

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
df = pd.read_csv('/content/drive/MyDrive/Diwali Sales Data.csv', encoding='unicode_escape')
```

```
df.head(5)
```

| | User_ID | Cust_name | Product_ID | Gender | Age Group | Age | Marital_Status | State | Zone | Occupation | Product_Category | Orders |
|---|---------|-----------|------------|--------|-----------|-----|----------------|----------------|----------|--------------|------------------|--------|
| 0 | 1002903 | Sanskriti | P00125942 | F | 26-35 | 28 | 0 | Maharashtra | Western | Healthcare | Auto | 11251 |
| 1 | 1000732 | Kartik | P00110942 | F | 26-35 | 35 | 1 | Andhra Pradesh | Southern | Govt | Auto | 11251 |
| 2 | 1001990 | Bindu | P00118542 | F | 26-35 | 35 | 1 | Uttar Pradesh | Central | Automobile | Auto | 11251 |
| 3 | 1001425 | Sudevi | P00237842 | M | 0-17 | 16 | 0 | Karnataka | Southern | Construction | Auto | 11251 |

Next steps: [Generate code with df](#) [New interactive sheet](#)

```
df.shape
```

(11251, 15)

```
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
```

```
df.describe()
```

| | User_ID | Age | Marital_Status | Orders | Amount | Status | unnamed1 | |
|-------|--------------|--------------|----------------|--------------|--------------|--------|----------|--|
| count | 1.125100e+04 | 11251.000000 | 11251.000000 | 11251.000000 | 11239.000000 | 0.0 | 0.0 | |
| mean | 1.003004e+06 | 35.421207 | 0.420318 | 2.489290 | 9453.610858 | NaN | NaN | |
| std | 1.716125e+03 | 12.754122 | 0.493632 | 1.115047 | 5222.355869 | NaN | NaN | |
| min | 1.000001e+06 | 12.000000 | 0.000000 | 1.000000 | 188.000000 | NaN | NaN | |
| 25% | 1.001492e+06 | 27.000000 | 0.000000 | 1.500000 | 5443.000000 | NaN | NaN | |
| 50% | 1.003065e+06 | 33.000000 | 0.000000 | 2.000000 | 8109.000000 | NaN | NaN | |
| 75% | 1.004430e+06 | 43.000000 | 1.000000 | 3.000000 | 12675.000000 | NaN | NaN | |
| max | 1.006040e+06 | 92.000000 | 1.000000 | 4.000000 | 23952.000000 | NaN | NaN | |

```
df.drop(['Status', 'unnamed1'], axis=1, inplace=True)
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 11251 entries, 0 to 11250
Data columns (total 13 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
dtypes: float64(3), int64(4), object(6)
memory usage: 1.3+ MB
```



```

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
dtypes: float64(1), int64(4), object(8)
memory usage: 1.1+ MB

```

```
pd.isnull(df)
```

| | User_ID | Cust_name | Product_ID | Gender | Age Group | Age | Marital_Status | State | Zone | Occupation | Product_Category | Orders |
|-------|---------|-----------|------------|--------|-----------|-------|----------------|-------|-------|------------|------------------|--------|
| 0 | False | False | False | False | False | False | False | False | False | False | False | False |
| 1 | False | False | False | False | False | False | False | False | False | False | False | False |
| 2 | False | False | False | False | False | False | False | False | False | False | False | False |
| 3 | False | False | False | False | False | False | False | False | False | False | False | False |
| 4 | False | False | False | False | False | False | False | False | False | False | False | False |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 11246 | False | False | False | False | False | False | False | False | False | False | False | False |
| 11247 | False | False | False | False | False | False | False | False | False | False | False | False |
| 11248 | False | False | False | False | False | False | False | False | False | False | False | False |
| 11249 | False | False | False | False | False | False | False | False | False | False | False | False |
| 11250 | False | False | False | False | False | False | False | False | False | False | False | False |

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

| | 0 |
|------------------|----|
| 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
```

```
df.shape
```

```
(11251, 13)
```

```
# drop null values
df.dropna(inplace=True)
```

```
df.shape
```

```
(11239, 13)
```

```
data_test = [['kapil',23], ['sonu',22],['manish' ]]  
df_test = pd.DataFrame(data_test, columns=['Name','Age'])  
df_test
```

| | Name | Age | |
|---|--------|------|--|
| 0 | kapil | 23.0 | |
| 1 | sonu | 22.0 | |
| 2 | manish | NaN | |

