

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

# importing dataset
data = pd.read_csv("Diwali Sales Data.csv", encoding='unicode_escape')

# top 5 and bottom 5 rows of dataset
data.head(5)
```

	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

	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

	Amount	Status	unnamed1
0	23952.0	NaN	NaN
1	23934.0	NaN	NaN
2	23924.0	NaN	NaN
3	23912.0	NaN	NaN
4	23877.0	NaN	NaN

```
data.tail(5)
```

	User_ID	Cust_name	Product_ID	Gender	Age Group	Age
11246	1000695	Manning	P00296942	M	18-25	19

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						

Amount	State	Zone	Occupation	Product_Category	Orders
11246	Maharashtra	Western	Chemical	Office	4
370.0					
11247	Haryana	Northern	Healthcare	Veterinary	3
367.0					
11248	Madhya Pradesh	Central	Textile	Office	4
213.0					
11249	Karnataka	Southern	Agriculture	Office	3
206.0					
11250	Maharashtra	Western	Healthcare	Office	3
188.0					

	Status	unnamed1
11246	NaN	NaN
11247	NaN	NaN
11248	NaN	NaN
11249	NaN	NaN
11250	NaN	NaN

no of rows and columns

data.shape

```
print(f'Number of rows in dataset:{data.shape[0]}')
```

```
print(f'Number of columns in dataset:{data.shape[1]}')
```

Number of rows in dataset:11251

Number of columns in dataset:15

Display columns

data.columns

```
Index(['User_ID', 'Cust_name', 'Product_ID', 'Gender', 'Age Group',
      'Age',
      'Marital_Status', 'State', 'Zone', 'Occupation',
      'Product_Category',
      'Orders', 'Amount', 'Status', 'unnamed1'],
      dtype='object')
```

stats about dataset

```
data.describe(include='all')
```

		User_ID	Cust_name	Product_ID	Gender	Age	Group
Age \	count	1.125100e+04	11251	11251	11251	11251	11251
	11251.000000						
	unique	NaN	1250	2351	2	7	
	NaN						
	top	NaN	Vishakha	P00265242	F	26-35	
	NaN						
	freq	NaN	42	53	7842	4543	
	NaN						
	mean	1.003004e+06	NaN	NaN	NaN	NaN	
	35.421207						
	std	1.716125e+03	NaN	NaN	NaN	NaN	
	12.754122						
	min	1.000001e+06	NaN	NaN	NaN	NaN	
	12.000000						
	25%	1.001492e+06	NaN	NaN	NaN	NaN	
	27.000000						
	50%	1.003065e+06	NaN	NaN	NaN	NaN	
	33.000000						
	75%	1.004430e+06	NaN	NaN	NaN	NaN	
	43.000000						
	max	1.006040e+06	NaN	NaN	NaN	NaN	
	92.000000						
		Marital_Status	State	Zone	Occupation		
Product_Category \	count	11251.000000	11251	11251	11251	11251	
	11251						
	unique	NaN	16	5	15		
	18						
	top	NaN	Uttar Pradesh	Central	IT Sector	Clothing &	
	Apparel						
	freq	NaN	1946	4296	1588		
	2655						
	mean	0.420318	NaN	NaN	NaN		
	NaN						
	std	0.493632	NaN	NaN	NaN		
	NaN						
	min	0.000000	NaN	NaN	NaN		
	NaN						
	25%	0.000000	NaN	NaN	NaN		
	NaN						
	50%	0.000000	NaN	NaN	NaN		
	NaN						
	75%	1.000000	NaN	NaN	NaN		
	NaN						
	max	1.000000	NaN	NaN	NaN		
	NaN						

	Orders	Amount	Status	unnamed1
count	11251.000000	11239.000000	0.0	0.0
unique	NaN	NaN	NaN	NaN
top	NaN	NaN	NaN	NaN
freq	NaN	NaN	NaN	NaN
mean	2.489290	9453.610858	NaN	NaN
std	1.115047	5222.355869	NaN	NaN
min	1.000000	188.000000	NaN	NaN
25%	1.500000	5443.000000	NaN	NaN
50%	2.000000	8109.000000	NaN	NaN
75%	3.000000	12675.000000	NaN	NaN
max	4.000000	23952.000000	NaN	NaN

information about dataset

data.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

checking if any column is null

data.isnull().sum()

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
Status        11251
unnamed1      11251
dtype: int64

# handling null vaues
average = data.Amount.mean()
average

np.float64(9453.610857727557)

data.Amount.fillna(round(average,0),inplace=True)

# remove Status and unnamed1 columns
data.drop(columns=['Status','unnamed1'],inplace=True)

data.columns

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

# check whether duplicate values or not
data.duplicated().sum()

np.int64(8)

# handling duplicate values
data[data.duplicated()==True]

