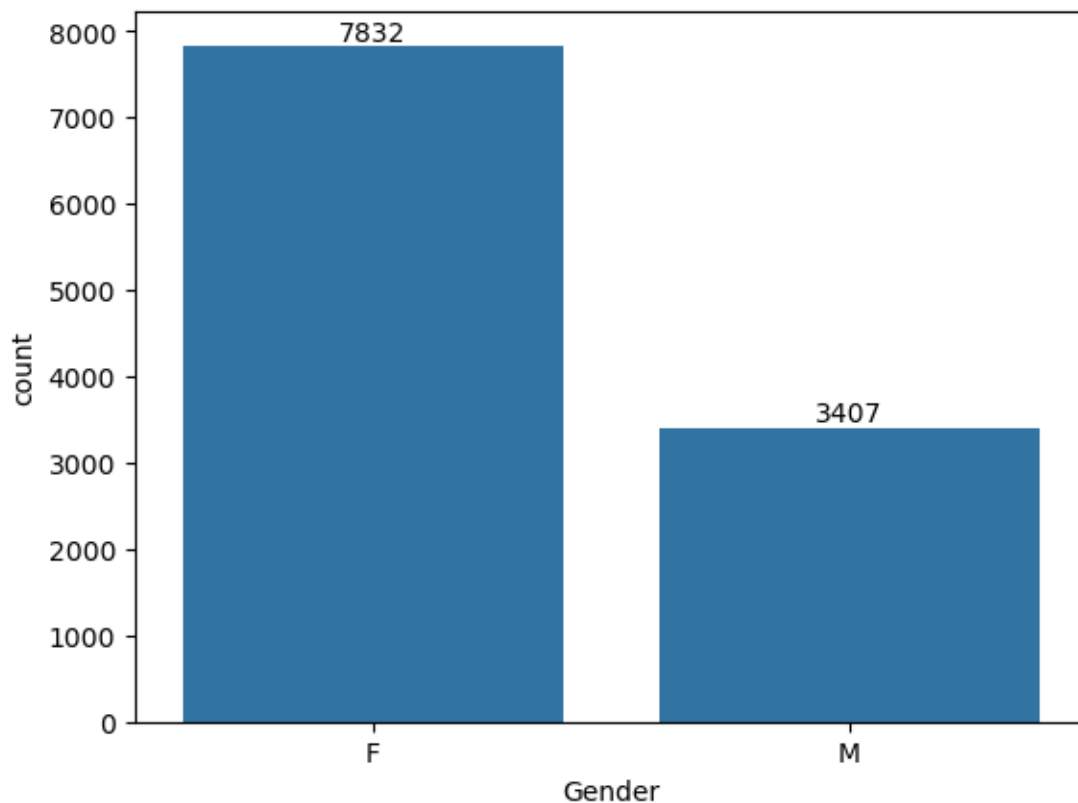
```
[42]:
          Age
count  11239.000000
mean    35.410357
std     12.753866
min     12.000000
25%     27.000000
50%     33.000000
75%     43.000000
max     92.000000
```

```
[43]: df[discrete_count].describe()
```

```
[43]:
```

	Orders	Amount
count	11239.000000	11239.000000
mean	2.489634	9453.610553
std	1.114967	5222.355168
min	1.000000	188.000000
25%	2.000000	5443.000000
50%	2.000000	8109.000000
75%	3.000000	12675.000000
max	4.000000	23952.000000

