In [100]: import pandas as pd
 import matplotlib.pyplot as plt
 import seaborn as sns
 import os
 from matplotlib import pyplot

Out[101]:

	User_ID	Product_ID	Gender	Age	Occupation	City_Category	Stay_In_Current_City_Years	M
0	1000001	P00069042	F	0- 17	10	А	2	
1	1000001	P00248942	F	0- 17	10	А	2	
2	1000001	P00087842	F	0- 17	10	А	2	
3	1000001	P00085442	F	0- 17	10	А	2	
4	1000002	P00285442	М	55+	16	С	4+	
5	1000003	P00193542	М	26- 35	15	А	3	
6	1000004	P00184942	М	46- 50	7	В	2	
7	1000004	P00346142	M	46- 50	7	В	2	
8	1000004	P0097242	M	46- 50	7	В	2	
9	1000005	P00274942	М	26- 35	20	А	1	

```
In [102]: ap.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 537577 entries, 0 to 537576
          Data columns (total 12 columns):
          User ID
                                         537577 non-null int64
          Product ID
                                         537577 non-null object
          Gender
                                         537577 non-null object
                                         537577 non-null object
          Age
          Occupation
                                         537577 non-null int64
                                         537577 non-null object
          City Category
          Stay In Current City Years
                                         537577 non-null object
          Marital Status
                                         537577 non-null int64
                                         537577 non-null int64
          Product Category 1
          Product Category 2
                                         370591 non-null float64
                                         164278 non-null float64
          Product Category 3
          Purchase
                                         537577 non-null int64
          dtypes: float64(2), int64(5), object(5)
          memory usage: 49.2+ MB
In [103]: | ap.isnull().sum()
Out[103]: User ID
                                               0
          Product ID
                                               0
                                               0
          Gender
          Age
                                               0
                                               0
          Occupation
          City Category
                                               0
          Stay In Current City Years
                                               0
          Marital Status
                                               0
          Product Category 1
                                               0
          Product Category 2
                                         166986
          Product Category 3
                                         373299
          Purchase
                                               0
          dtype: int64
In [104]: ap.columns
Out[104]: Index(['User ID', 'Product ID', 'Gender', 'Age', 'Occupation', 'City C
          ategory',
                  'Stay In Current City Years', 'Marital Status', 'Product Catego
          ry 1',
                  'Product Category 2', 'Product Category 3', 'Purchase'],
                 dtype='object')
```

```
In [110]: ap.sort_values('User_ID').head(10)
#ap['User_ID'].value_counts().count()
```

Out[110]:

	User_ID	Product_ID	Gender	Age	Occupation	City_Category	Stay_In_Current_City_Year
0	1000001	P00069042	F	0- 17	10	А	
390151	1000001	P00255842	F	0- 17	10	А	
390150	1000001	P0097142	F	0- 17	10	А	
350797	1000001	P00289942	F	0- 17	10	А	
311713	1000001	P00210342	F	0- 17	10	А	
311712	1000001	P00248442	F	0- 17	10	А	
311711	1000001	P00051442	F	0- 17	10	А	
311710	1000001	P00183942	F	0- 17	10	А	
311709	1000001	P00178342	F	0- 17	10	А	
467663	1000001	P00058142	F	0- 17	10	А	

```
In [111]: ap['User_ID'].value_counts().count()
```