```

	User_ID	Cust_name	Product_ID	Gender	Age Group	Age
Marital_Status \						
19	1001883	Praneet	P00029842	M	51-55	54
1						
4404	1004725	Jackson	P00150842	F	36-45	37
1						
5703	1003208	Bowman	P00171642	F	26-35	31
0						
5908	1001260	Dheeraj	P00344042	M	26-35	28
0						
6173	1001325	Reese	P00111742	F	26-35	27
1						
8651	1000083	Gute	P00242842	M	26-35	35


```

...
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      23952.0
1      23934.0
2      23924.0
3      23912.0
4      23877.0
...      ...
11246      370.0
11247      367.0
11248      213.0
11249      206.0
11250      188.0

```

```
[11243 rows x 13 columns]
```

```
# datatypes of columns
```

```
data.dtypes
```

```
User_ID          int64
Cust_name        object
Product_ID       object
Gender           object
Age Group        object
Age              int64
Marital_Status   int64
State            object
Zone             object
Occupation       object
Product_Category object
Orders           int64
Amount           float64
dtype: object
```

```
# Amount datatype convert from float64 to int64
```

```
data.Amount=data.Amount.astype('int64')
```

```
data.dtypes
```

```
User_ID          int64
Cust_name        object
Product_ID       object
Gender           object
Age Group        object
Age              int64
Marital_Status   int64
State            object
Zone             object
Occupation       object
Product_Category object
Orders           int64
Amount           int64
dtype: object
```

```
# rename columns
```

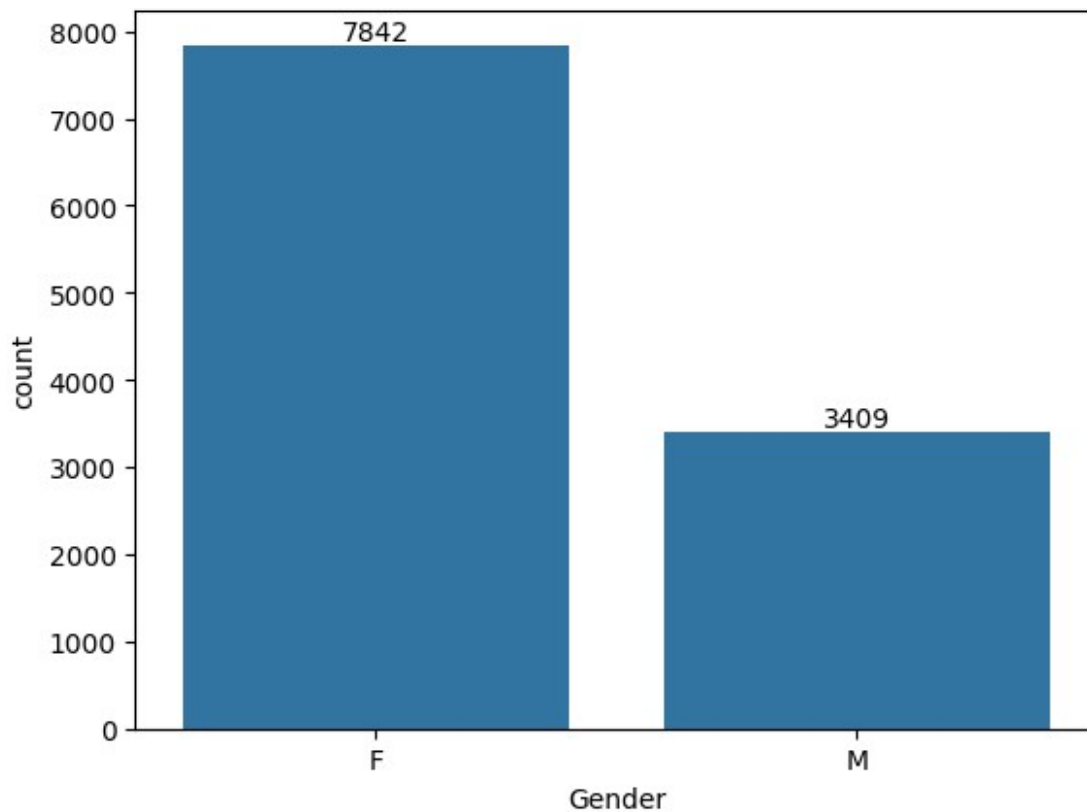
```
data.rename(columns={'Marital_Status':'Married_Or_Not'},inplace=True)
data.columns
```