Next steps: [Generate code with df_test](#) [New interactive sheet](#)

```
df_test.dropna()
```

| | Name | Age | |
|---|-------|------|--|
| 0 | kapil | 23.0 | |
| 1 | sonu | 22.0 | |

```
df_test
```

| | Name | Age | |
|---|--------|------|--|
| 0 | kapil | 23.0 | |
| 1 | sonu | 22.0 | |
| 2 | manish | NaN | |

Next steps: [Generate code with df_test](#) [New interactive sheet](#)

```
# change data type  
df['Amount'] = df['Amount'].astype('int')
```

```
df['Amount'].dtype
```

```
dtype('int64')
```

```
df.columns
```

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

```
# Rename column  
df.rename(columns={'Marital_Status':'Shaadi'})
```

| | User_ID | Cust_name | Product_ID | Gender | Age Group | Age | Shaadi | State | Zone | Occupation | Product_Category | Order |
|-------|---------|-------------|------------|--------|-----------|-----|--------|----------------|----------|-----------------|------------------|------------|
| 0 | 1002903 | Sanskriti | P00125942 | F | 26-35 | 28 | 0 | Maharashtra | Western | Healthcare | | Auto |
| 1 | 1000732 | Kartik | P00110942 | F | 26-35 | 35 | 1 | Andhra Pradesh | Southern | Govt | | Auto |
| 2 | 1001990 | Bindu | P00118542 | F | 26-35 | 35 | 1 | Uttar Pradesh | Central | Automobile | | Auto |
| 3 | 1001425 | Sudevi | P00237842 | M | 0-17 | 16 | 0 | Karnataka | Southern | Construction | | Auto |
| 4 | 1000588 | Joni | P00057942 | M | 26-35 | 28 | 1 | Gujarat | Western | Food Processing | | Auto |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | | ... |
| 11246 | 1000695 | Manning | P00296942 | M | 18-25 | 19 | 1 | Maharashtra | Western | Chemical | | Office |
| 11247 | 1004089 | Reichenbach | P00171342 | M | 26-35 | 33 | 0 | Haryana | Northern | Healthcare | | Veterinary |
| 11248 | 1001209 | Oshin | P00201342 | F | 36-45 | 40 | 0 | Madhya Pradesh | Central | Textile | | Office |
| 11249 | 1004023 | Noonan | P00059442 | M | 36-45 | 37 | 0 | Karnataka | Southern | Agriculture | | Office |

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
Index: 11239 entries, 0 to 11250  
Data columns (total 13 columns):
```

```
# Column      Non-Null Count  Dtype
---  -
0    User_ID      11239 non-null    int64
1    Cust_name     11239 non-null    object
2    Product_ID    11239 non-null    object
3    Gender        11239 non-null    object
4    Age Group     11239 non-null    object
5    Age           11239 non-null    int64
6    Marital_Status 11239 non-null    int64
7    State         11239 non-null    object
8    Zone          11239 non-null    object
9    Occupation    11239 non-null    object
10   Product_Category 11239 non-null    object
11   Orders        11239 non-null    int64
12   Amount        11239 non-null    int64
dtypes: int64(5), object(8)
memory usage: 1.2+ MB
```

```
# describe() method return description of the data in dataframe(i.e count,mean,std etc)
df.describe()
```

| | User_ID | Age | Marital_Status | Orders | Amount |
|--------------|--------------|--------------|----------------|--------------|--------------|
| count | 1.123900e+04 | 11239.000000 | 11239.000000 | 11239.000000 | 11239.000000 |
| mean | 1.003004e+06 | 35.410357 | 0.420055 | 2.489634 | 9453.610553 |
| std | 1.716039e+03 | 12.753866 | 0.493589 | 1.114967 | 5222.355168 |
| min | 1.000001e+06 | 12.000000 | 0.000000 | 1.000000 | 188.000000 |
| 25% | 1.001492e+06 | 27.000000 | 0.000000 | 2.000000 | 5443.000000 |
| 50% | 1.003064e+06 | 33.000000 | 0.000000 | 2.000000 | 8109.000000 |
| 75% | 1.004426e+06 | 43.000000 | 1.000000 | 3.000000 | 12675.000000 |
| max | 1.006040e+06 | 92.000000 | 1.000000 | 4.000000 | 23952.000000 |

