fbnj94530

December 15, 2024

[3]:
```
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
import seaborn as sns
```

[24]:
```
df = pd.read_csv(r'C:\Users\Asus\OneDrive\Desktop\EDA Project\Diwali Sales␣
↪Analysis\Diwali Sales Data.csv', encoding ='unicode_escape')
df
```

[24]:

	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	
…	…	…	…	…	…	…	…	
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	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]
```

1 To inspect the data

[25]:
```
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
```

2 Droping unrelated/blank columns

[26]:
```
df.drop(['Status', 'unnamed1'], axis = 1, inplace = True)
df
```

[26]:

	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	
…	…	…	…	…	…	…	…	
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

[11251 rows x 13 columns]

#We have dropped the Status and unnamed1 columns as they were containing null values

3 Checking for null values

[28]:
```
df.isnull().sum()
```

[28]:
```
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
```

#We can see that Amount has 12 null values

4 Dropping null values

[30]:
```
df.shape
```

[30]:
```
(11251, 13)
```

#We can see 11251 rows and 13 columns

[32]:
```
df.dropna(inplace = True)
df.shape
```

[32]:
```
(11239, 13)
```

#Now we have removed the null values

5 To Check all the Column Names

[34]:
```
df.columns
```

[34]:
```
Index(['User_ID', 'Cust_name', 'Product_ID', 'Gender', 'Age Group', 'Age',
       'Marital_Status', 'State', 'Zone', 'Occupation', 'Product_Category',
       'Orders', 'Amount'],
      dtype='object')
```

6 To get Descriptive Analysis

[36]:
```
df.describe()
```

[36]:

	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	9453.610858
std	1.716039e+03	12.753866	0.493589	1.114967	5222.355869
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

#Here we get descriptive analysis of all the columns

[38]:
```
df[["Age","Orders","Amount"]].describe()
```

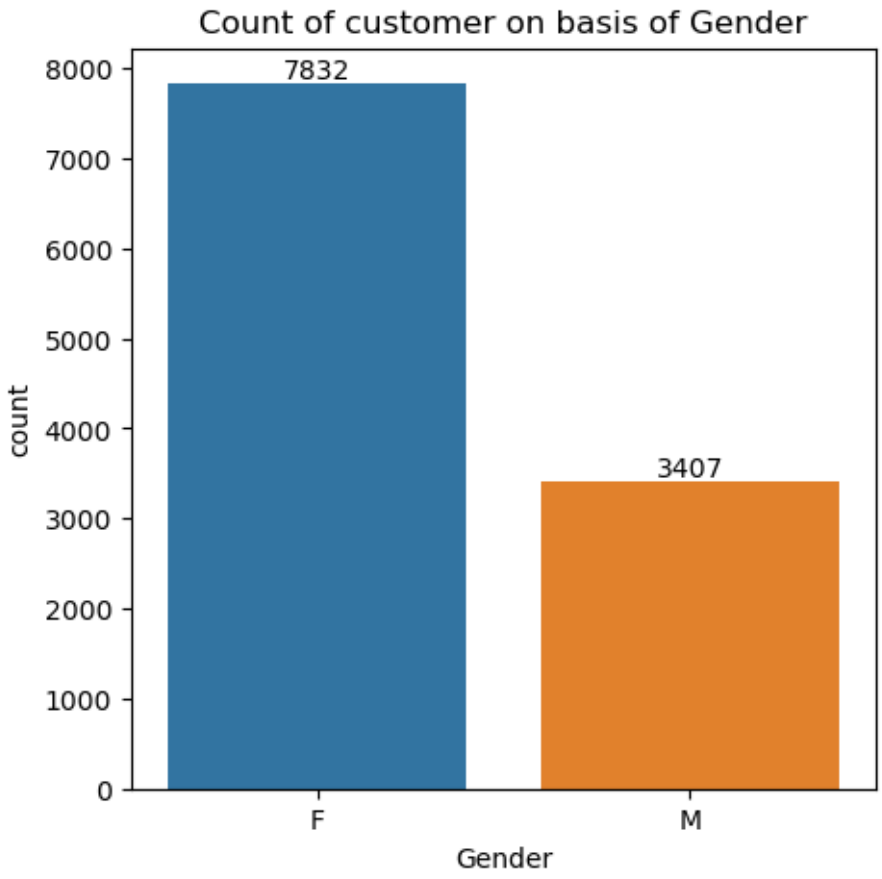
[38]:

	Age	Orders	Amount
count	11239.000000	11239.000000	11239.000000
mean	35.410357	2.489634	9453.610858
std	12.753866	1.114967	5222.355869
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

#To get Descriptive analysis of selected columns

7 Exploratory Data Analysis

[42]:
```
plt.figure(figsize = (5,5))
plt.title("Count of customer on basis of Gender")
ax = sns.countplot(x = "Gender", data = df)
for bars in ax.containers:
    ax.bar_label(bars)
```



#We can see that Female Customer are more than Male Customer

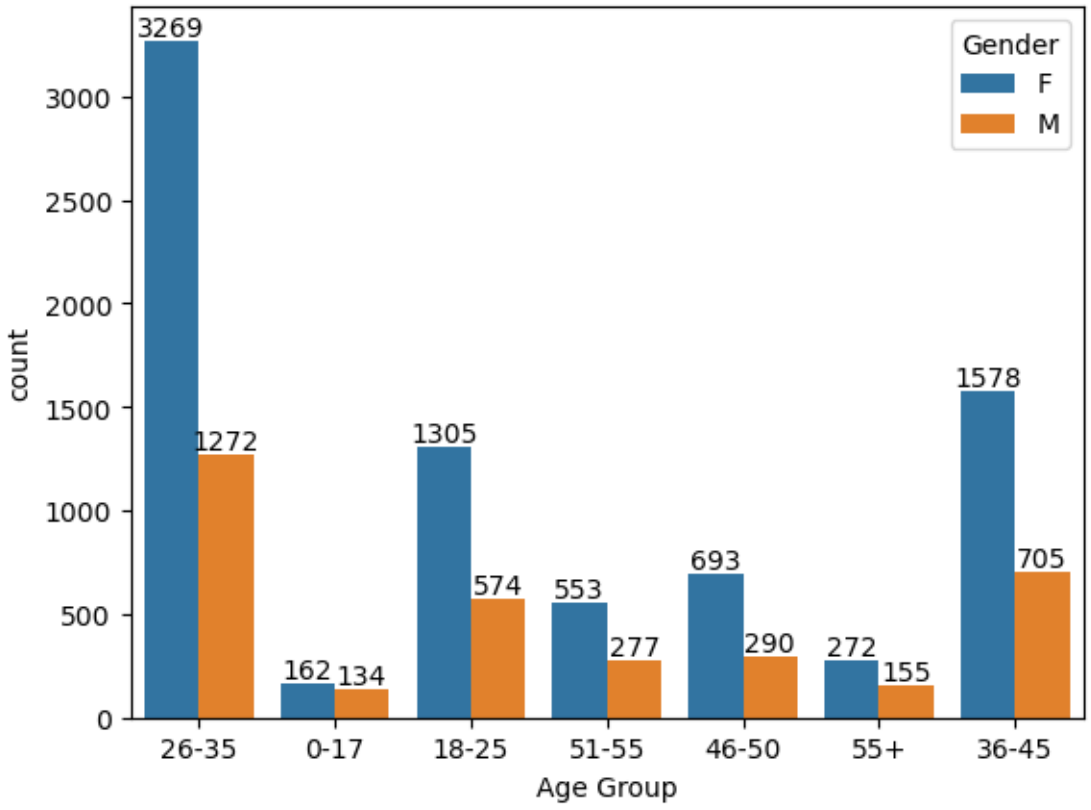
[48]:
```
sales_gen = df.groupby(["Gender"], as_index=False)["Amount"].sum().