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


Exploratory Data Analysis

Gender

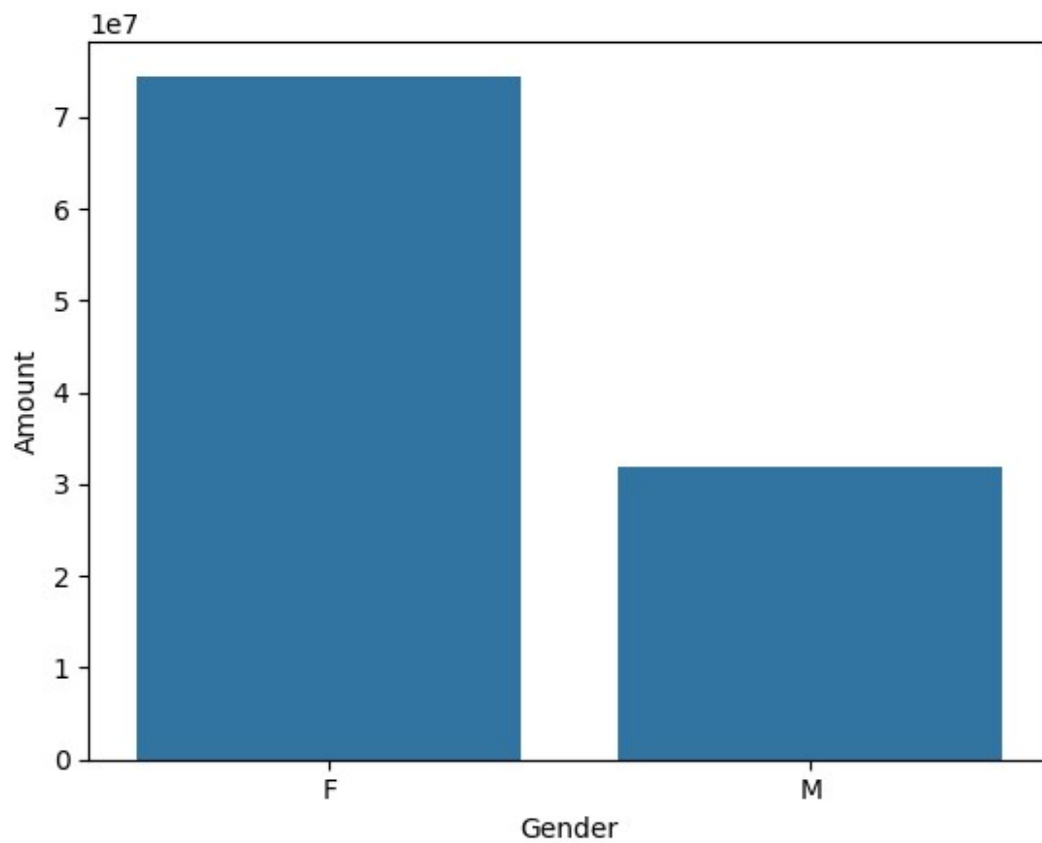
```
count_plot = sns.countplot(x='Gender',data=data,stat='count')
for bars in count_plot.containers:
    count_plot.bar_label(bars)
```



```
# genderwise amount
am =data.groupby(['Gender'],as_index=False)
['Amount'].sum().sort_values(by='Amount',ascending=False)
am
```