```
# use describe() for specific columns
df[['Age','Orders','Amount']].describe()
```

| | Age | Orders | Amount |
|--------------|--------------|--------------|--------------|
| count | 11239.000000 | 11239.000000 | 11239.000000 |
| mean | 35.410357 | 2.489634 | 9453.610553 |
| std | 12.753866 | 1.114967 | 5222.355168 |
| min | 12.000000 | 1.000000 | 188.000000 |
| 25% | 27.000000 | 2.000000 | 5443.000000 |
| 50% | 33.000000 | 2.000000 | 8109.000000 |
| 75% | 43.000000 | 3.000000 | 12675.000000 |
| max | 92.000000 | 4.000000 | 23952.000000 |

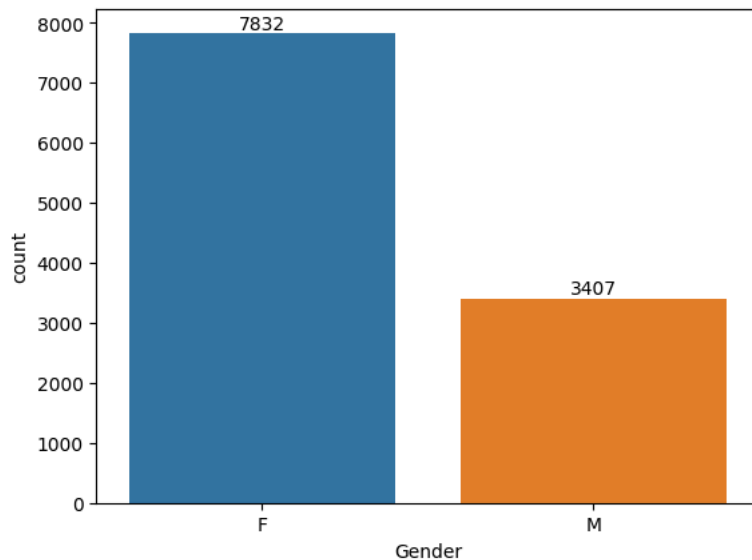
Expolaratory Data analysis

Gender

```
df.columns
```

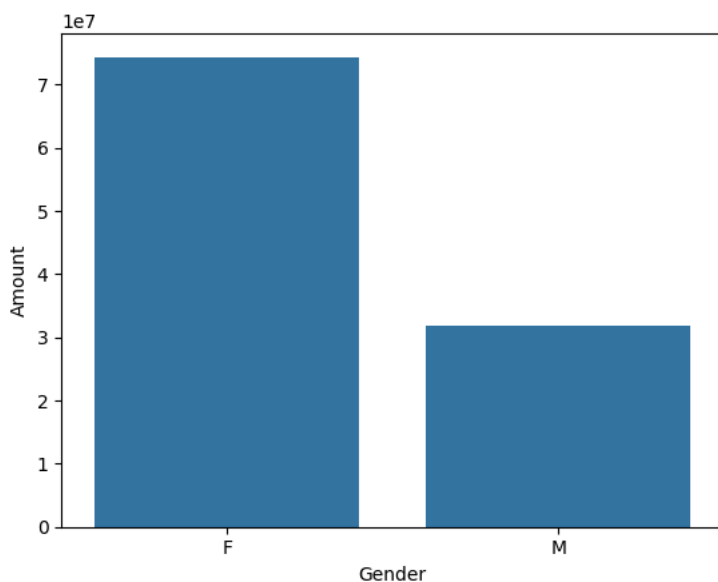
```
Index(['User_ID', 'Cust_name', 'Product_ID', 'Gender', 'Age Group', 'Age',
      'Marital_Status', 'State', 'Zone', 'Occupation', 'Product_Category',
      'Orders', 'Amount'],
      dtype='object')
```

```
ax = sns.countplot(x='Gender', data=df , hue = 'Gender')
for bars in ax.containers:
    ax.bar_label(bars)
```