 ↪sort_values(by="Amount", ascending=False)
sales_gen
```

[48]:

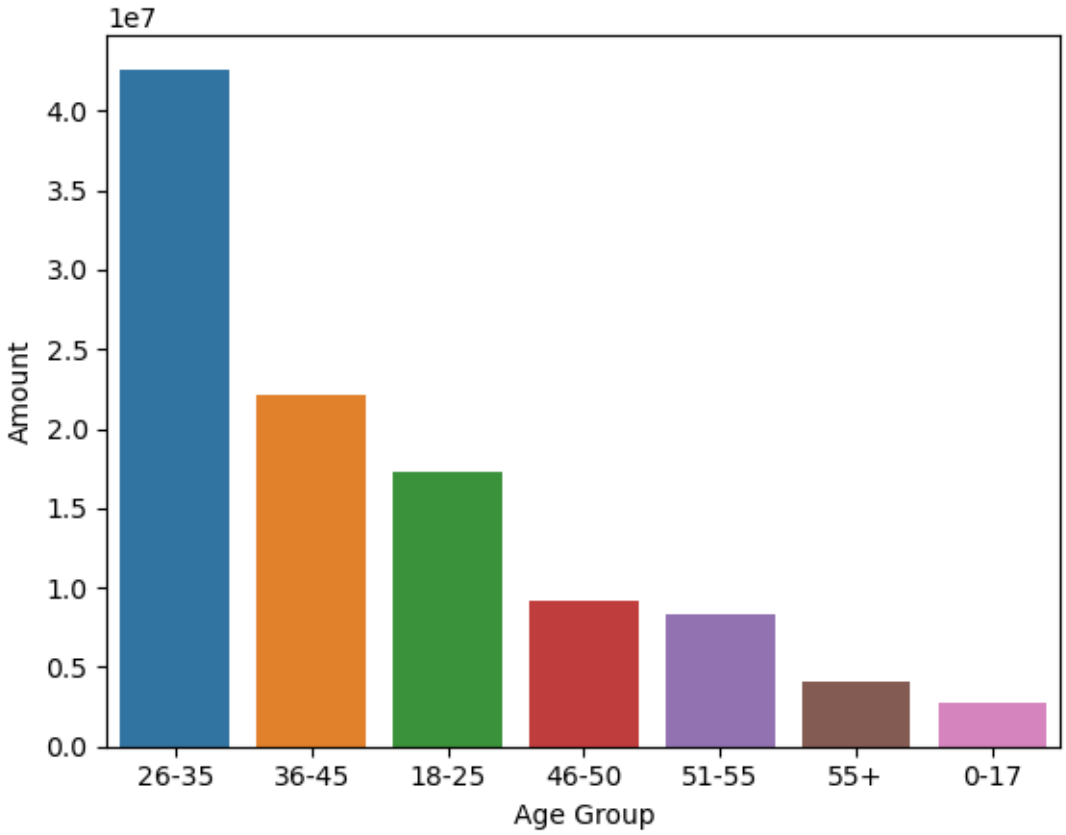
	Gender	Amount
0	F	74335856.43
1	M	31913276.00

#Female Customer have purchased more than Male customers

[51]:
```
ax = sns.countplot(x = "Age Group", data = df, hue = "Gender")
for bars in ax.containers:
    ax.bar_label(bars)
```



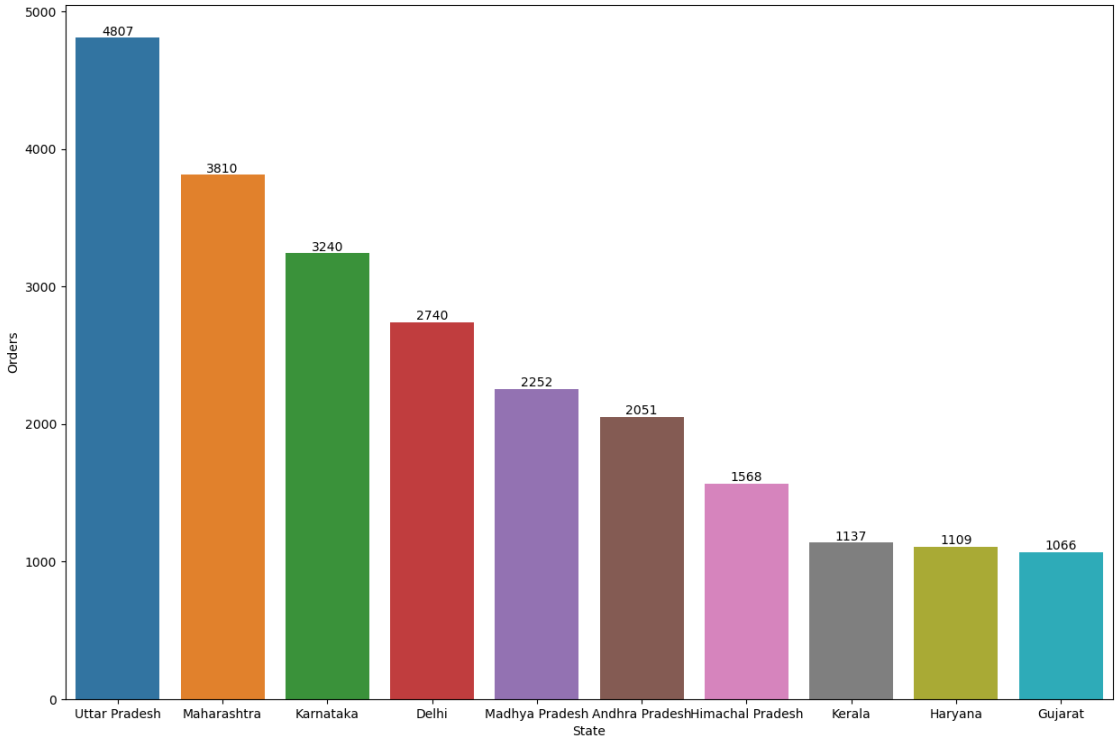
[71]:
```
sales_age = df.groupby(["Age Group"], as_index=False)["Amount"].sum().