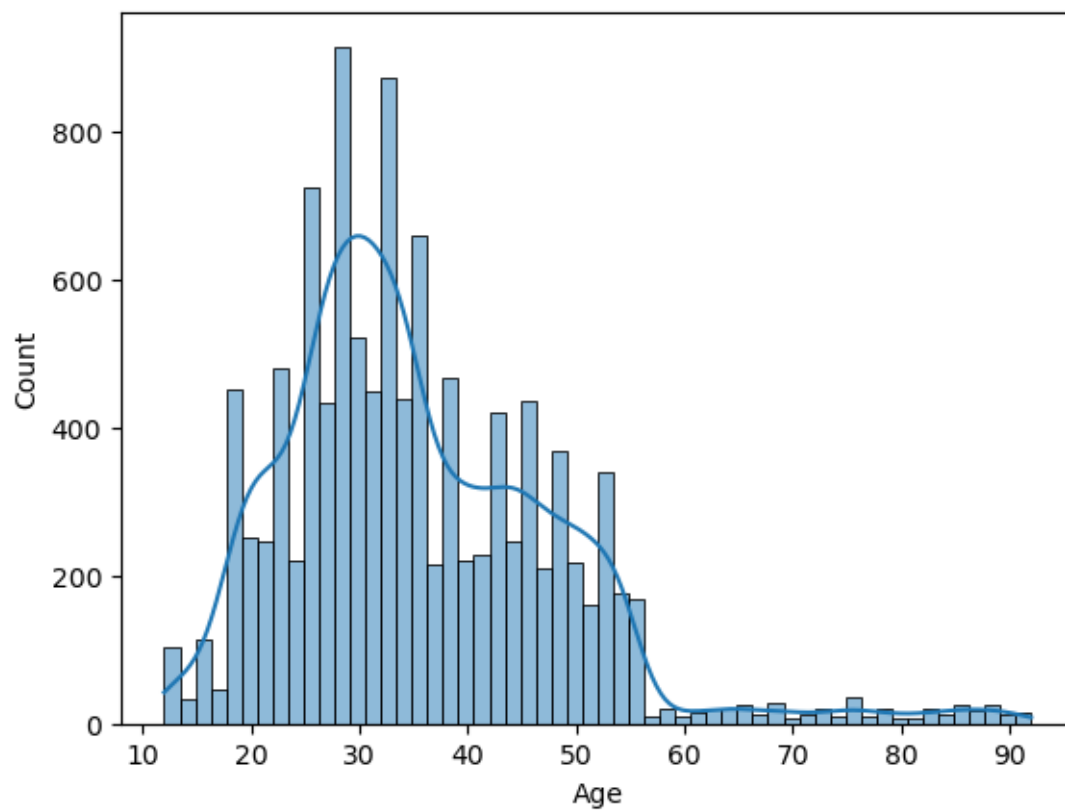
```
[61]: count = sns.countplot(x=df['Gender'])  
for bars in count.containers:  
    count.bar_label(bars)
```



from above bar plot there are 7832 females and 3407 males

```
[62]: sns.histplot(df['Age'],kde=True)
```

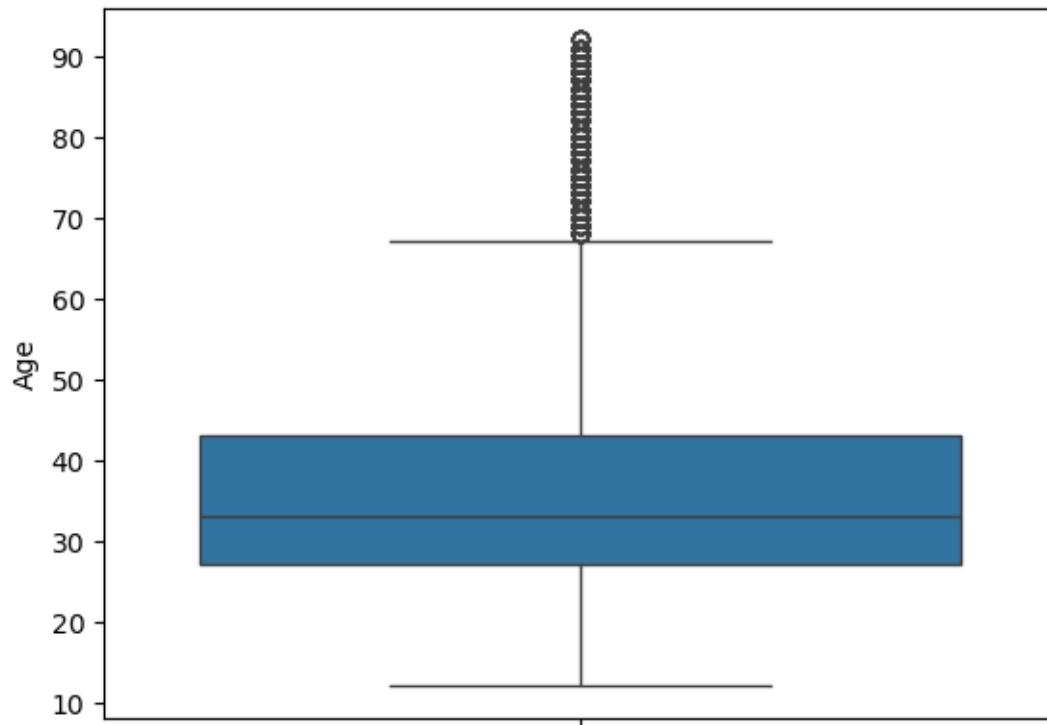
```
[62]: <Axes: xlabel='Age', ylabel='Count'>
```



from above observation it is a right skewed

```
[52]: sns.boxplot(df['Age'])
```