```
sales_gen = df.groupby(['Gender'], as_index=False)['Amount'].sum().sort_values(by='Amount', ascending=False)
sns.barplot(x='Gender', y='Amount', data=sales_gen)
```

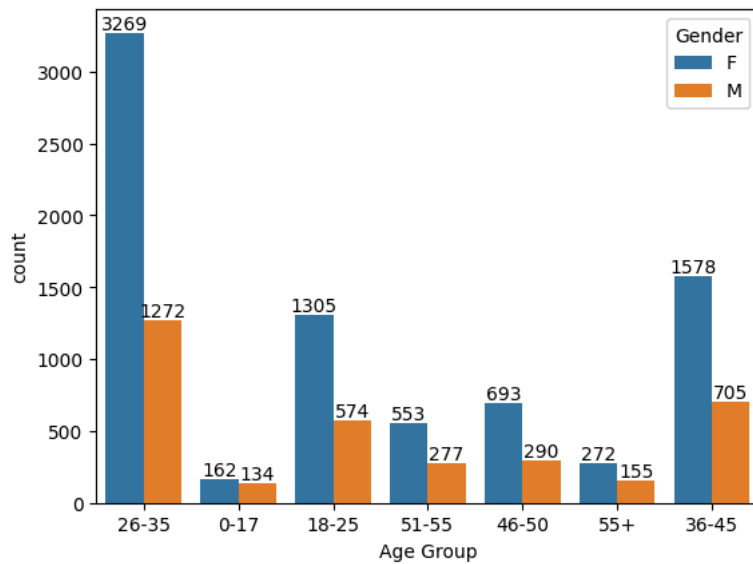
<Axes: xlabel='Gender', ylabel='Amount'>



from above graphs we can see that most of the buyers are female and even the power of purchasing power of the females are the greater then men

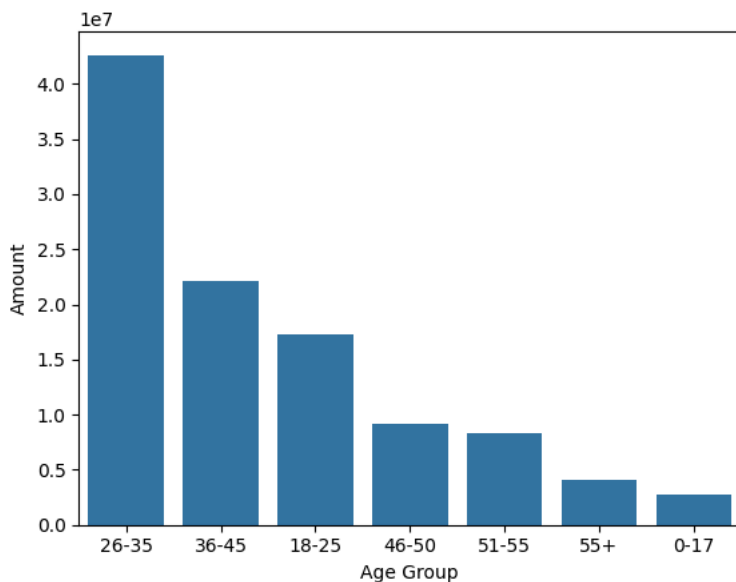
✓ Age

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