Out[111]: 5891

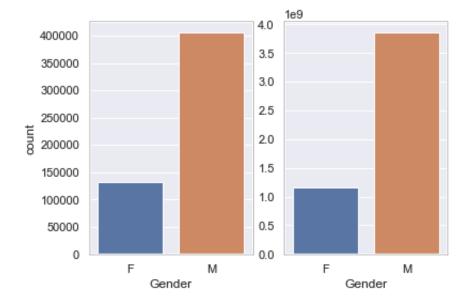
```
In [112]: ap['Gender'].unique()
```

Out[112]: array(['F', 'M'], dtype=object)

```
In [113]: plt.subplot(1,2,1)
    sns.countplot(ap['Gender']) #attendance

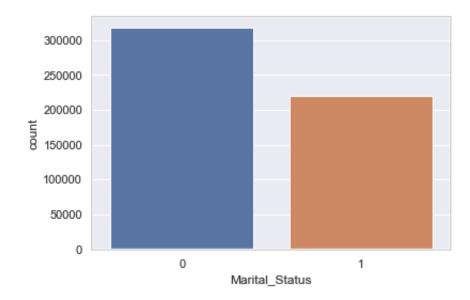
m_purchase = ap.groupby(['Gender'])['Purchase'].sum()
    plt.subplot(1,2,2)
    sns.barplot(m_purchase.index, m_purchase.values) #dollar value
```

Out[113]: <matplotlib.axes._subplots.AxesSubplot at 0x1a2f7b9630>



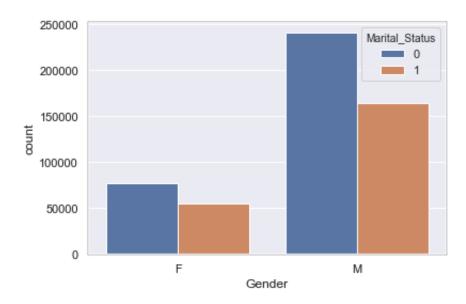
In [114]: sns.countplot(ap['Marital_Status'])

Out[114]: <matplotlib.axes._subplots.AxesSubplot at 0x1a2aa27b00>



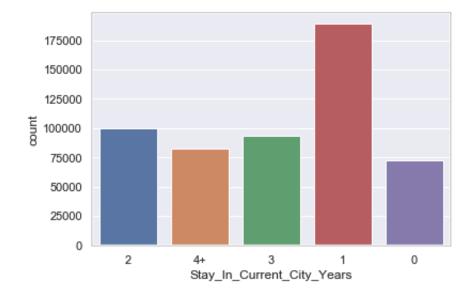
```
In [115]: sns.countplot(ap['Gender'], hue = ap['Marital_Status'])
```

Out[115]: <matplotlib.axes._subplots.AxesSubplot at 0x1a16bc5e80>



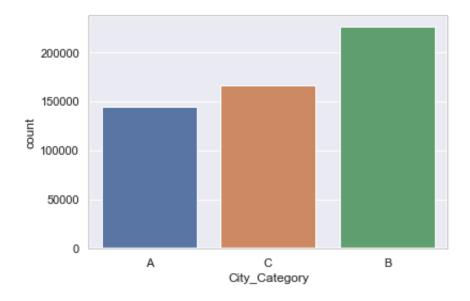
```
In [116]: sns.countplot(ap['Stay_In_Current_City_Years'])
```

Out[116]: <matplotlib.axes._subplots.AxesSubplot at 0x1a32272630>



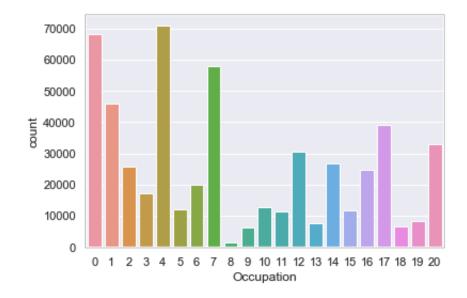
```
In [117]: sns.countplot(ap['City_Category'])
```

Out[117]: <matplotlib.axes._subplots.AxesSubplot at 0x1a322c1160>



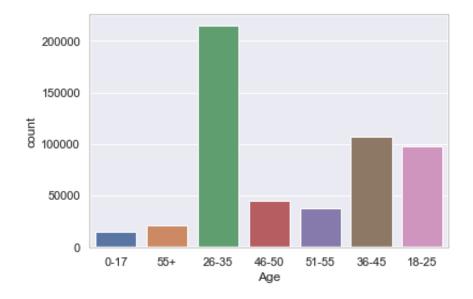
```
In [118]: sns.countplot(ap['Occupation'])
```