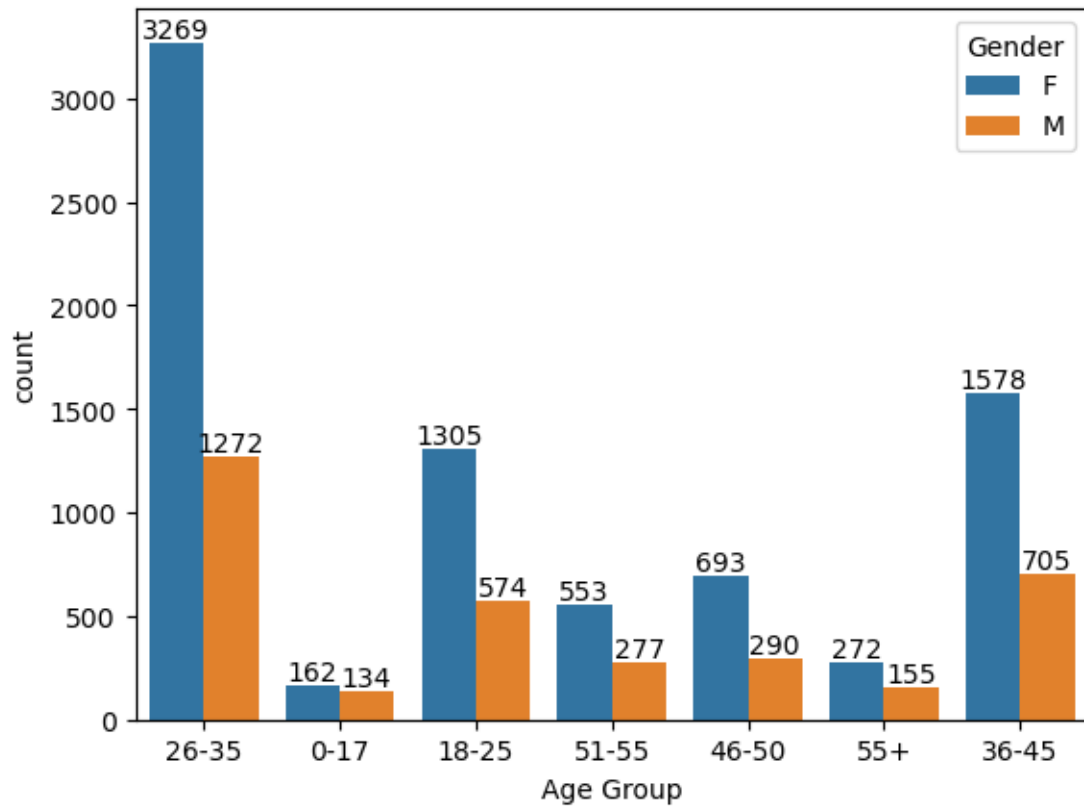
```
[52]: <Axes: ylabel='Age'>
```



```
[63]: df['Age'].unique()
```