  ↪sort_values(by="Amount", ascending=False)
ax = sns.barplot(x = "Age Group", y = "Amount", data = sales_age)
```



#Here we can see that our most of the Customer belongs to age group 26-35 (Female)

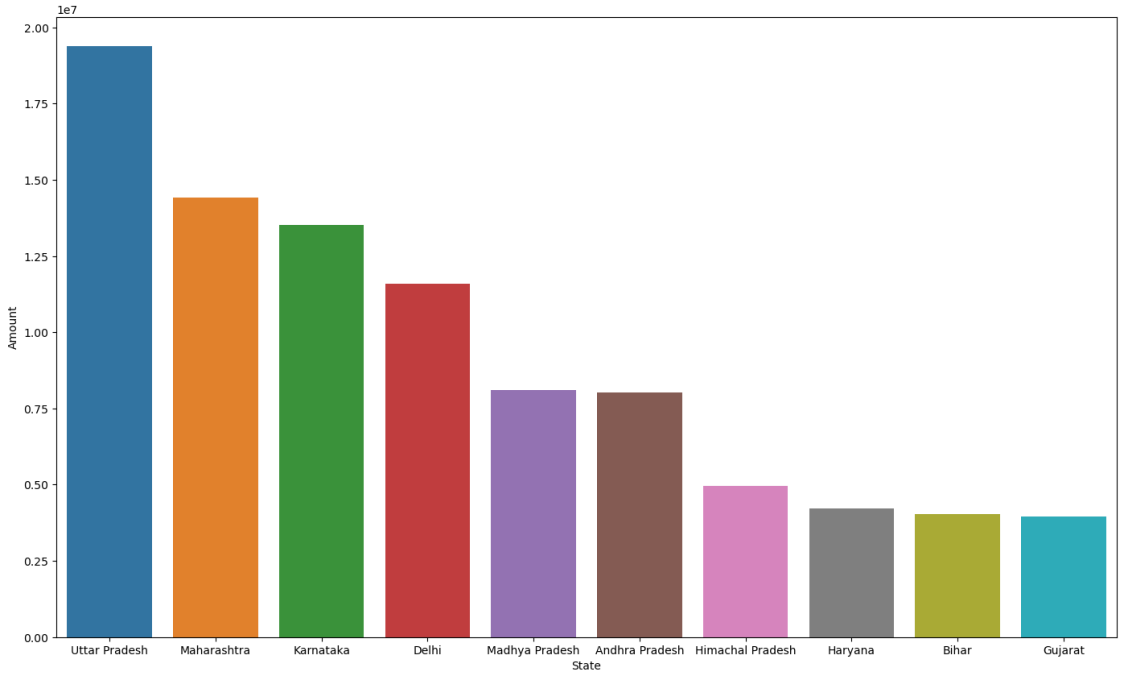
8 Top 10 States

[69]:
```
plt.figure(figsize=(15,10))
sales_state = df.groupby(["State"], as_index=False)["Orders"].sum().