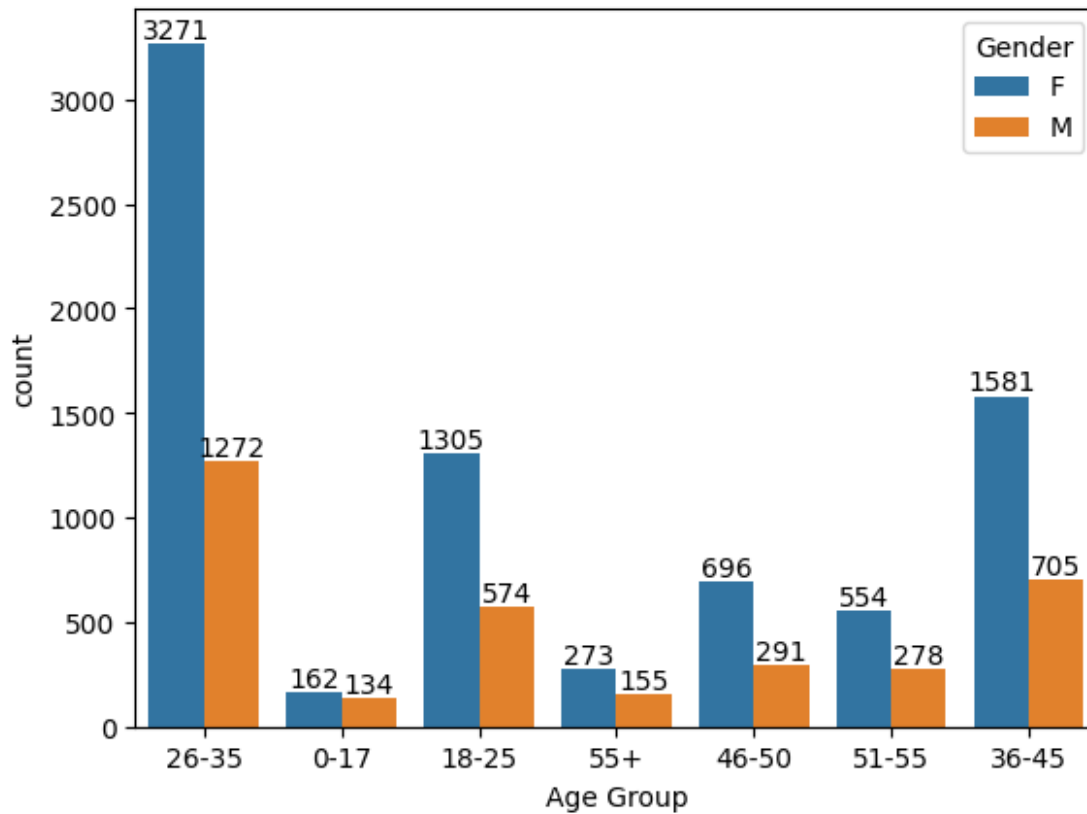
	Gender	Amount
0	F	74430393
1	M	31932184

```
sns.barplot(data=am,x= 'Gender',y='Amount')
<Axes: xlabel='Gender', ylabel='Amount'>
```



