

## 1 Setting The Working Environment

#### In [1]:

- 1 import numpy as np
- 2 import pandas as pd
- 3 import matplotlib.pyplot as plt
- 4 %matplotlib inline
- 5 import seaborn as sns

## **2 Importing The Dataset**

```
In [2]:
```

1 | df = pd.read\_csv('Diwali Sales Data.csv', encoding= 'unicode\_escape')

# 3 Exploring Dataset

#### 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	
0	1002903	Sanskriti	P00125942	F	26-35	28	0	Maharashtra	V
1	1000732	Kartik	P00110942	F	26-35	35	1	Andhra Pradesh	Sc
2	1001990	Bindu	P00118542	F	26-35	35	1	Uttar Pradesh	(
3	1001425	Sudevi	P00237842	М	0-17	16	0	Karnataka	Sc
4	1000588	Joni	P00057942	М	26-35	28	1	Gujarat	٧
4									•

#### In [5]:

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				
d+vnos: $flort64(2)$ $int64(4)$ $objoct(8)$							

dtypes: float64(3), int64(4), object(8)

memory usage: 1.3+ MB

#### In [6]:

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

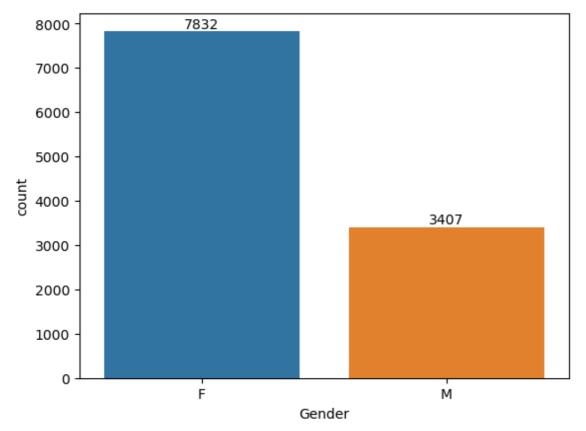
```
In [7]:
 1 pd.isnull(df).sum()
Out[7]:
User_ID
                     0
Cust_name
Product_ID
                     0
Gender
Age Group
                     0
Age
Marital_Status
                     0
State
                     0
Zone
                     0
Occupation
                     0
Product_Category
Orders
                     0
Amount
                     12
dtype: int64
In [8]:
 1 df.dropna(inplace=True)
In [9]:
   df['Amount'] = df['Amount'].astype('int')
In [10]:
 1 df['Amount'].dtypes
Out[10]:
dtype('int32')
```

# Gender



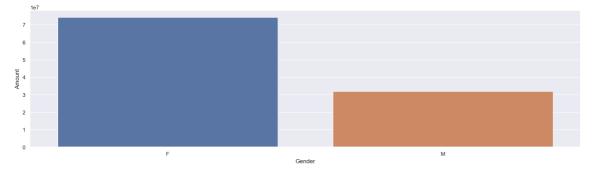
#### In [11]:

```
ax = sns.countplot(x = 'Gender', data = df)
2
3
  for bars in ax.containers:
       ax.bar_label(bars)
```



#### In [49]:

```
sales_gen = df.groupby(['Gender'], as_index=False)['Amount'].sum().sort_values(by='A
1
2
  sns.barplot(x = 'Gender',y= 'Amount' ,data = sales_gen)
3
  plt.show()
```



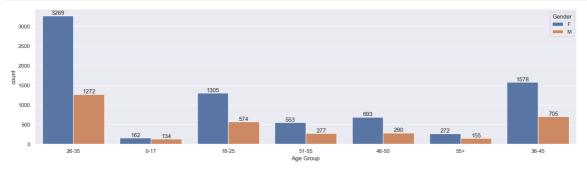
From above graphs we can see that most of the buyers are females and even the purchasing power of females are greater than men

### Age



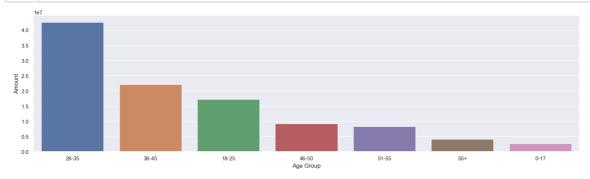
### In [42]:

```
ax = sns.countplot(data = df, x = 'Age Group', hue = 'Gender')
2
3
  for bars in ax.containers:
4
      ax.bar_label(bars)
```



### In [48]:

```
sales_age = df.groupby(['Age Group'], as_index=False)['Amount'].sum().sort_values(by
3
  sns.barplot(x = 'Age Group',y= 'Amount' ,data = sales_age)
4
  plt.show()
```



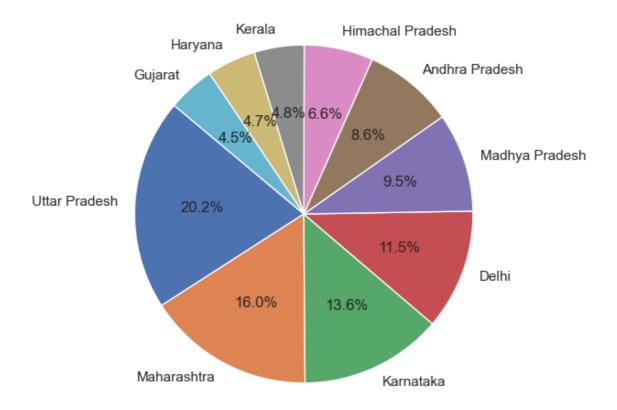
From above graphs we can see that most of the buyers are of age group between 26-35 yrs female

### **State**



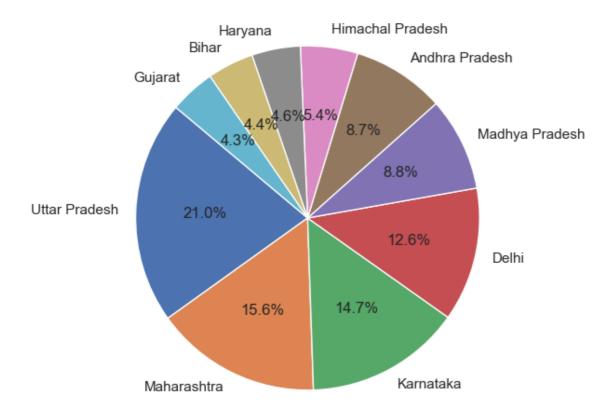
#### In [45]:

```
sales_state = df.groupby(['State'], as_index=False)['Orders'].sum().sort_values(by='
  plt.figure(figsize=(10, 6))
  plt.pie(sales_state['Orders'], labels=sales_state['State'], autopct='%1.1f%%', start
5
  plt.show()
```



#### In [46]:

```
sales_state = df.groupby(['State'], as_index=False)['Amount'].sum().sort_values(by='
2
3
  plt.figure(figsize=(10, 6))
5
  plt.pie(sales_state['Amount'], labels=sales_state['State'], autopct='%1.1f%%', start
  plt.show()
```



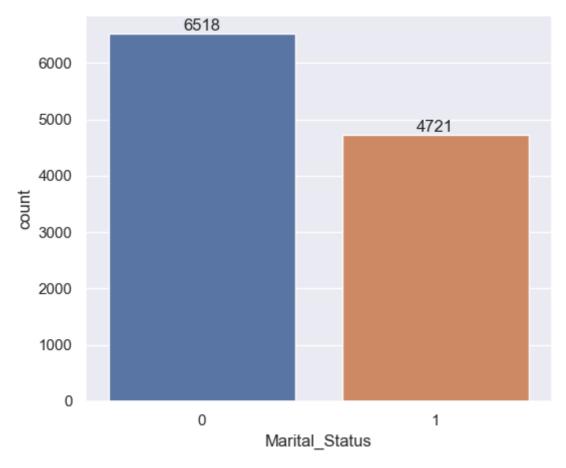
From above graphs we can see that most of the orders & total sales/amount are from Uttar Pradesh, Maharashtra and Karnataka respectively

## **Marital Status**



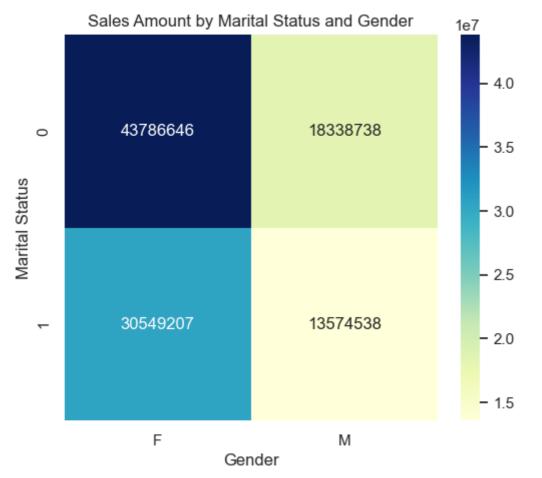
#### In [63]:

```
1  ax = sns.countplot(data = df, x = 'Marital_Status')
2
3  sns.set(rc={'figure.figsize':(7,5)})
4  for bars in ax.containers:
    ax.bar_label(bars)
```



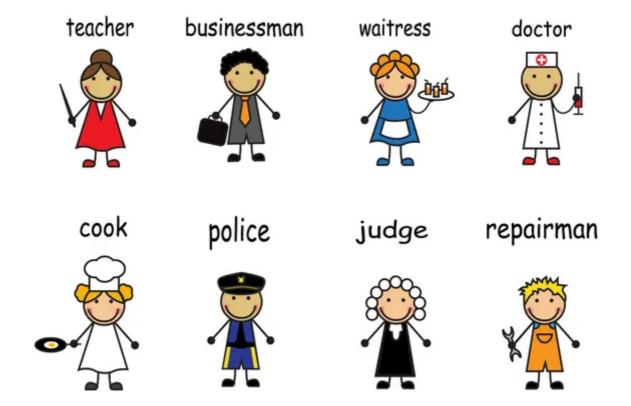
#### In [61]:

```
sales_state_pivot = df.pivot_table(index='Marital_Status', columns='Gender', values=
 2
   sns.set(rc={'figure.figsize': (6, 5)})
4
   sns.heatmap(sales_state_pivot, annot=True, fmt='.0f', cmap='YlGnBu')
 5
   plt.xlabel('Gender')
 6
   plt.ylabel('Marital Status')
 7
   plt.title('Sales Amount by Marital Status and Gender')
10
   plt.show()
11
```



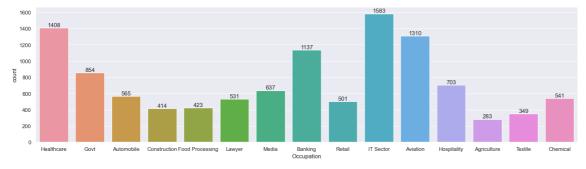
From above graphs we can see that most of the buyers are married (women) and they have high purchasing power

## **Occupation**



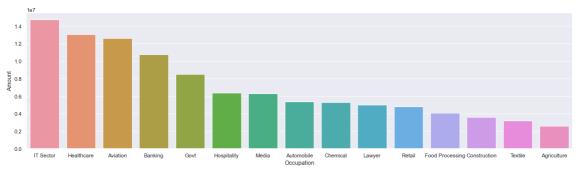
### In [57]:

```
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)
```



#### In [65]:

```
sales_state = df.groupby(['Occupation'], as_index=False)['Amount'].sum().sort_values
2
  sns.set(rc={'figure.figsize':(20,5)})
  sns.barplot(data = sales_state, x = 'Occupation',y= 'Amount')
5
  plt.show()
```



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

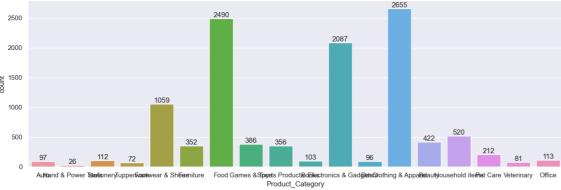
## **Product Category**



#### In [67]:

```
sns.set(rc={'figure.figsize':(16,5)})
ax = sns.countplot(data = df, x = 'Product_Category')

for bars in ax.containers:
    ax.bar_label(bars)
2655
```



### In [64]:

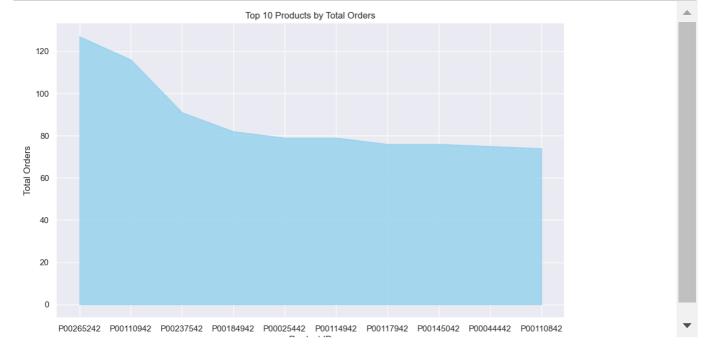
```
sales_state = df.groupby(['Product_Category'], as_index=False)['Amount'].sum().sort_
sns.set(rc={'figure.figsize':(20,5)})
sns.barplot(data = sales_state, x = 'Product_Category',y= 'Amount')
plt.show()
```

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

### **Product ID**

#### In [55]:

```
top_products = df.groupby('Product_ID')['Orders'].sum().nlargest(10).sort_values(asc
fig1, ax1 = plt.subplots(figsize=(12, 7))
ax1.fill_between(top_products.index, top_products.values, color='skyblue', alpha=0.7
ax1.set_xlabel('Product ID')
ax1.set_ylabel('Total Orders')
ax1.set_title('Top 10 Products by Total Orders')
plt.show()
```



## 4 Conclusion:

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

Thank you!