

DIWALI SALES ANALYSIS

```
# import python libraries
```

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

```
# import csv file
```

```
df = pd.read_csv(r"C:\Diwali Sales Data.csv",encoding=
'unicode_escape')
```

```
df
```

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

	State	Zone	Occupation	Product_Category
0	Maharashtra	Western	Healthcare	Auto
1	Andhra Pradesh	Southern	Govt	Auto
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  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
...
11246      370.0    NaN      NaN
11247      367.0    NaN      NaN
11248      213.0    NaN      NaN
11249      206.0    NaN      NaN
11250      188.0    NaN      NaN

```

[11251 rows x 15 columns]

df.head()

```

      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

df.tail()

	User_ID	Cust_name	Product_ID	Gender	Age Group	Age
Marital_Status \						
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
Amount \					
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
```

```
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.shape
```

```
(11251, 15)
```

```
#drop unrelated/blank columns
```

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

```
#check for null values
```

```
pd.isnull(df).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

dtype: int64

drop null values

df.dropna(inplace=True)

df

	User_ID	Cust_name	Product_ID	Gender	Age Group	Age
0	1002903	Sanskriti	P00125942	F	26-35	28
1	1000732	Kartik	P00110942	F	26-35	35
2	1001990	Bindu	P00118542	F	26-35	35
3	1001425	Sudevi	P00237842	M	0-17	16
4	1000588	Joni	P00057942	M	26-35	28
...

11246	1000695	Manning	P00296942	M	18-25	19
11247	1004089	Reichenbach	P00171342	M	26-35	33
11248	1001209	Oshin	P00201342	F	36-45	40
11249	1004023	Noonan	P00059442	M	36-45	37
11250	1002744	Brumley	P00281742	F	18-25	19

	State	Zone	Occupation	Product_Category
0	Maharashtra	Western	Healthcare	Auto
1	Andhra Pradesh	Southern	Govt	Auto
2	Uttar Pradesh	Central	Automobile	Auto
3	Karnataka	Southern	Construction	Auto
4	Gujarat	Western	Food Processing	Auto
...
11246	Maharashtra	Western	Chemical	Office
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

```

```
[11239 rows x 13 columns]
```

```
# change data type
```

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

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

```
dtype('int32')
```

```
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 \							
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

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

...
...
11246	Maharashtra	Western	Chemical	Office
4	Haryana	Northern	Healthcare	Veterinary
3	Madhya Pradesh	Central	Textile	Office
11248	Karnataka	Southern	Agriculture	Office
4	Maharashtra	Western	Healthcare	Office
11249
3
11250
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]
```

```
# describe() method returns description of the data in the 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
mean	1.003004e+06	35.410357	0.420055	2.489634
std	1.716039e+03	12.753866	0.493589	1.114967
min	1.000001e+06	12.000000	0.000000	1.000000
25%	1.001492e+06	27.000000	0.000000	2.000000
50%	1.003064e+06	33.000000	0.000000	2.000000
75%	1.004426e+06	43.000000	1.000000	3.000000
max	1.006040e+06	92.000000	1.000000	4.000000

```
# 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

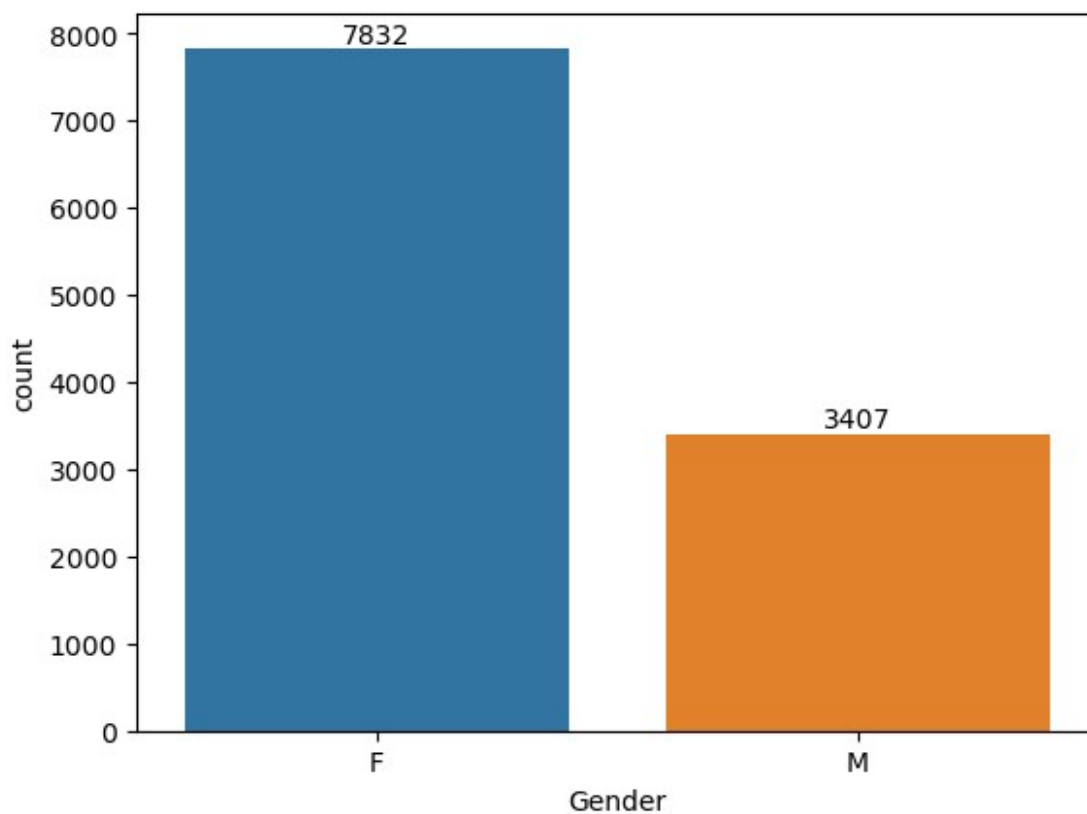
EXPLORATORY DATA ANALYSIS

GENDER