 ↪sort_values(by="Orders", ascending=False).head(10)
ax = sns.barplot(x = "State", y = "Orders", data = sales_state)
for bars in ax.containers:
    ax.bar_label(bars)
```



#Here we can see that top three states with most no. of Orders are UP,Maharashtra and Karnataka

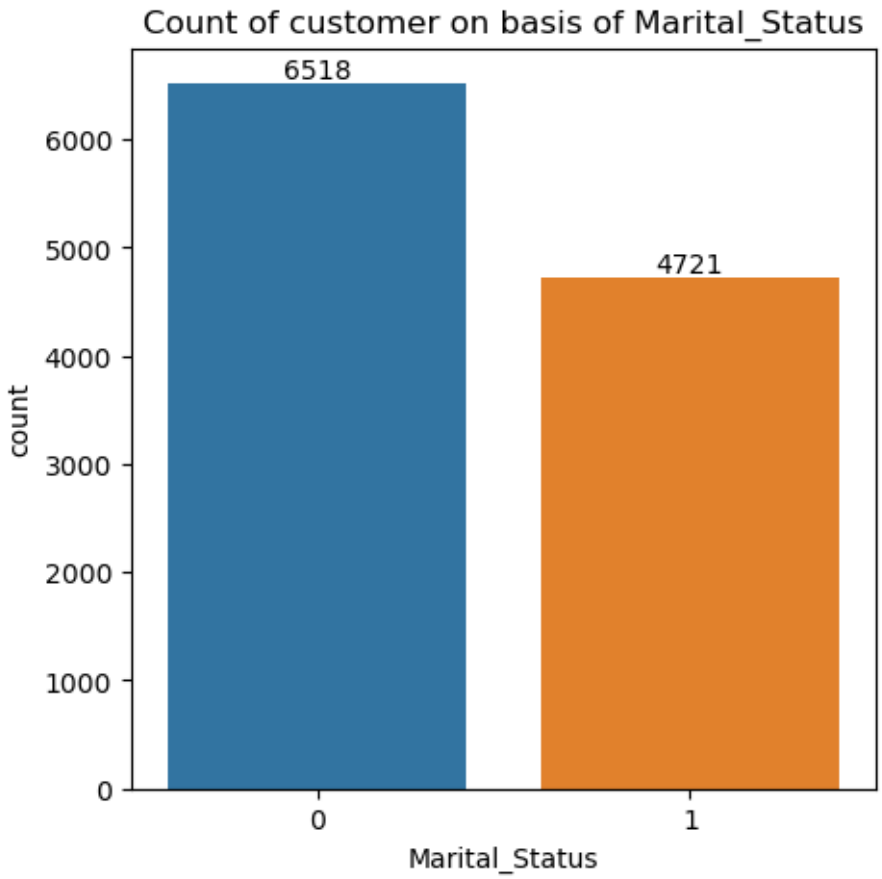
[76]:
```
plt.figure(figsize=(17,10))
sales_amount = df.groupby(["State"], as_index=False)["Amount"].sum().
  ↪sort_values(by="Amount", ascending=False).head(10)
ax = sns.barplot(x = "State", y = "Amount", data = sales_amount)
```



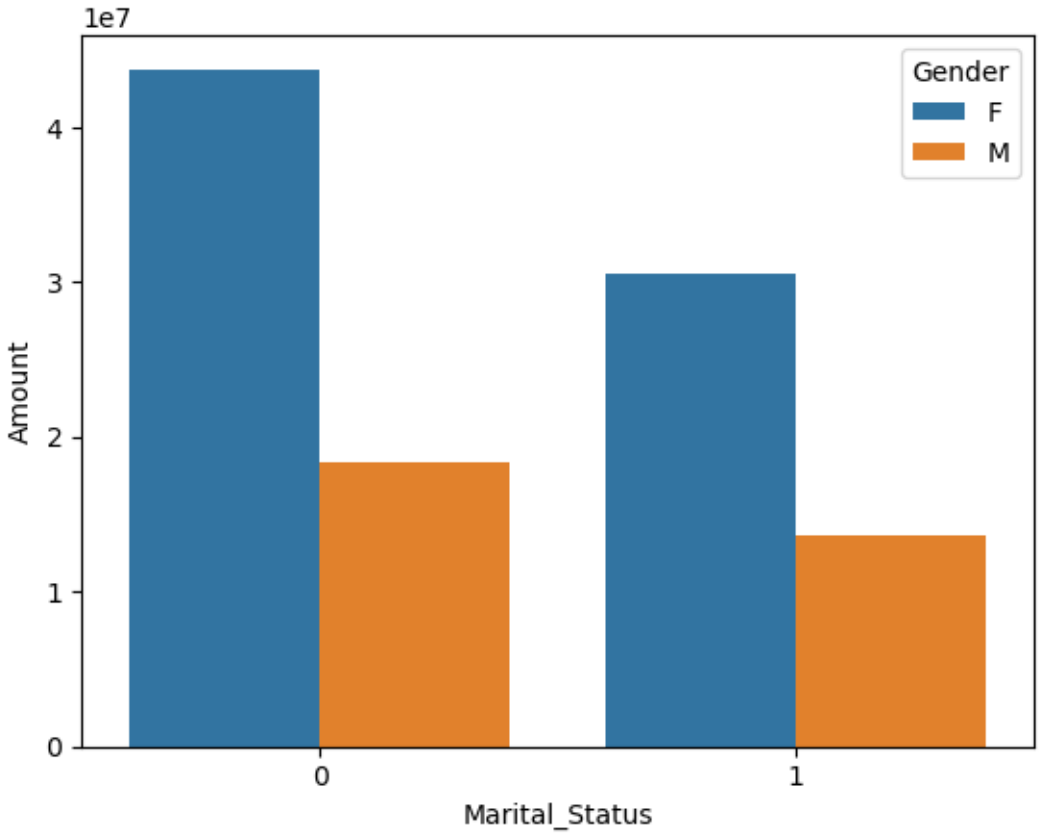
#Top 10 States based on Amount spend and here we can see the order change after Himachal Pradesh compared to above

9 Marital_Status

[78]:
```
plt.figure(figsize = (5,5))
plt.title("Count of customer on basis of Marital_Status")
ax = sns.countplot(x = "Marital_Status", data = df)
for bars in ax.containers:
    ax.bar_label(bars)
```



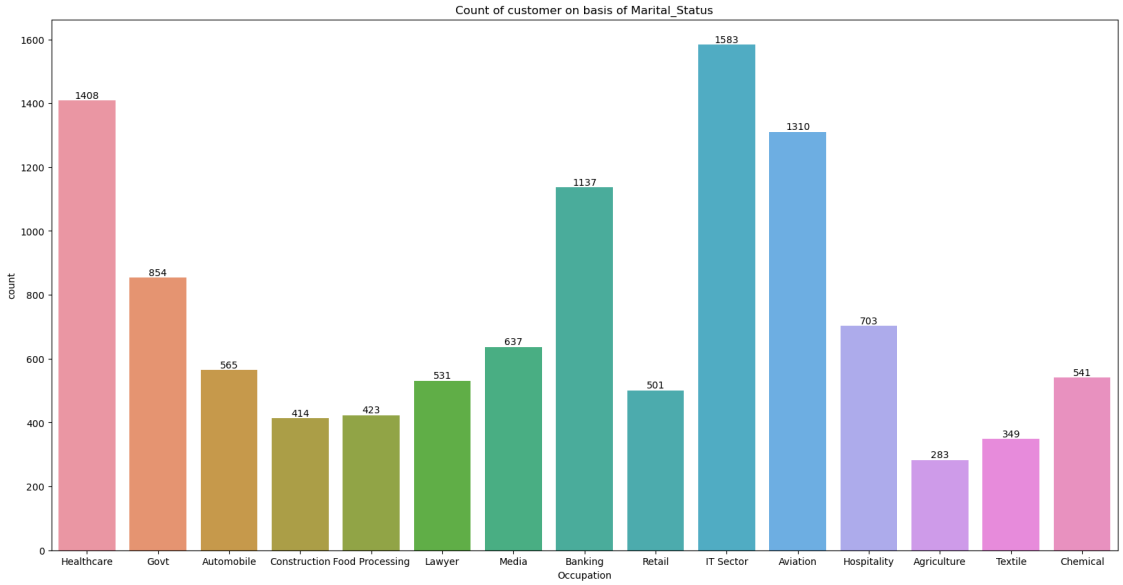
[84]:
```
sales_marital = df.groupby(["Marital_Status","Gender"],␣
↪as_index=False)["Amount"].sum().sort_values(by="Amount", ascending=False).
↪head(10)
ax = sns.barplot(x = "Marital_Status", y = "Amount", data = sales_marital, hue␣
↪= "Gender")
```



#Here we can see that most of our buyers are Unmarried

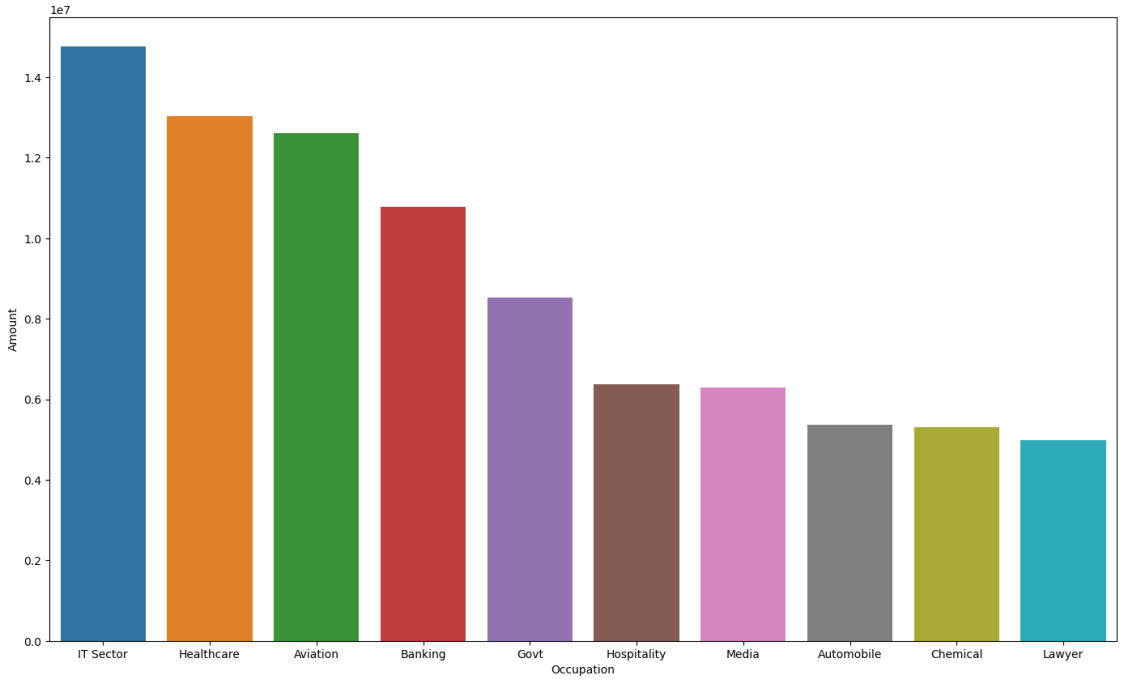
10 Occupation

[85]:
```
plt.figure(figsize = (20,10))
plt.title("Count of customer on basis of Occupation")
ax = sns.countplot(x = "Occupation", data = df)
for bars in ax.containers:
    ax.bar_label(bars)
```



#Most of our Customer are from IT Sector

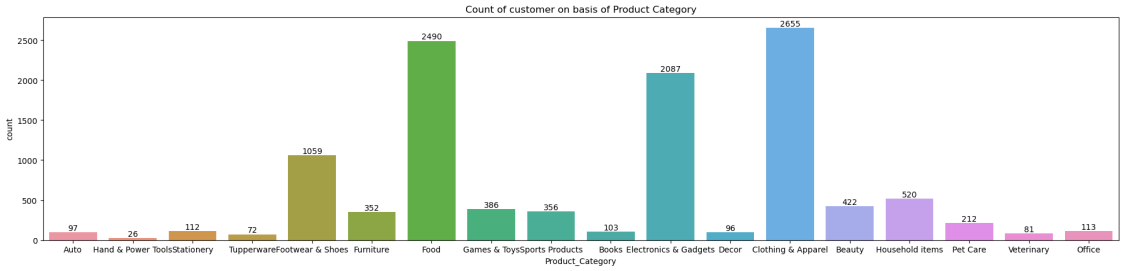
[87]:
```
plt.figure(figsize=(17,10))
sales_occupation = df.groupby(["Occupation"], as_index=False)["Amount"].sum().
↪sort_values(by="Amount", ascending=False).head(10)
ax = sns.barplot(x = "Occupation", y = "Amount", data = sales_occupation)
```



#We can see that most of our buyers are from IT Sector,Healthcare,Aviation and banking Sector

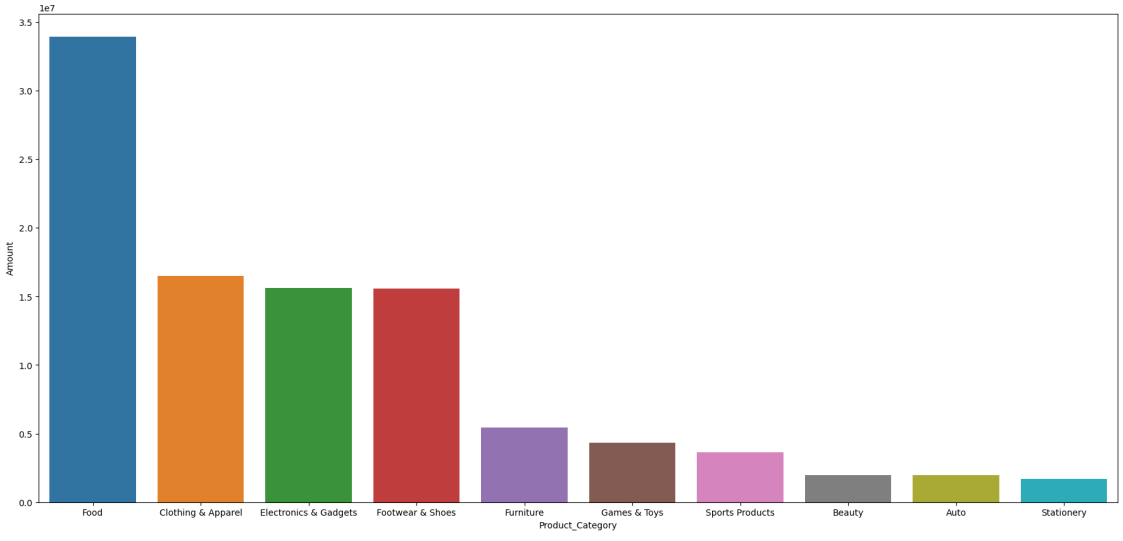
11 Product Category

[98]:
```
plt.figure(figsize = (24,5))
plt.title("Count of customer on basis of Product Category")
ax = sns.countplot(x = "Product_Category", data = df)
for bars in ax.containers:
    ax.bar_label(bars)
```



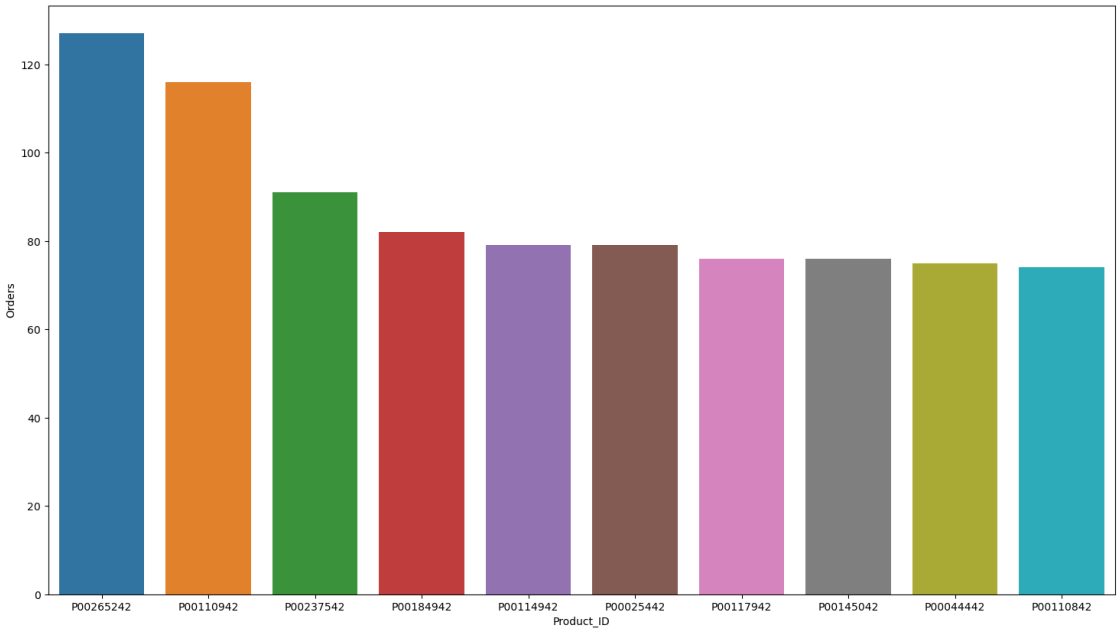
#Top 3 products are Clothing & Apparel, Food and Electronic and Gadgets

[101]:
```
plt.figure(figsize=(22,10))
sales_product = df.groupby(["Product_Category"], as_index=False)["Amount"].
  ↪sum().sort_values(by="Amount", ascending=False).head(10)
ax = sns.barplot(x = "Product_Category", y = "Amount", data = sales_product)
```



#Here we see that the most amount spend on was Food, Clothing & Apparel and Electronics & Gadget

[111]:
```
plt.figure(figsize=(18,10))
sales_productID = df.groupby(["Product_ID"], as_index=False)["Orders"].sum().
↪sort_values(by="Orders", ascending=False).head(10)
ax = sns.barplot(x = "Product_ID", y = "Orders", data = sales_productID)
```



#Most ordered product ID id P0026524

[ ]: