Out[5]:

	User_ID	Cust_name	Product_ID	Gender	Age Group	Age	Marital_Status	State	Zone	Occupation	Product_Category
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	М	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	М	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	М	36-45	37	0	Karnataka	Southern	Agriculture	Office
11250	1002744	Brumley	P00281742	F	18-25	19	0	Maharashtra	Western	Healthcare	Office

11251 rows × 15 columns

In [3]: 1 df.shape

Out[3]: (11251, 15)

In [4]: 1 df.head()

Out[4]:

	User_ID	Cust_name	Product_ID	Gender	Age Group	Age	Marital_Status	State	Zone	Occupation	Product_Category	Ord
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	М	0-17	16	0	Karnataka	Southern	Construction	Auto	
4	1000588	Joni	P00057942	M	26-35	28	1	Gujarat	Western	Food Processing	Auto	
4 =												

•

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

In [5]:

1 df.info()

```
In [7]:
           1 #check for null values
           2 pd.isnull(df).sum()
 Out[7]: User_ID
                              0
         Cust_name
                              0
         Product_ID
                              0
         Gender
                              0
         Age Group
                              0
         Age
                              0
         Marital_Status
                              0
         State
                              0
                              0
         Zone
         Occupation 0
                              0
         Product_Category
                              0
         Orders
                              0
         Amount
                             12
         dtype: int64
 In [8]:
           1 # drop null values
           2 df.dropna(inplace=True)
 In [9]:
           1 # change data type
           2 df['Amount'] = df['Amount'].astype('int')
           1 df['Amount'].dtypes
In [10]:
Out[10]: dtype('int32')
In [11]:
           1 df.columns
Out[11]: Index(['User_ID', 'Cust_name', 'Product_ID', 'Gender', 'Age Group', 'Age',
                 'Marital_Status', 'State', 'Zone', 'Occupation', 'Product_Category',
                'Orders', 'Amount'],
               dtype='object')
```

```
In [12]: 1 #rename column
```

2 df.rename(columns= {'Marital_Status':'Shaadi'})

Out[12]:

	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	М	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	М	18-25	19	1	Maharashtra	Western	Chemical	Office	4
11247	1004089	Reichenbach	P00171342	М	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	М	36-45	37	0	Karnataka	Southern	Agriculture	Office	;
11250	1002744	Brumley	P00281742	F	18-25	19	0	Maharashtra	Western	Healthcare	Office	;

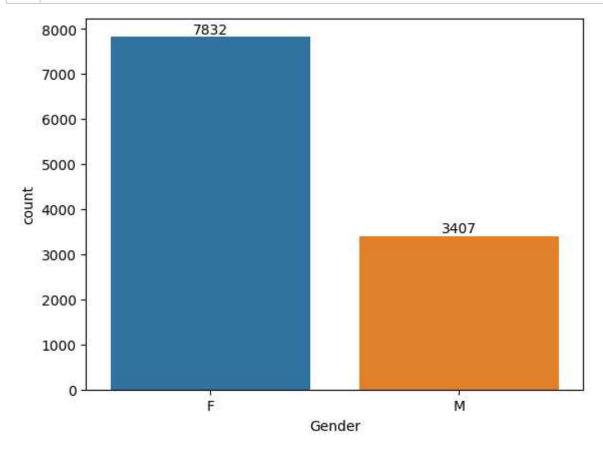
11239 rows × 13 columns

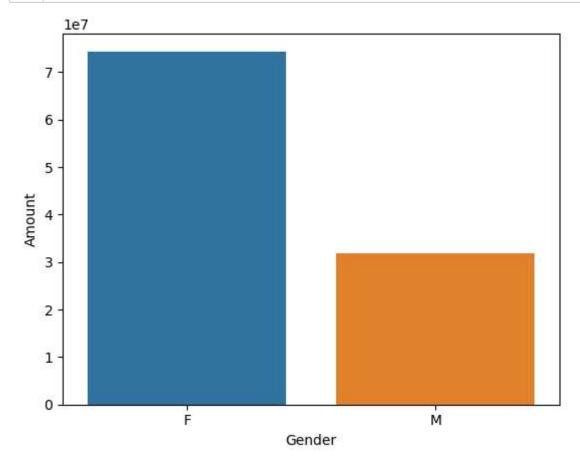
```
In [13]:
            1 # describe() method returns description of the data in the DataFrame (i.e. count, mean, std, etc)
            2 df.describe()
Out[13]:
                       User ID
                                       Age Marital Status
                                                               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
                 1.716039e+03
                                  12.753866
                                                 0.493589
                                                              1.114967
                                                                        5222.355168
                 1.000001e+06
                                  12.000000
                                                 0.000000
                                                              1.000000
                                                                         188.000000
             min
                 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
In [14]:
            1 # use describe() for specific columns
            2 df[['Age', 'Orders', 'Amount']].describe()
Out[14]:
```

	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

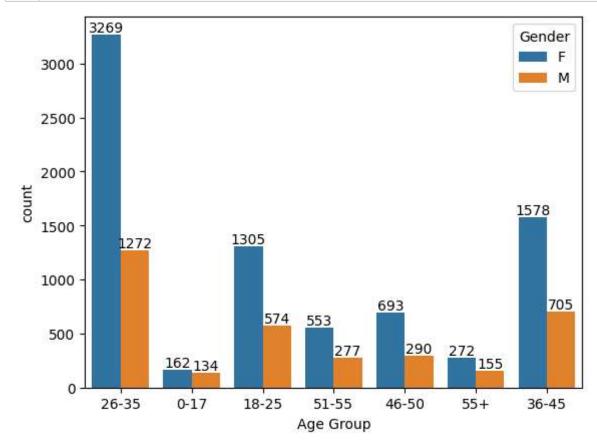
Gender

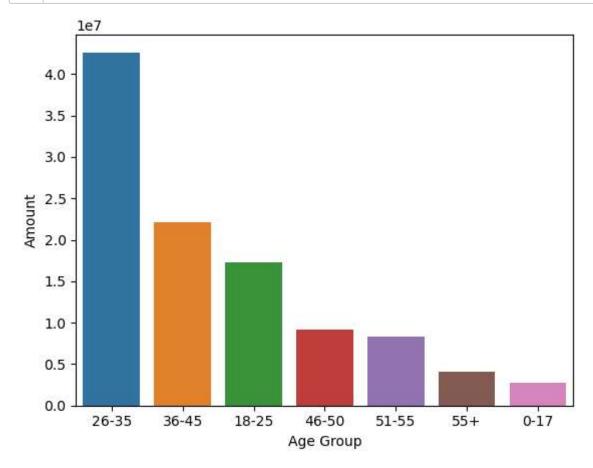




From above graphs we can see that most of the buyers are females and even the purchasing power of females are greater than men

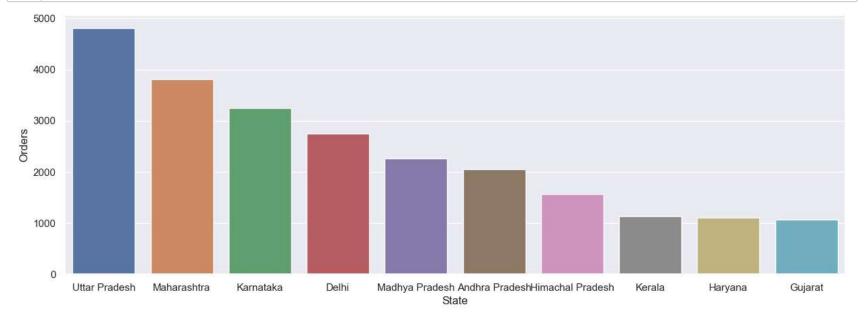
Age

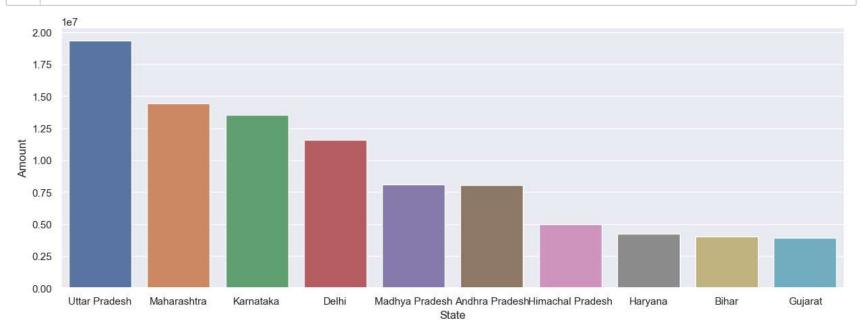




From above graphs we can see that most of the buyers are of age group between 26-35 yrs female

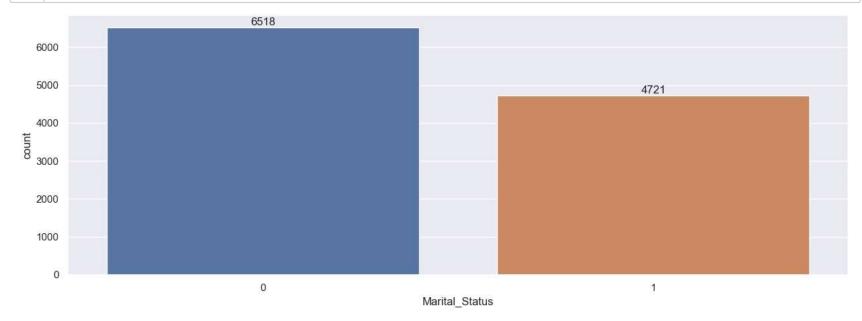
State

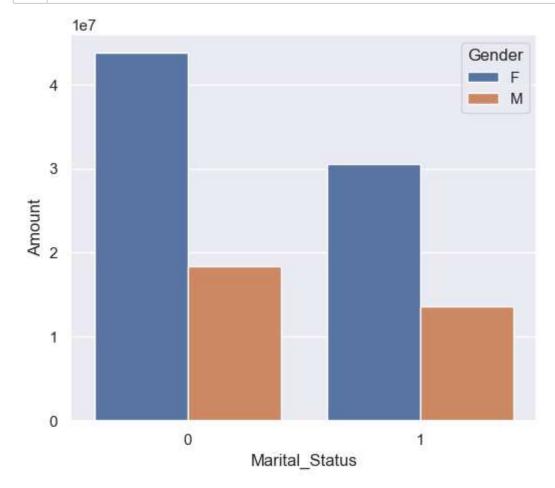




From above graphs we can see that most of the orders & total sales/amount are from Uttar Pradesh, Maharashtra and Karnataka respectively

Marital Status

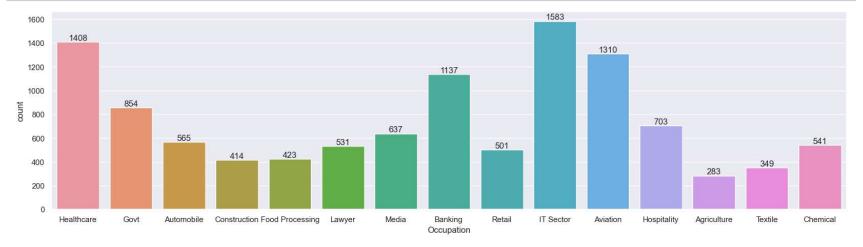


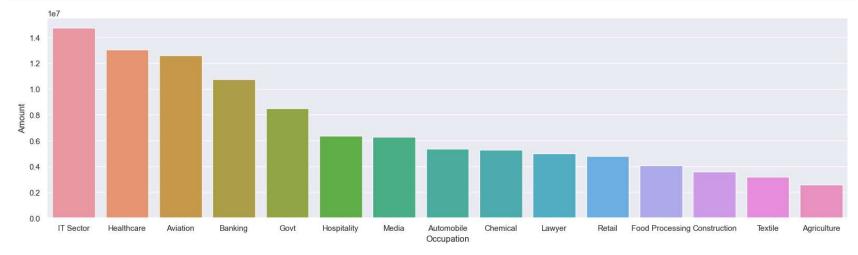


From above graphs we can see that most of the buyers are married (women) and they have high purchasing power

Occupation

```
In [23]: 1 sns.set(rc={'figure.figsize':(20,5)})
2 ax = sns.countplot(data = df, x = 'Occupation')
3
4 for bars in ax.containers:
    ax.bar_label(bars)
```

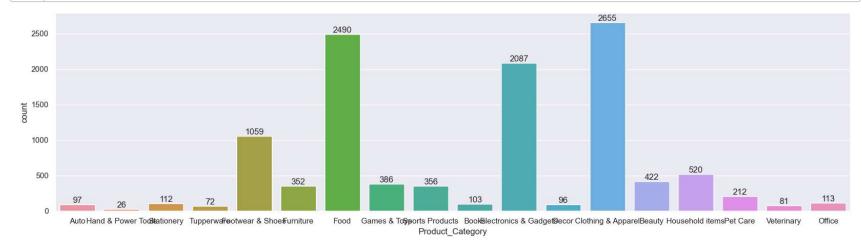


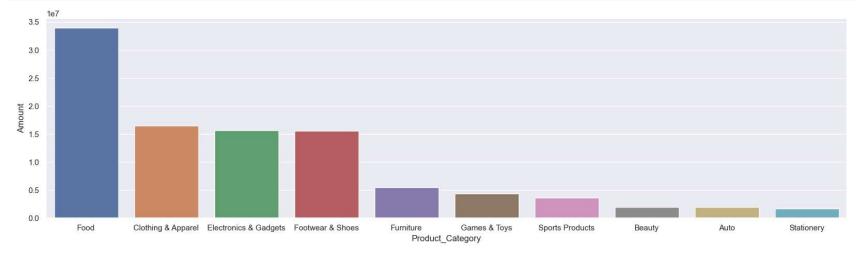


From above graphs we can see that most of the buyers are working in IT, Healthcare and Aviation sector

Product Category

```
In [25]: 1 sns.set(rc={'figure.figsize':(20,5)})
2 ax = sns.countplot(data = df, x = 'Product_Category')
3
4 for bars in ax.containers:
    ax.bar_label(bars)
```





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