Out[118]: <matplotlib.axes._subplots.AxesSubplot at 0x1a1a5bc320>



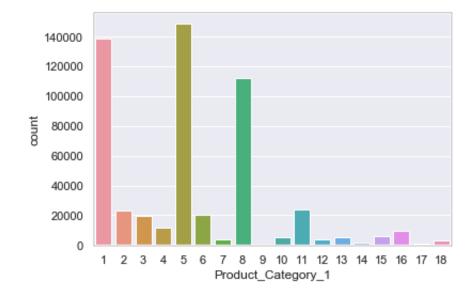
```
In [119]: sns.countplot(ap['Age'])
```

Out[119]: <matplotlib.axes._subplots.AxesSubplot at 0x1a3235bfd0>



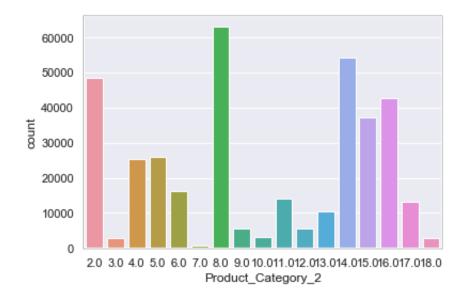
```
In [120]: sns.countplot(ap['Product_Category_1'])
```

Out[120]: <matplotlib.axes._subplots.AxesSubplot at 0x1a32240e48>



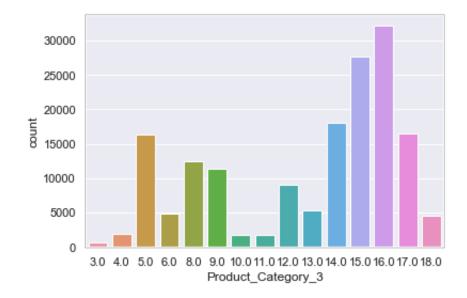
```
In [121]: sns.countplot(ap['Product_Category_2'])
```

Out[121]: <matplotlib.axes._subplots.AxesSubplot at 0x1a2f5aa588>



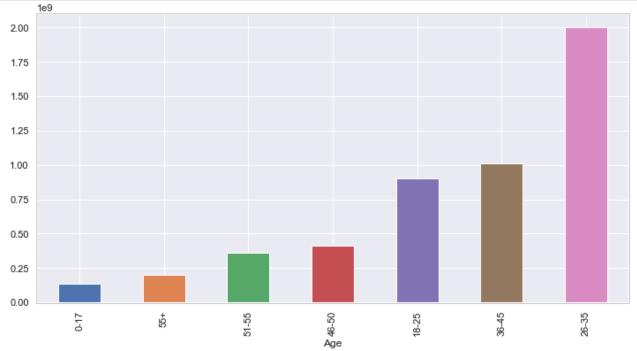
```
In [122]: sns.countplot(ap['Product_Category_3'])
```

Out[122]: <matplotlib.axes._subplots.AxesSubplot at 0x1a322f7668>



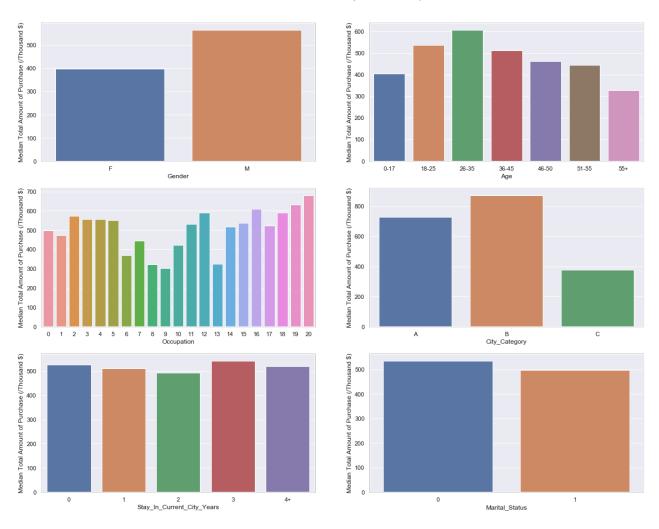
```
In [123]: fig1, ax1 = plt.subplots(figsize=(12,7))
    sns.countplot(ap['Age'],hue=ap['Gender'])

def plot(group,column,plot):
    ax=plt.figure(figsize=(12,6))
    ap.groupby(group)[column].sum().sort_values().plot(plot)
    plot('Age','Purchase','bar')
```



```
In [124]: # Bar charts - show median instead of mean of total amount of purchase b
          import numpy as np
          fig5, axes = plt.subplots(3,2,figsize=(20,16))
          fig5.suptitle('Median Amount of Purchase by Customer Groups', fontsize =
          sns.barplot(x='Gender', y='Tot_Purchase', data = ap_customer, estimator
          sns.barplot(x='Age', y='Tot Purchase', data = ap customer, estimator = n
                      ax = axes[0][1], order = ['0-17', '18-25', '26-35', '36-45',
          sns.barplot(x='Occupation', y='Tot Purchase', data = ap customer, estimate
          sns.barplot(x='City_Category', y='Tot_Purchase', data = ap customer, est
                      ci = None, ax = axes[1][1], order = ('A', 'B', 'C')
          sns.barplot(x='Stay In Current City Years', y='Tot Purchase', data = ap
                      ci = None, ax = axes[2][0], order = ('0', '1', '2', '3', '4+
          sns.barplot(x='Marital Status', y='Tot Purchase', data = ap customer, es
          for ax in fig5.axes:
              plt.sca(ax)
              plt.ylabel('Median Total Amount of Purchase (/Thousand $)')
          plt.savefig('fig5')
```

Median Amount of Purchase by Customer Groups

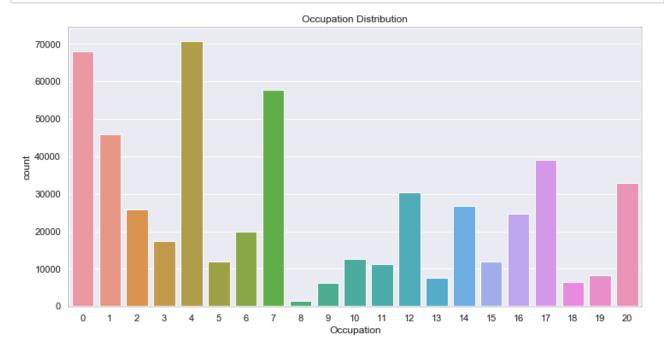


In [143]: # Generate new features - total # products purchased by customer; and to
tot_item = ap['User_ID'].value_counts().sort_index()
tot_purchase = ap.groupby('User_ID').sum()['Purchase']
tot = pd.concat([tot_item, tot_purchase], axis = 1, keys = ['Tot_Product
ap = pd.merge(ap, tot, left_on = 'User_ID', right_index = True)
ap.head()

Out[143]:

 Tot_Products_x	Tot_Purchase_x	Tot_Products_y	Tot_Purchase_y	Tot_Products_x	Tot_Purchase_
 34	333481	34	333481	34	33348
 34	333481	34	333481	34	33348
 34	333481	34	333481	34	33348
 34	333481	34	333481	34	33348
 34	333481	34	333481	34	33348

```
In [144]: #Occupation
    plt.figure(figsize=(12,6))
    sns.countplot(ap['Occupation'])
    plt.title('Occupation Distribution')
    plt.show()
```



Question:- 1) We need to Analysis the data based on the multiple variables on which the purchase is dependent. So we will analyze all variable such as Gender, Age, Occupation, City in which they Stay, Product Category from which they shop, Marital Status. 2) Once we study these small variable we will come to know the impact of each Variable on the purchase Level. Now from this we need to analyze on each user to check what was their total purchase. As this will give us a view of their total purchasing power. Now Analzing each variable again we will come to know which of them impact the most to affect the purchasing power of each Individual.

Question What we Exactly want to Predict :- 1) We want to check the purchasing power of user once they login with their Details. 2) We want to predict the customer recommendation based on product Category for each new user once we have their details. 3) Inbound goods prediction based on sale for each city category.

```
In [ ]:
```