```
# total amount vs age group
sales_gen = df.groupby(['Age Group'], as_index=False)['Amount'].sum().sort_values(by='Amount', ascending=False)
sns.barplot(x='Age Group', y='Amount', data=sales_gen)
```

<Axes: xlabel='Age Group', ylabel='Amount'>



from above graph we can see that the most of buyers are of age group between 26-35yrs female

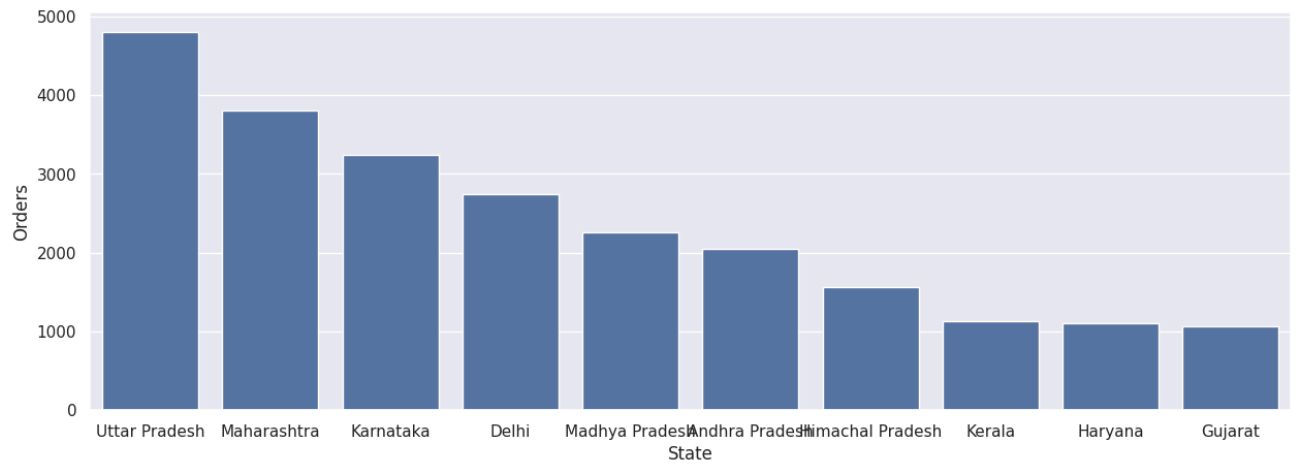
state

```
df.columns
```

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

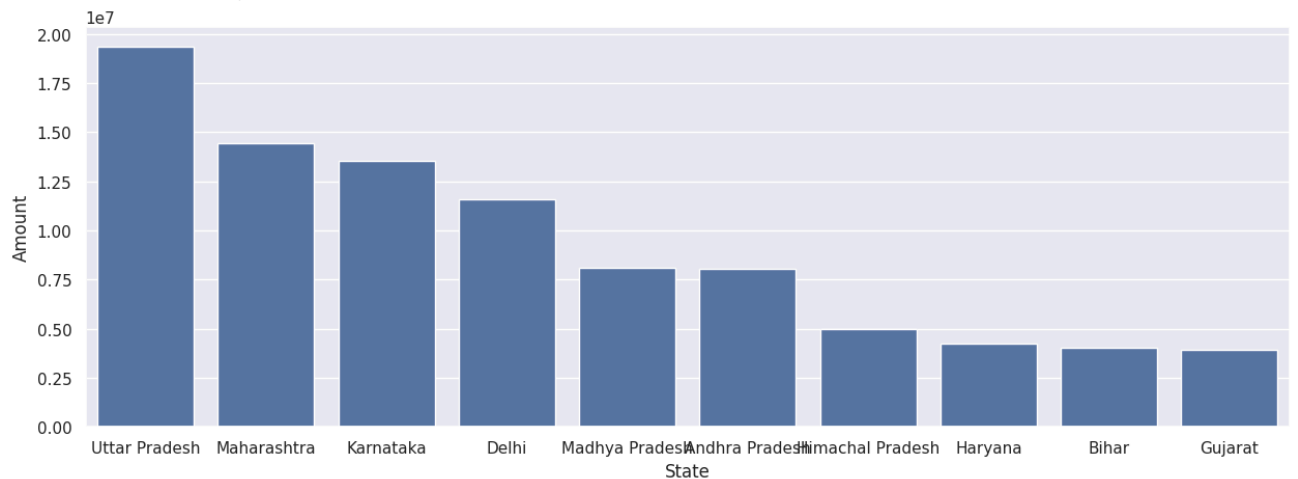
```
# total number of order from top 10 state
sales_state = df.groupby(['State'], as_index=False)['Orders'].sum().sort_values(by='Orders', ascending=False).head(10)
sns.set(rc={'figure.figsize':(15,5)})
sns.barplot(x='State', y='Orders', data=sales_state)
```

<Axes: xlabel='State', ylabel='Orders'>