```
# plotting a bar chart for Gender and it's count
```

```
ax = sns.countplot(x = 'Gender', data = df)
```

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

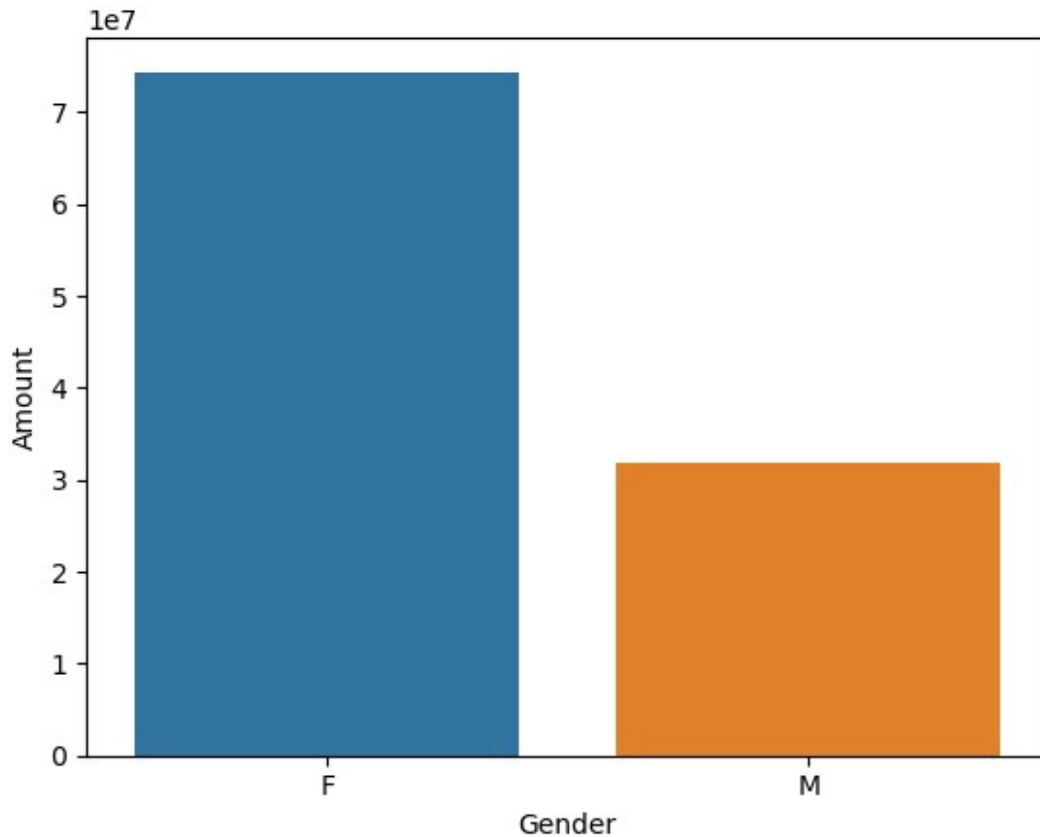