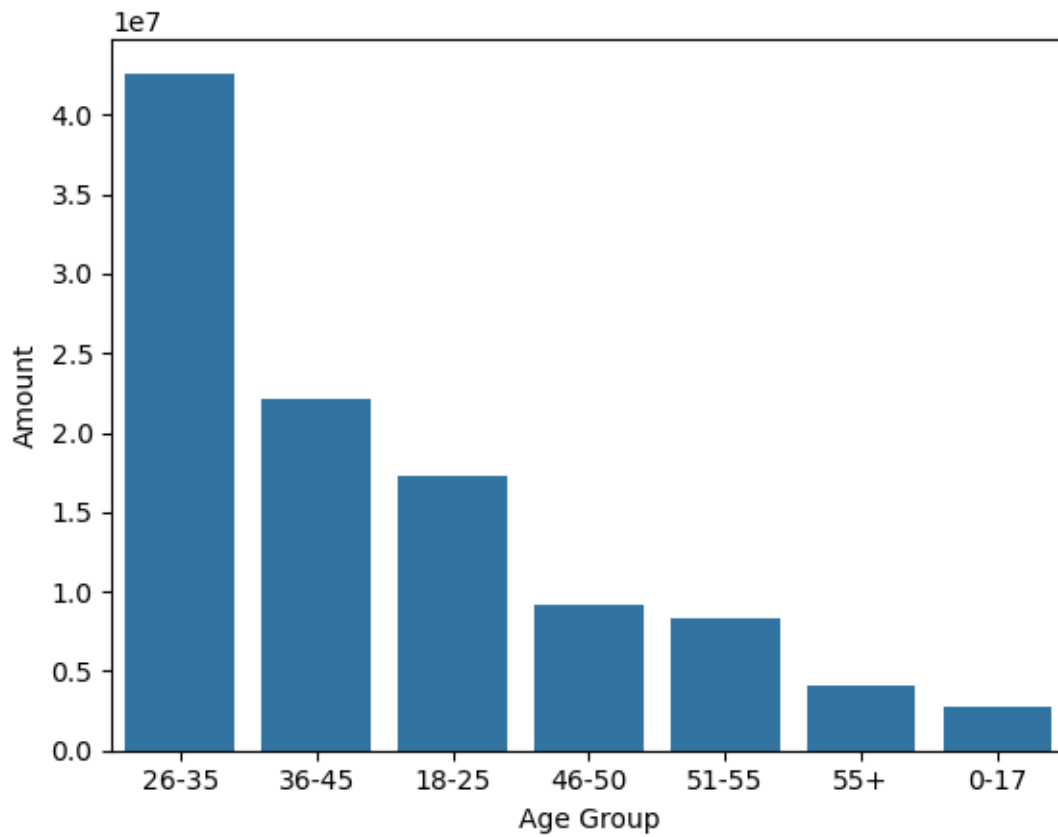
```
[63]: array([28, 35, 16, 25, 26, 34, 20, 24, 29, 54, 19, 46, 30, 53, 83, 33, 40,
          39, 32, 36, 55, 27, 72, 45, 43, 47, 22, 52, 18, 21, 38, 37, 23, 49,
          42, 50, 48, 31, 44, 41, 66, 15, 51, 77, 87, 79, 71, 88, 58, 82, 62,
          92, 12, 63, 17, 13, 67, 90, 56, 75, 81, 64, 73, 84, 14, 76, 86, 89,
          68, 61, 91, 85, 70, 80, 65, 74, 69, 78, 57, 60, 59])
```

```
[66]: counts = sns.countplot(data = df, x = 'Age Group', hue = 'Gender')
      for bars in counts.containers:
          counts.bar_label(bars)
```



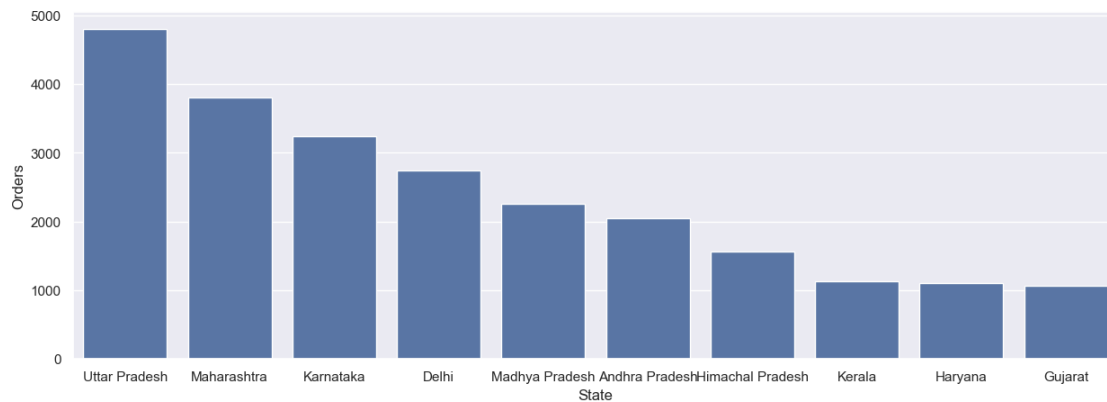
```
[67]: sales_age = df.groupby(['Age Group'], as_index=False)['Amount'].sum().
      ↪sort_values(by='Amount', ascending=False)
      sns.barplot(x = 'Age Group',y= 'Amount' ,data = sales_age)
```

```
[67]: <Axes: xlabel='Age Group', ylabel='Amount'>
```