```
#total amount/sales top 10 state
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(x='State', y='Amount', data=sales_state)
```

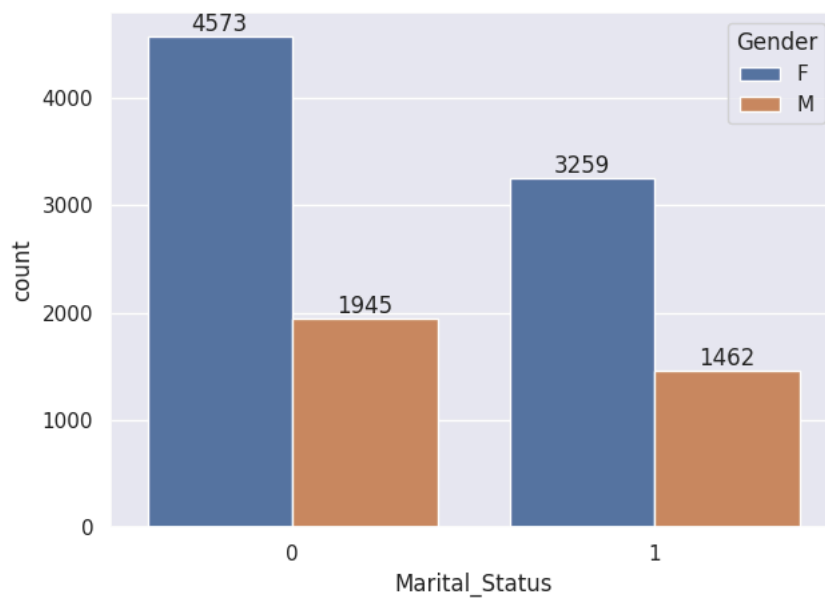
<Axes: xlabel='State', ylabel='Amount'>



from above graph we can see that unexpectedly most of the orders are from uttar pradesh maharashtra and karnataka respectively but total sales/amount from up karnataka and then maharashtra

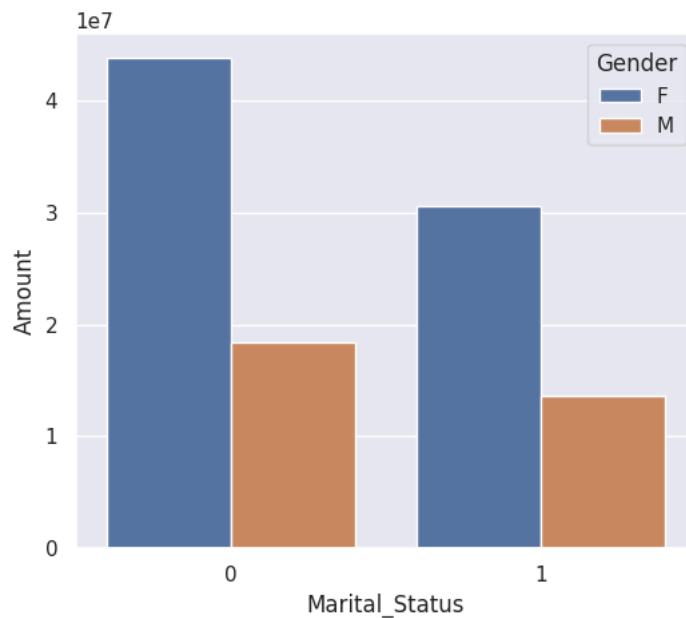
✓ Marital Status

```
ax = sns.countplot(data=df, x='Marital_Status', hue = 'Gender')
sns.set(rc={'figure.figsize':(7,5)})
for bars in ax.containers:
    ax.bar_label(bars)
```



```
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')
```

<Axes: xlabel='Marital_Status', ylabel='Amount'>



from above graphs we can see that most of the buyers are married (women) and they have purchasing power

✓ occupation

```
# Set figure size
sns.set(rc={'figure.figsize':(20,5)})