Age_Group

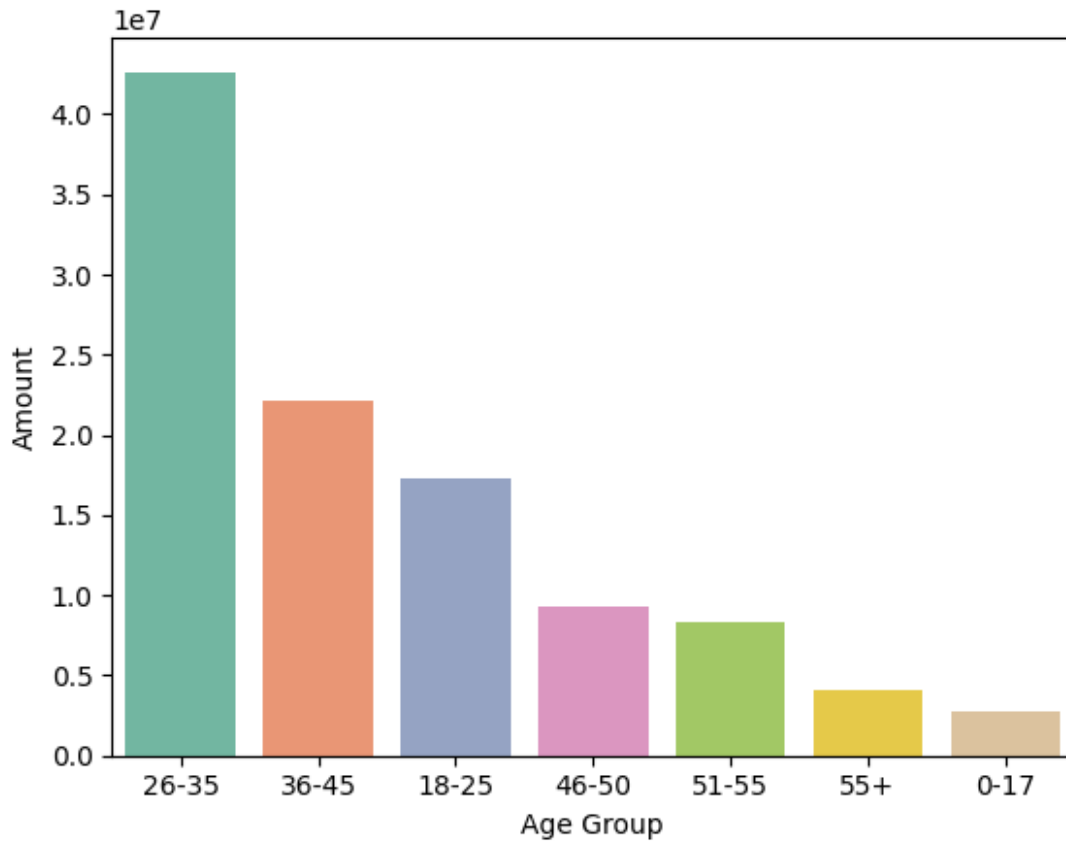
```
age = sns.countplot(data = data, x = 'Age Group', hue = 'Gender')  
for bars in age.containers:  
    age.bar_label(bars)
```