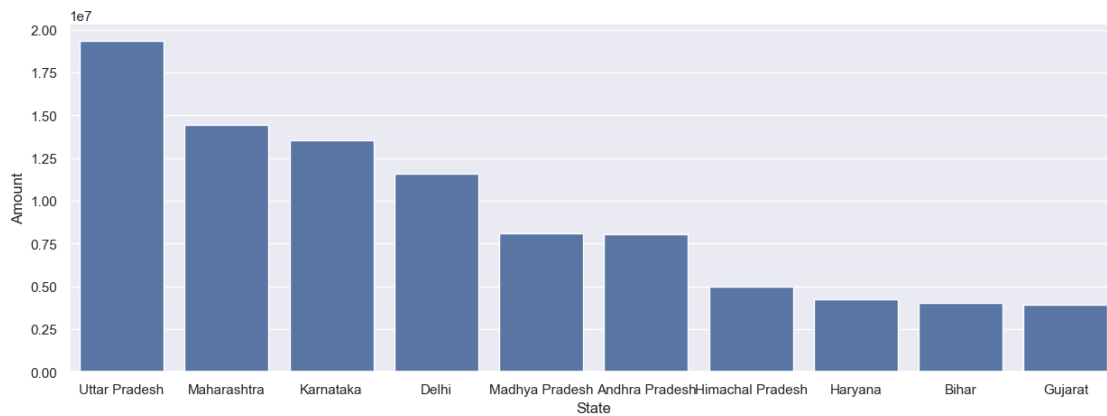
```
[69]: sales_state = df.groupby(['State'], as_index=False)['Orders'].sum().
      ↪sort_values(by='Orders', ascending=False).head(10)
      sns.set(rc={'figure.figsize':(15,5)})
      sns.barplot(data = sales_state, x = 'State',y= 'Orders')
```

```
[69]: <Axes: xlabel='State', ylabel='Orders'>
```



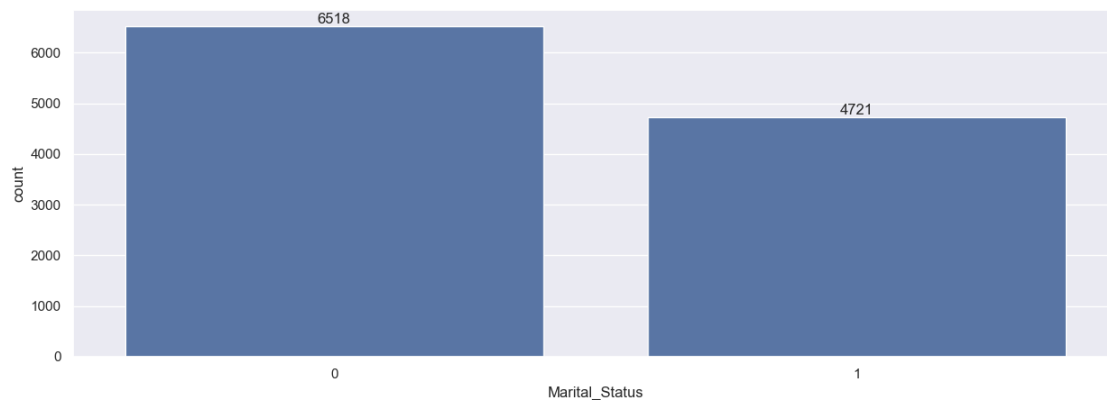
```
[70]: sales_state = df.groupby(['State'], as_index=False)['Amount'].sum().
      ↪sort_values(by='Amount', ascending=False).head(10)
sns.set(rc={'figure.figsize':(15,5)})
sns.barplot(data = sales_state, x = 'State',y= 'Amount')
```

```
[70]: <Axes: xlabel='State', ylabel='Amount'>
```



```
[71]: ax = sns.countplot(data = df, x = 'Marital_Status')

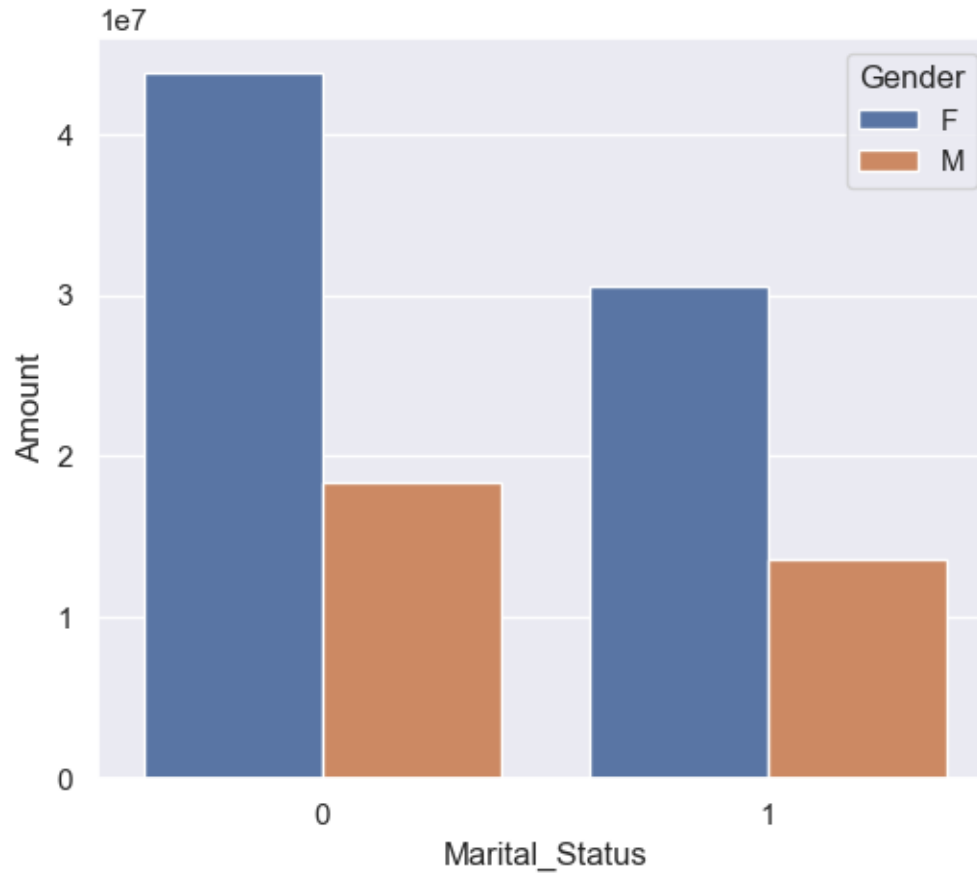
sns.set(rc={'figure.figsize':(7,5)})
for bars in ax.containers:
    ax.bar_label(bars)
```



```
[72]: sales_state = df.groupby(['Marital_Status', 'Gender'],
      ↪as_index=False)['Amount'].sum().sort_values(by='Amount', ascending=False)
```