```
# plotting a bar chart for gender vs total amount
```

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

```
sns.barplot(x = 'Gender', y = 'Amount', data = sales_gender)
```

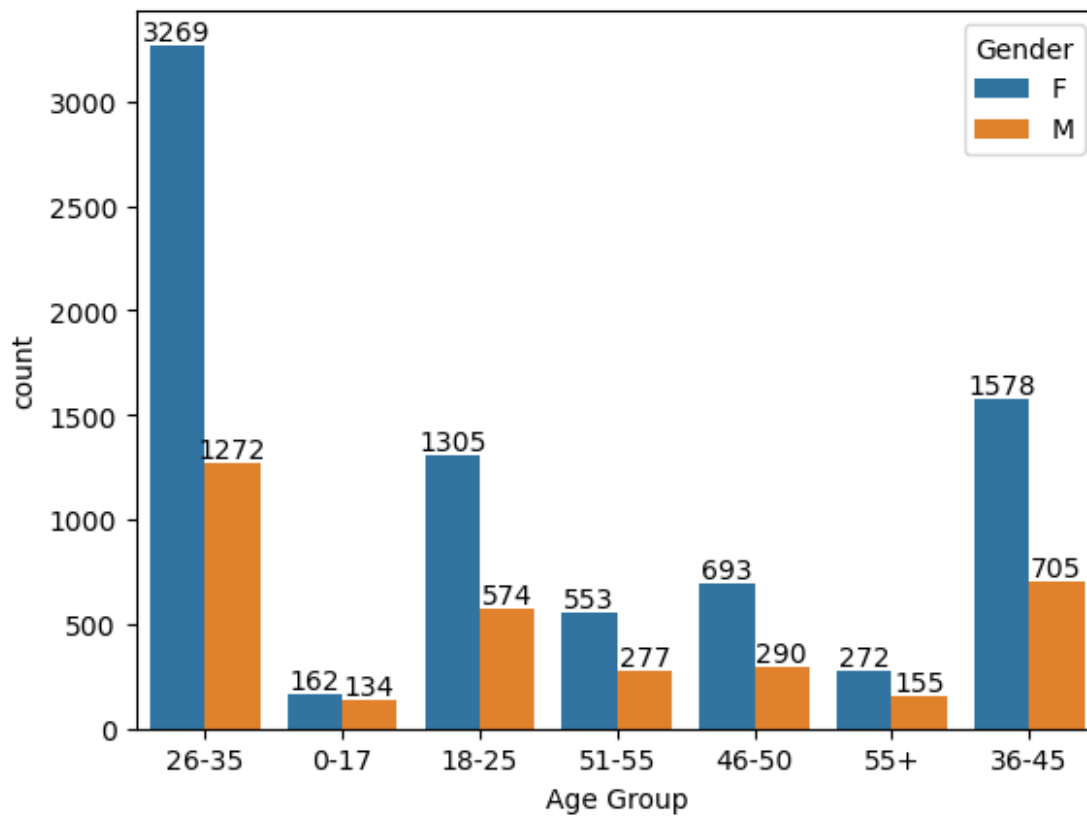
```
<Axes: xlabel='Gender', ylabel='Amount'>
```



From above graphs we can see that most of the buyers are females and even the purchasing power of females are greater than men

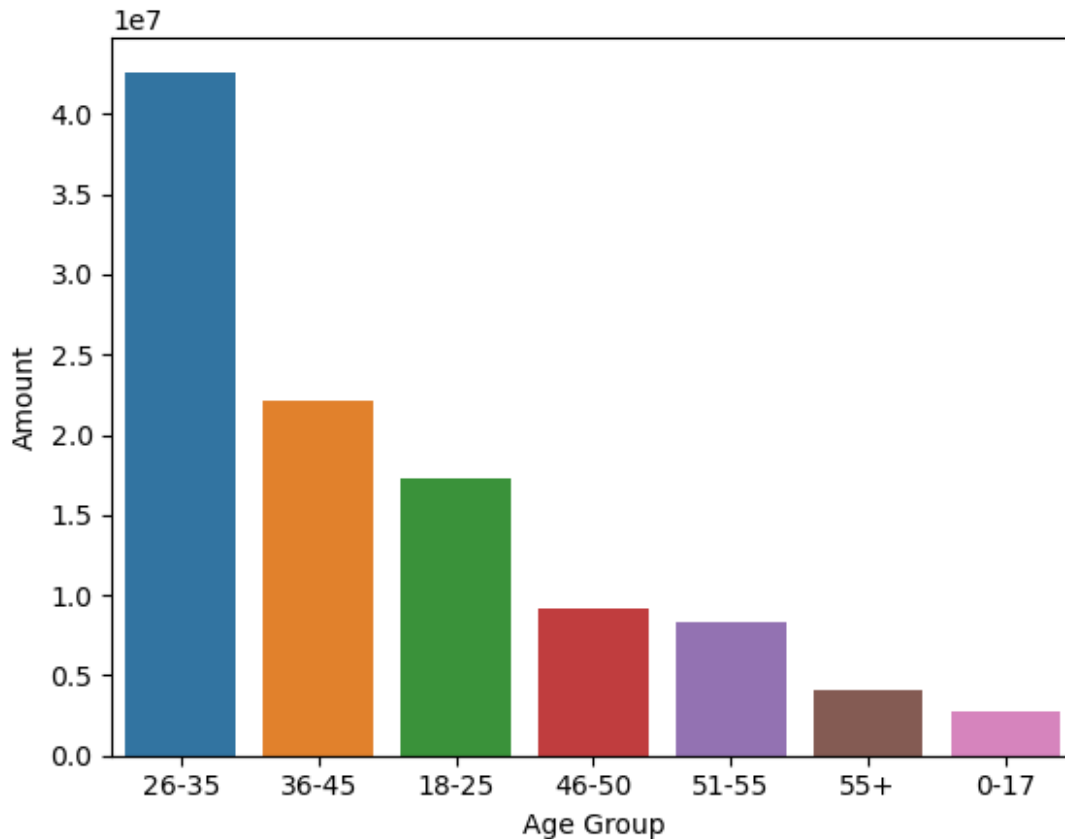
AGE

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



```
# Total Amount vs Age Group
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)
<Axes: xlabel='Age Group', ylabel='Amount'>
```



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

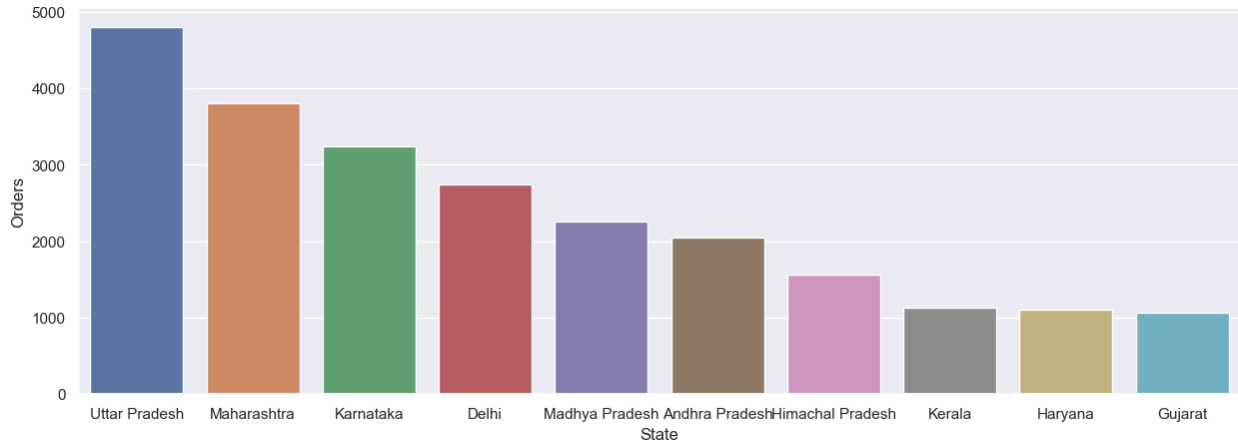
STATE

```
# total number of orders from top 10 states

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')

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

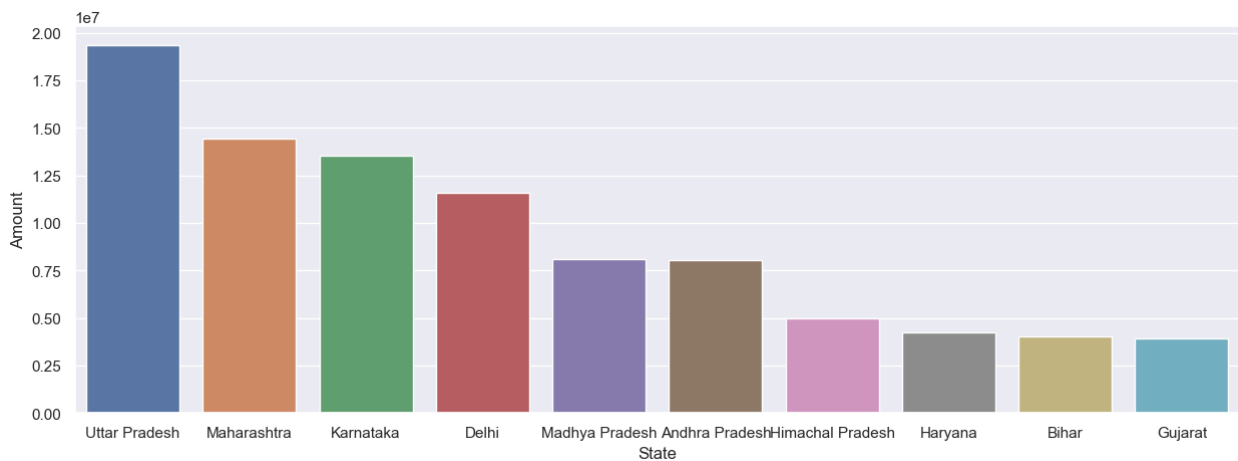


```
# total amount/sales from top 10 states
```

```
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')
```

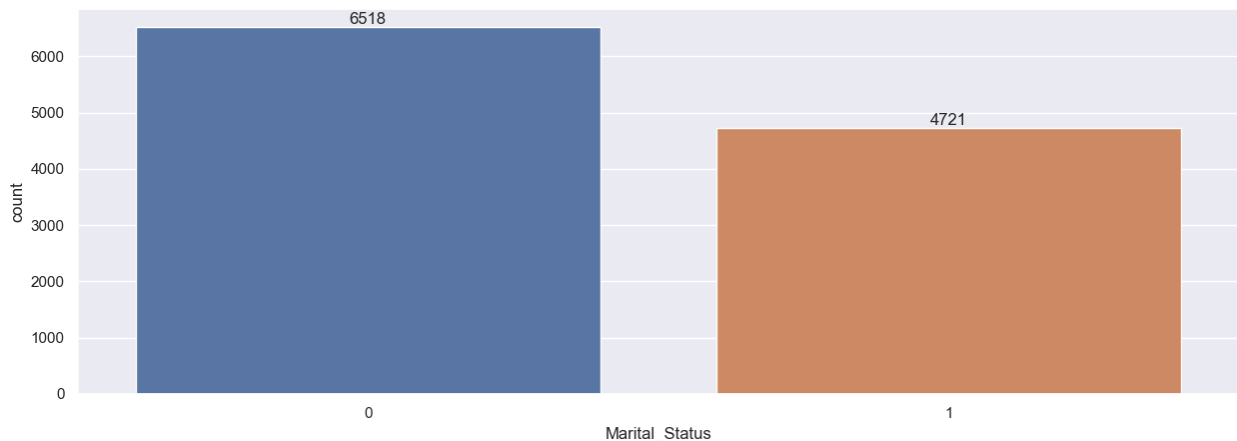
```
<Axes: xlabel='State', ylabel='Amount'>
```



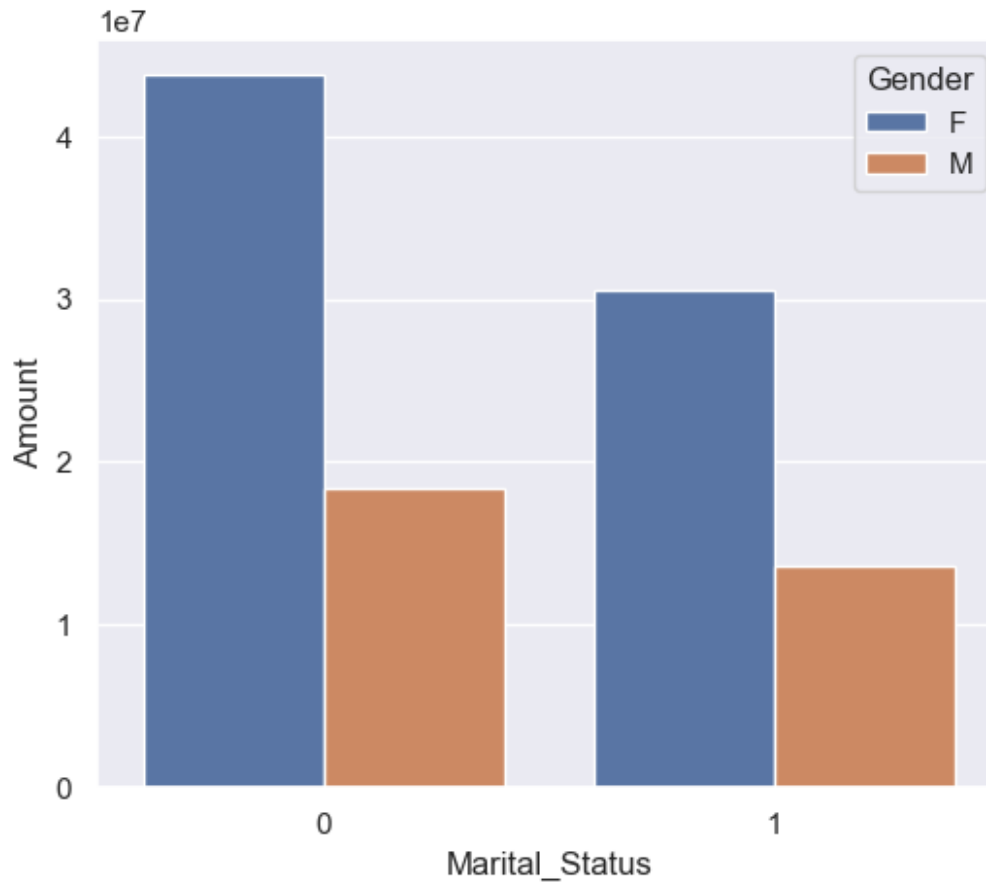
From above graphs we can see that most of the orders & total sales/amount are from Uttar Pradesh, Maharashtra and Karnataka respectively

MARITAL STATUS

```
ax = sns.countplot(data = df, x = 'Marital_Status')  
  
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'>
```

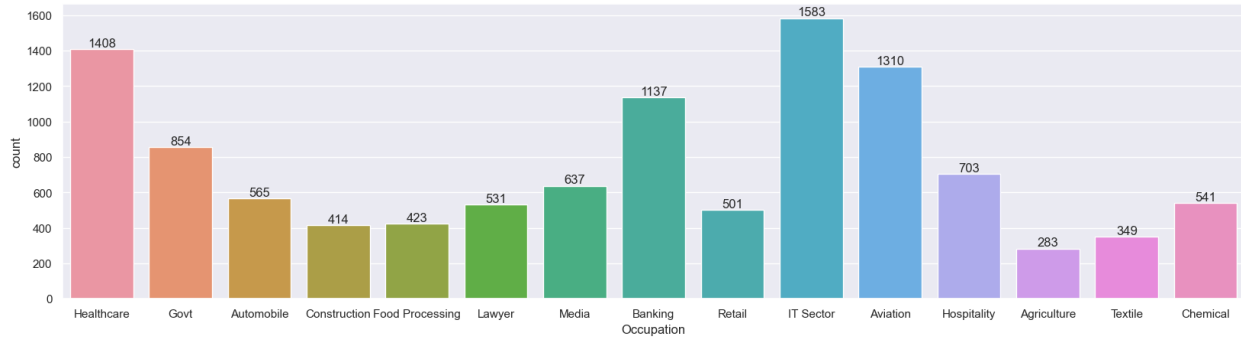


From above graphs we can see that most of the buyers are married (women) and they have high purchasing power

OCCUPATION

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

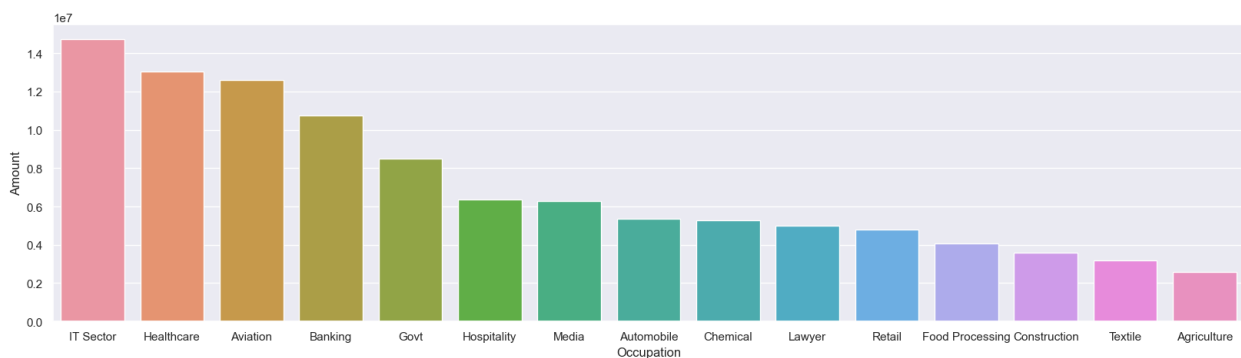
for bars in ax.containers:
    ax.bar_label(bars)
```



```
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')

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

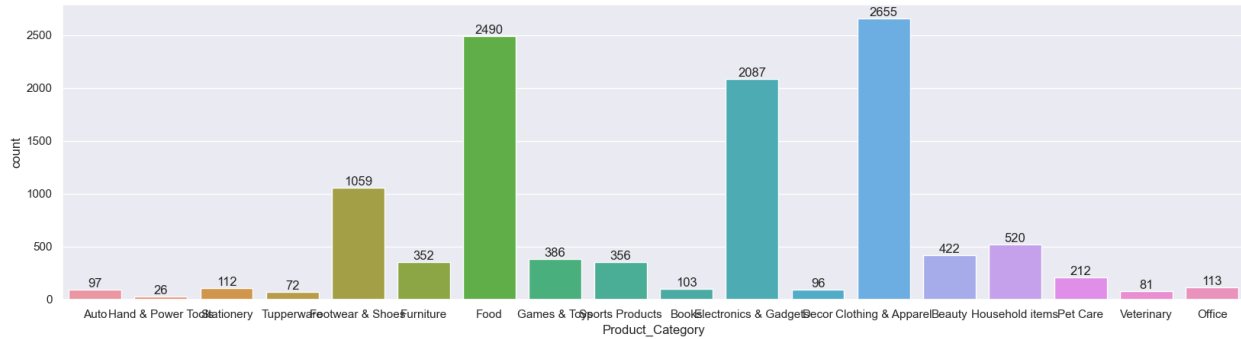


From above graphs we can see that most of the buyers are working in IT, Healthcare and Aviation sector

PRODUCT CATEGORY

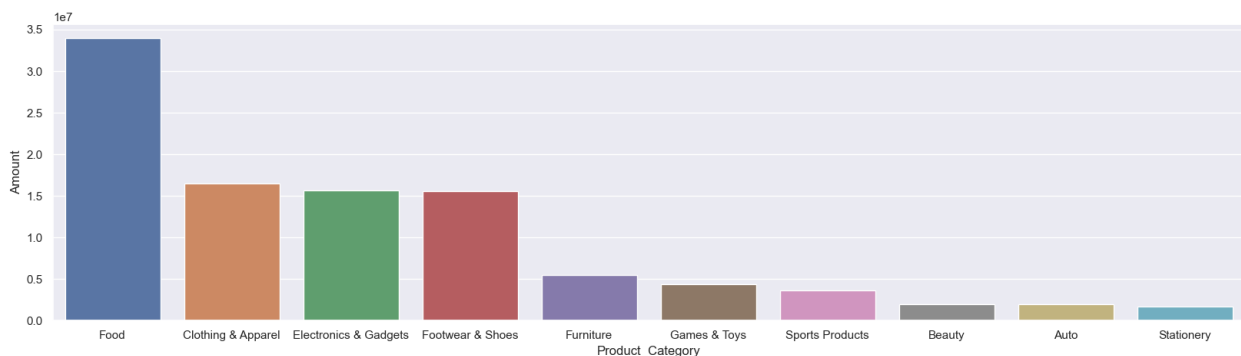
```
sns.set(rc={'figure.figsize':(20,5)})
ax = sns.countplot(data = df, x = 'Product_Category')

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

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

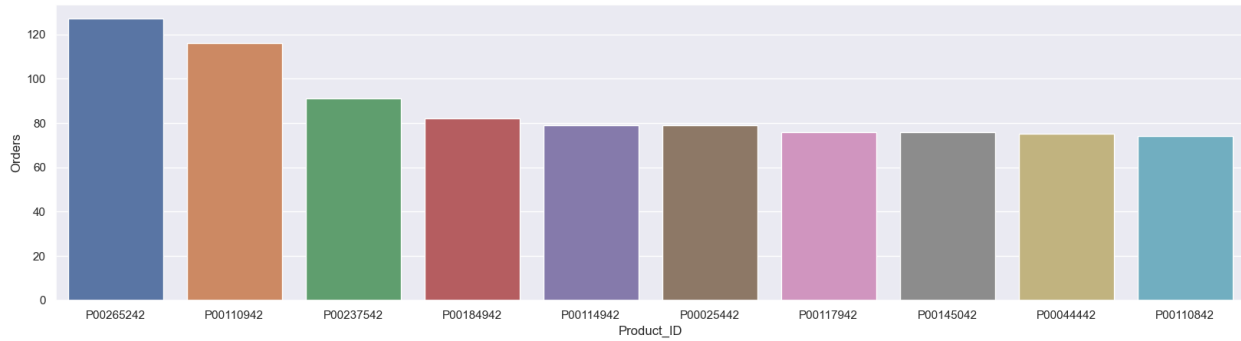
sns.set(rc={'figure.figsize':(20,5)})
sns.barplot(data = sales_state, x = 'Product_Category',y= 'Amount')
<Axes: xlabel='Product_Category', ylabel='Amount'>
```



From above graphs we can see that most of the sold products are from Food, Clothing and Electronics category

```
sales_state = df.groupby(['Product_ID'], as_index=False)
['Orders'].sum().sort_values(by='Orders', ascending=False).head(10)

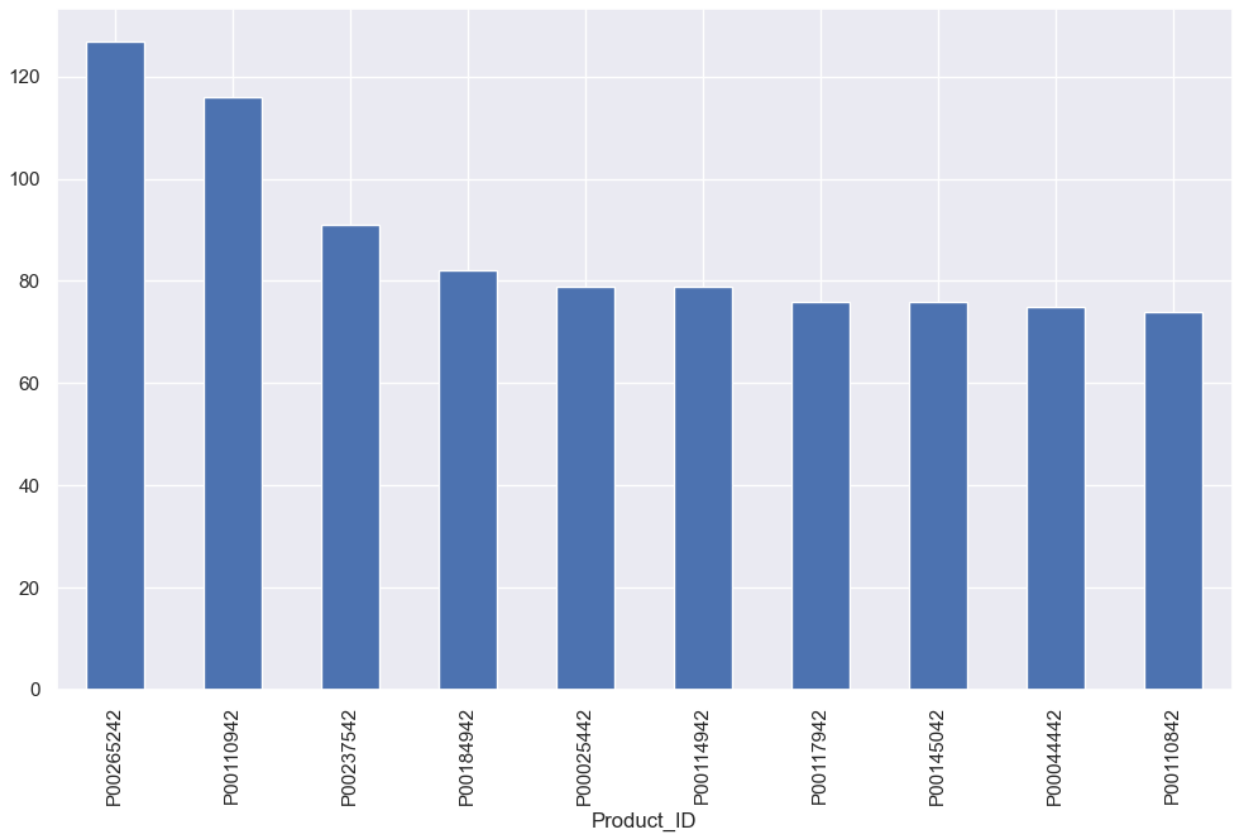
sns.set(rc={'figure.figsize':(20,5)})
sns.barplot(data = sales_state, x = 'Product_ID',y= 'Orders')
<Axes: xlabel='Product_ID', ylabel='Orders'>
```



top 10 most sold products

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

<Axes: xlabel='Product_ID'>



CONCLUSION:

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