# Countplot with colorful bars
ax = sns.countplot(data=df, x='Occupation', palette="Set3")

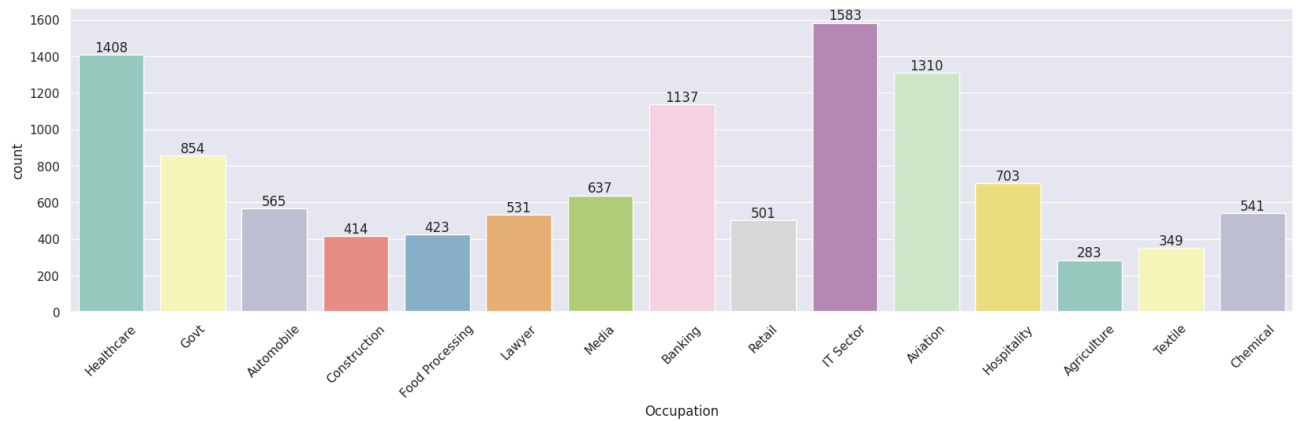
# Add labels on top of bars
for bars in ax.containers:
    ax.bar_label(bars)

plt.xticks(rotation=45)
plt.show()
```


/tmp/ipython-input-3654508151.py:5: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set

```
ax = sns.countplot(data=df, x='Occupation', palette="Set3")
```



```
# Grouping occupation-wise total sales amount
```

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

```
sns.set(rc={'figure.figsize':(20,5)})
```

```
# Colorful barplot
```

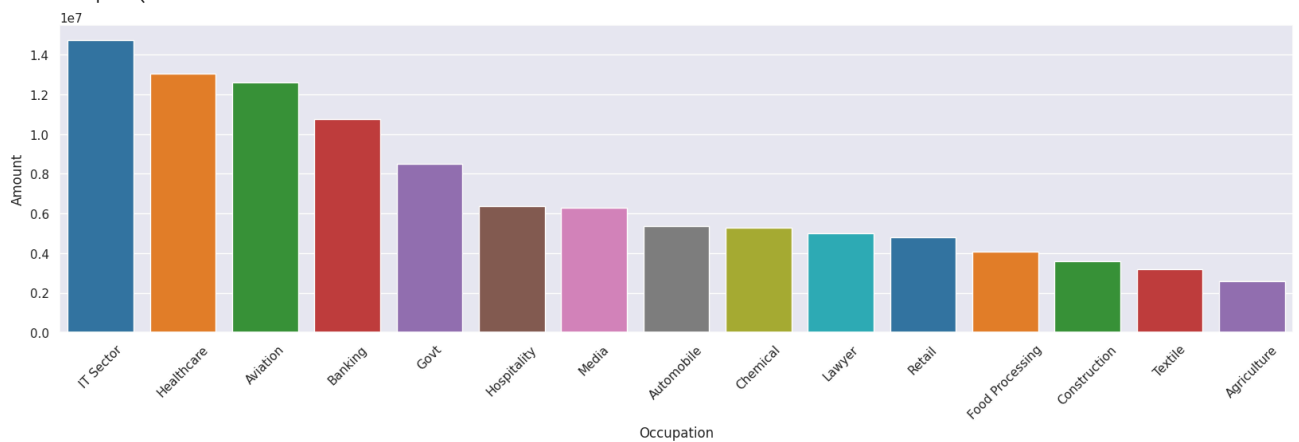
```
sns.barplot(
    data=sales_state,
    x='Occupation',
    y='Amount',
    palette='tab10' #palette change karne se alag-alag colors aayenge
)
```

```
plt.xticks(rotation=45) # x-labels readable banane ke liye
plt.show()
```

/tmp/ipython-input-1092369730.py:8: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set

```
sns.barplot(
```



from above graphs we can see that most of the buyers are working in it ,Aviation and healthcare sector

✓ Product category

```
sns.set(rc={'figure.figsize':(20,5)})

ax = sns.countplot(data=df, x='Product_Category', palette="Set2")

# Add count labels
for bars in ax.containers:
    ax.bar_label(bars)

plt.xticks(rotation=45)
plt.show()
```

/tmp/ipython-input-2892381369.py:3: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set