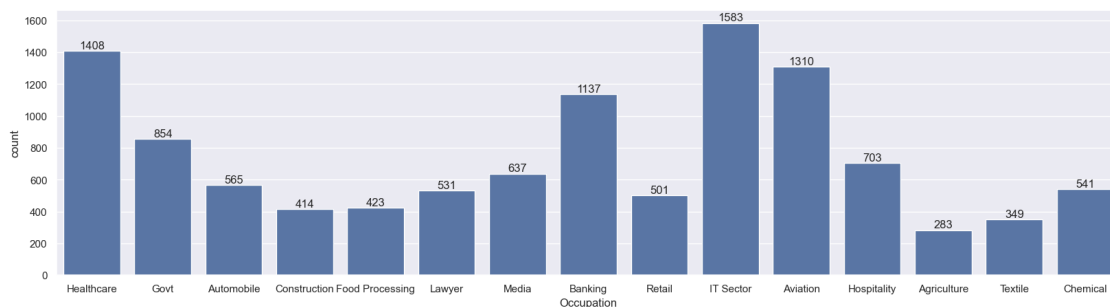
```
sns.set(rc={'figure.figsize':(6,5)})
sns.barplot(data = sales_state, x = 'Marital_Status',y= 'Amount', hue='Gender')
```

[72]: <Axes: xlabel='Marital_Status', ylabel='Amount'>



```
[73]: sns.set(rc={'figure.figsize':(20,5)})
ax = sns.countplot(data = df, x = 'Occupation')

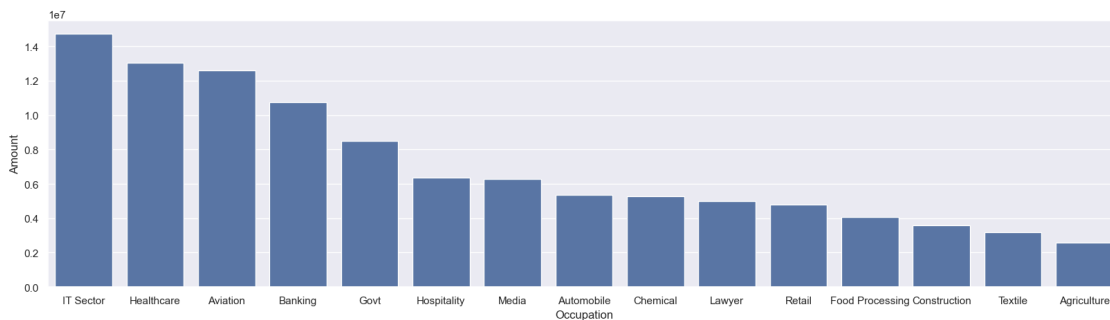
for bars in ax.containers:
    ax.bar_label(bars)
```



```
[74]: sales_state = df.groupby(['Occupation'], as_index=False)['Amount'].sum().
      ↪sort_values(by='Amount', ascending=False)

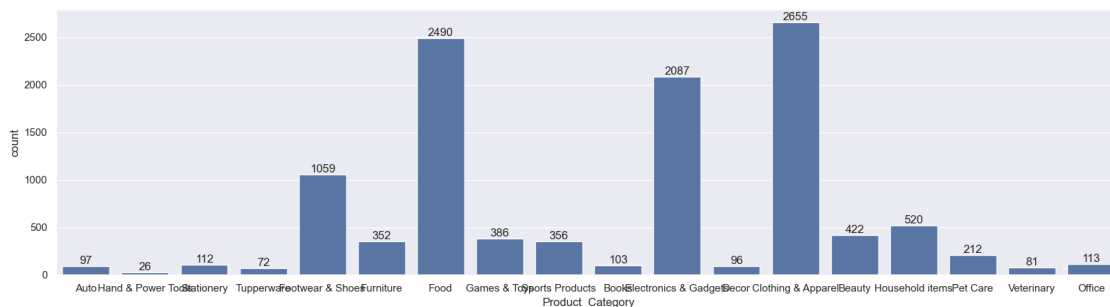
sns.set(rc={'figure.figsize':(20,5)})
sns.barplot(data = sales_state, x = 'Occupation',y= 'Amount')
```

[74]: <Axes: xlabel='Occupation', ylabel='Amount'>



```
[75]: sns.set(rc={'figure.figsize':(20,5)})
ax = sns.countplot(data = df, x = 'Product_Category')

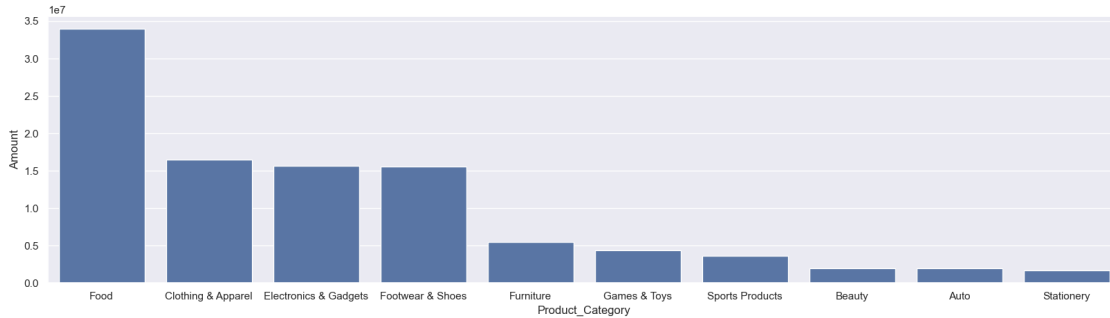
for bars in ax.containers:
    ax.bar_label(bars)
```



```
[76]: sales_state = df.groupby(['Product_Category'], as_index=False)['Amount'].sum().
      ↪sort_values(by='Amount', ascending=False).head(10)

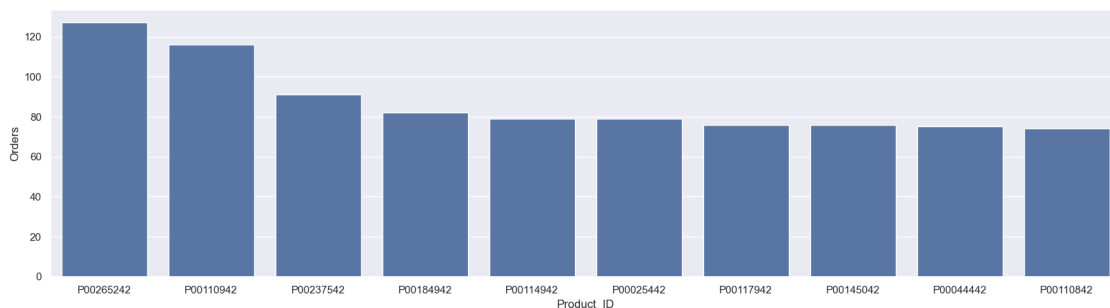
sns.set(rc={'figure.figsize':(20,5)})
sns.barplot(data = sales_state, x = 'Product_Category',y= 'Amount')
```