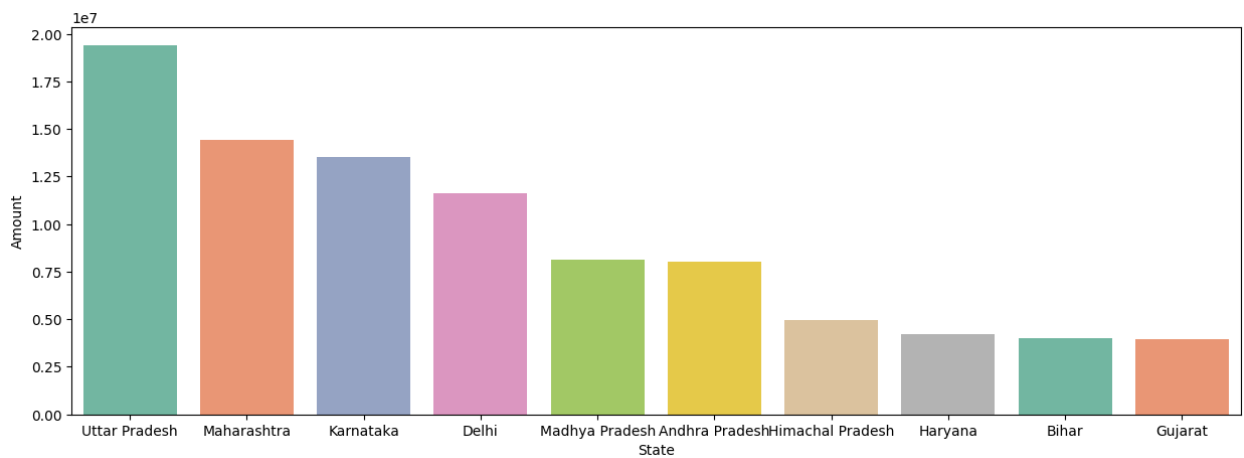
```
import warnings
warnings.filterwarnings('ignore')

# age_group wise amount
aml = data.groupby(['Age Group'],as_index=False)
['Amount'].sum().sort_values(by='Amount',ascending=False)
sns.barplot(data=aml,x='Age Group',y= 'Amount', palette='Set2')

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



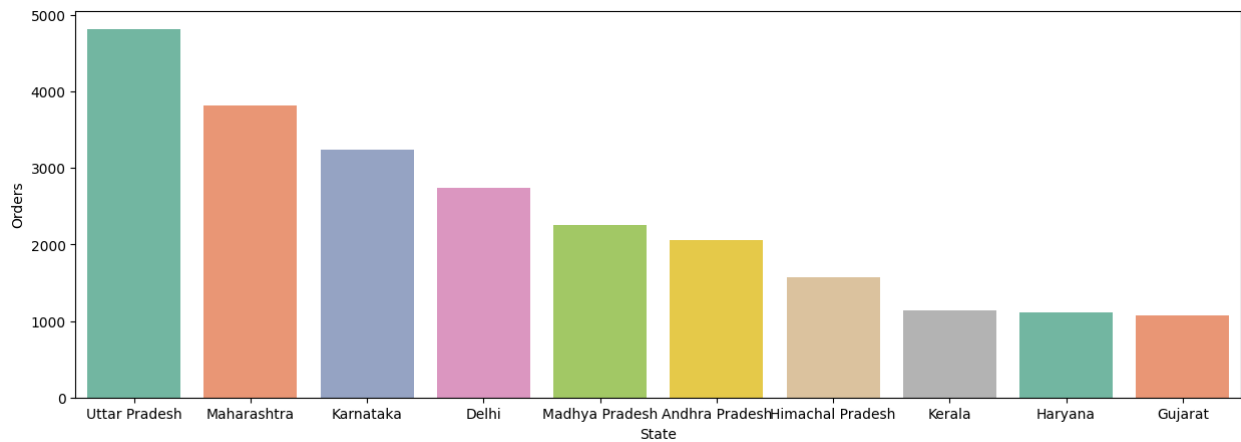
```
# top 10 states-wise total amount
states = data.groupby(['State'], as_index=False)
['Amount'].sum().sort_values(by= 'Amount', ascending=False).head(10)
plt.figure(figsize=(15,5))
sns.barplot(data= states, x='State', y='Amount', palette='Set2')
plt.show()
```



```

# top 10 state wise orders
# top 10 states
states = data.groupby(['State'], as_index=False)
['Orders'].sum().sort_values(by= 'Orders', ascending=False).head(10)
plt.figure(figsize=(15,5))
sns.barplot(data= states, x='State', y='Orders', palette='Set2')
plt.show()

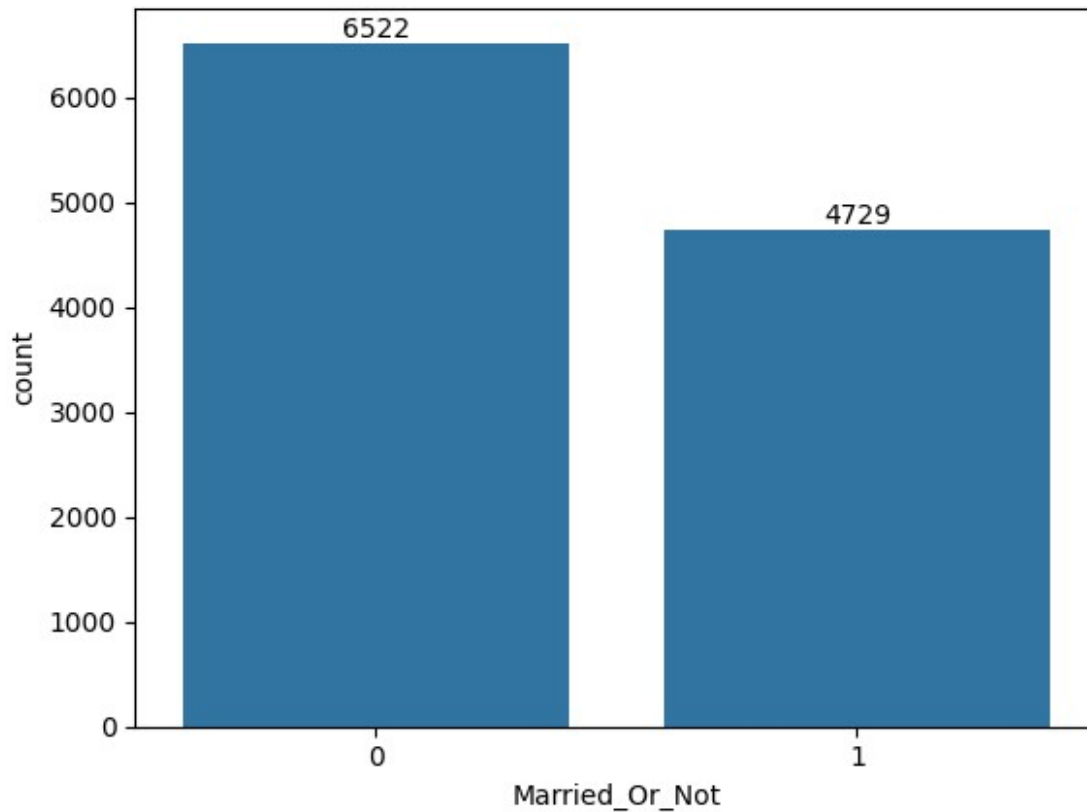
```



```

married = sns.countplot(data = data, x = 'Married_Or_Not')
plt.figure(figsize=(4,5))
for bars in married.containers:
    married.bar_label(bars)

```



<Figure size 400x500 with 0 Axes>

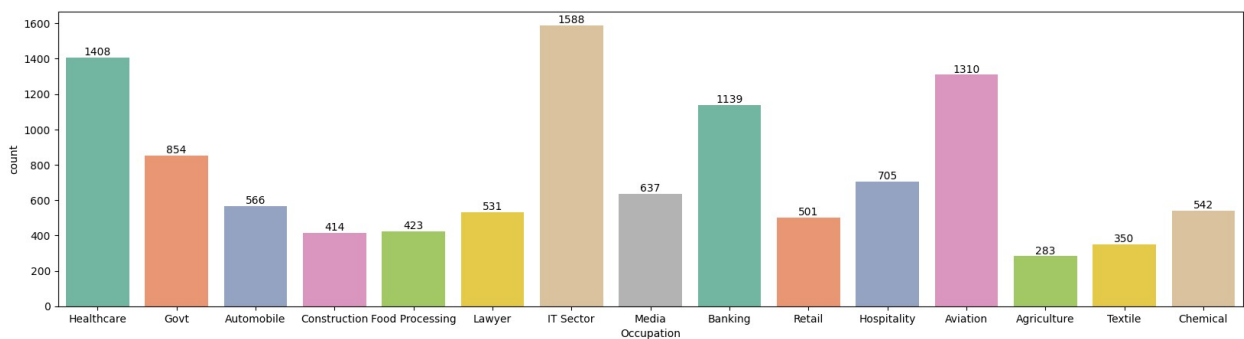
```
data.Married_Or_Not.unique()
```

```
array([0, 1])
```

Occupation

```
plt.figure(figsize=(20,5))  
ax = sns.countplot(data = data, x = 'Occupation',palette='Set2')
```

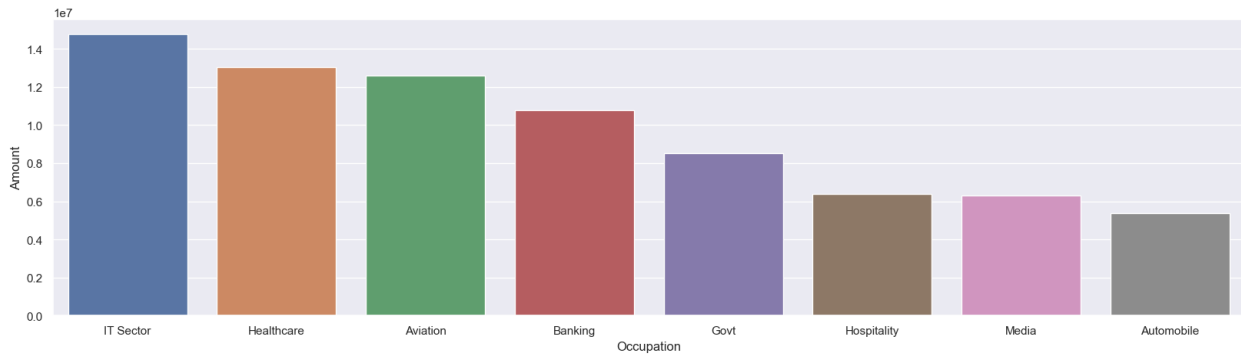
```
for bars in ax.containers:  
    ax.bar_label(bars)
```



```
# top 8 performing occupation-wise total amount
top_states = data.groupby(['Occupation'], as_index=False)
['Amount'].sum().sort_values(by='Amount', ascending=False).head(8)

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

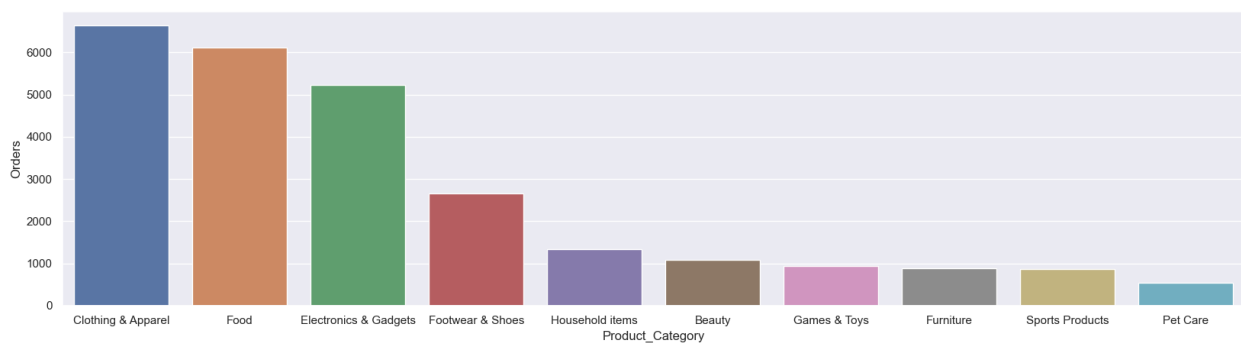
<Axes: xlabel='Occupation', ylabel='Amount'>
```



```
# top 10 most sold products
top_products = data.groupby(['Product_Category'], as_index=False)
['Orders'].sum().sort_values(by='Orders', ascending=False).head(10)

sns.set(rc={'figure.figsize':(20,5)})
sns.barplot(data = top_products, x = 'Product_Category',y=
'Orders',palette='deep')

<Axes: xlabel='Product_Category', ylabel='Orders'>
```



Conclusion

During Dwali Sales, 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