[76]: <Axes: xlabel='Product_Category', ylabel='Amount'>



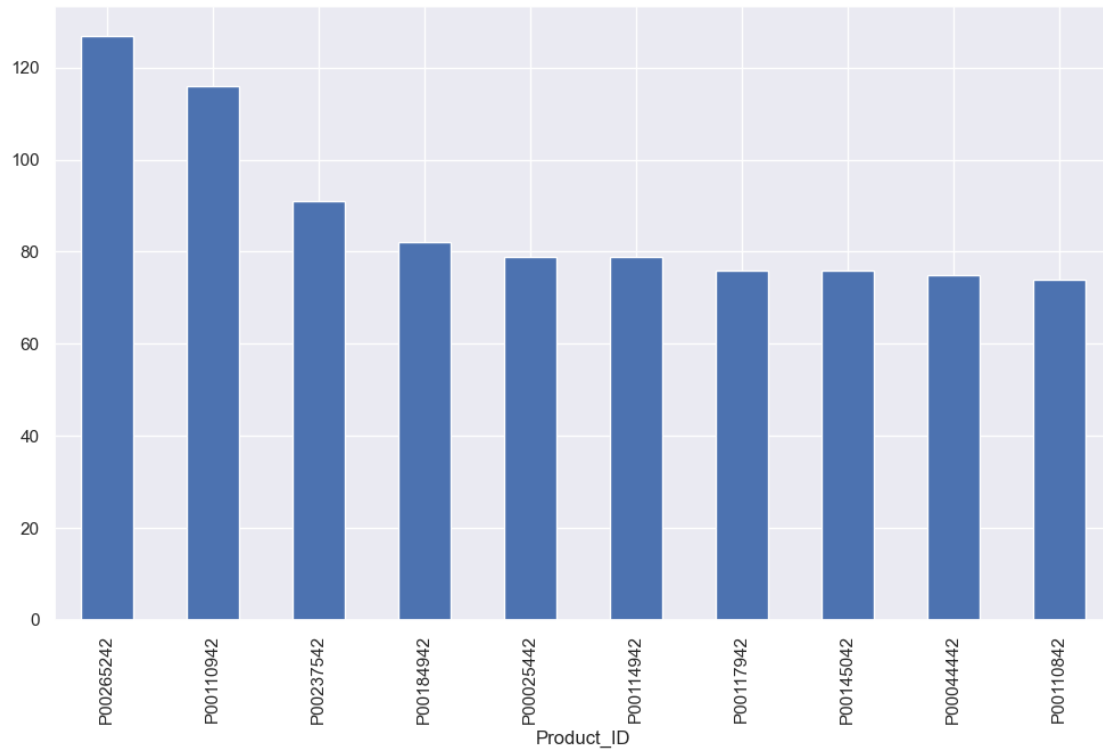
```
[77]: sales_state = df.groupby(['Product_ID'], as_index=False)['Orders'].sum().  
      ↪sort_values(by='Orders', ascending=False).head(10)  
  
sns.set(rc={'figure.figsize':(20,5)})  
sns.barplot(data = sales_state, x = 'Product_ID',y= 'Orders')
```

[77]: <Axes: xlabel='Product_ID', ylabel='Orders'>



```
[78]: fig1, ax1 = plt.subplots(figsize=(12,7))  
df.groupby('Product_ID')['Orders'].sum().nlargest(10).  
  ↪sort_values(ascending=False).plot(kind='bar')
```

[78]: <Axes: xlabel='Product_ID'>



Conclusion:

Married women age group 26-35 yrs from UP, Maharastra and Karnataka working in IT, Healthcare and Aviation are more likely to buy products from Food, Clothing and Electronics category

[]: