Mini Project

```
In [1]: import pandas as pd
    import numpy as np
    import os
    import seaborn as sns
    import matplotlib.pyplot as plt
    %matplotlib inline
```

1. Merge the files in one dataframe

```
In [2]:
    files = [file for file in os.listdir(r'C:\Users\keerti chouhan\Desktop\sales')
    sales_df = pd.DataFrame()
    for file in files:
        df = pd.read_csv('C:/Users/keerti chouhan/Desktop/sales/'+file)
        sales_df = pd.concat([sales_df,df])
```

In [3]: sales_df.head()

Out[3]:

	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address
0	176558	USB-C Charging Cable	2	11.95	04/19/19 8:46	917 1st St, Dallas, TX 75001
1	NaN	NaN	NaN	NaN	NaN	NaN
2	176559	Bose SoundSport Headphones	1	99.99	04/07/19 22:30	682 Chestnut St, Boston, MA 02215
3	176560	Google Phone	1	600	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001
4	176560	Wired Headphones	1	11.99	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001

In [4]: sales_df.tail()

Out[4]:

	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address
11681	259353	AAA Batteries (4-pack)	3	2.99	09/17/19 20:56	840 Highland St, Los Angeles, CA 90001
11682	259354	iPhone	1	700	09/01/19 16:00	216 Dogwood St, San Francisco, CA 94016
11683	259355	iPhone	1	700	09/23/19 7:39	220 12th St, San Francisco, CA 94016
11684	259356	34in Ultrawide Monitor	1	379.99	09/19/19 17:30	511 Forest St, San Francisco, CA 94016
11685	259357	USB-C Charging Cable	1	11.95	09/30/19 0:18	250 Meadow St, San Francisco, CA 94016

In [5]: sales_df.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 186850 entries, 0 to 11685
Data columns (total 6 columns):

Column # Non-Null Count Dtype ____ _____ Order ID 186305 non-null object 0 186305 non-null object 1 Product 2 Quantity Ordered 186305 non-null object 3 Price Each 186305 non-null object 4 Order Date 186305 non-null object

Purchase Address 186305 non-null object

dtypes: object(6)
memory usage: 10.0+ MB

2.Clean the data.

In [6]: sales_df.info()

5

<class 'pandas.core.frame.DataFrame'>
Int64Index: 186850 entries, 0 to 11685
Data columns (total 6 columns):

#	Column	Non-Null Count	Dtype
0	Order ID	186305 non-null	object
1	Product	186305 non-null	object
2	Quantity Ordered	186305 non-null	object
3	Price Each	186305 non-null	object
4	Order Date	186305 non-null	object
5	Purchase Address	186305 non-null	object

dtypes: object(6)
memory usage: 10.0+ MB

In [7]: sales_df.isna().sum()

Out[7]: Order ID 545 Product 545

Quantity Ordered 545 Price Each 545 Order Date 545

Purchase Address 545

dtype: int64

In [8]: sales_df=sales_df.dropna()

In [9]: sales_df

Out[9]:

Order ID		Product	Quantity Ordered	Price Each	Order Date	Purchase Address
0	176558	USB-C Charging Cable	2	11.95	04/19/19 8:46	917 1st St, Dallas, TX 75001
2	176559	Bose SoundSport Headphones	1	99.99	04/07/19 22:30	682 Chestnut St, Boston, MA 02215
3	176560	Google Phone	1	600	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001
4	176560	Wired Headphones	1	11.99	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001
5	176561	Wired Headphones	1	11.99	04/30/19 9:27	333 8th St, Los Angeles, CA 90001
11681	259353	AAA Batteries (4- pack)	3	2.99	09/17/19 20:56	840 Highland St, Los Angeles, CA 90001
11682	259354	iPhone	1	700	09/01/19 16:00	216 Dogwood St, San Francisco, CA 94016
11683	259355	iPhone	1	700	09/23/19 7:39	220 12th St, San Francisco, CA 94016
11684	259356	34in Ultrawide Monitor	1	379.99	09/19/19 17:30	511 Forest St, San Francisco, CA 94016
11685	259357	USB-C Charging Cable	1	11.95	09/30/19 0:18	250 Meadow St, San Francisco, CA 94016

186305 rows × 6 columns

```
In [10]: sales df.isna().sum()
Out[10]: Order ID
                             0
         Product
                             0
         Quantity Ordered
                             0
         Price Each
                             0
         Order Date
                             0
         Purchase Address
                             0
         dtype: int64
In [11]: sales_df.info()
         <class 'pandas.core.frame.DataFrame'>
         Int64Index: 186305 entries, 0 to 11685
         Data columns (total 6 columns):
              Column
                                Non-Null Count
                                                 Dtype
              ----
                                -----
                                                 ----
              Order ID
          0
                                186305 non-null object
          1
              Product
                                186305 non-null object
          2
              Quantity Ordered 186305 non-null
                                                object
          3
              Price Each
                                186305 non-null
                                                 object
          4
              Order Date
                                186305 non-null
                                                 object
          5
              Purchase Address 186305 non-null object
         dtypes: object(6)
         memory usage: 9.9+ MB
```

3. Change the object type column into integer type or float type.

```
In [12]: sales df[sales df["Price Each"]=="Price Each"].index
Out[12]: Int64Index([ 519,
                             1149, 1155, 2878, 2893, 3036, 3209,
                                                                       3618, 4138,
                      4645,
                     . . .
                             9325, 9502, 9615, 9954, 10000, 10387, 11399, 11468,
                      8644,
                     11574],
                    dtype='int64', length=355)
In [13]:
         sales df.drop(sales df[sales df["Price Each"]=="Price Each"].index,inplace=Tru
In [14]: | sales_df[sales_df["Price Each"]=="Price Each"].index
Out[14]: Int64Index([], dtype='int64')
In [15]: sales_df["Price Each"]=sales_df["Price Each"].astype('float64')
In [16]: sales df["Order ID"]=sales df["Order ID"].astype('int64')
```

```
In [17]: sales df["Quantity Ordered"]=sales df["Quantity Ordered"].astype('int32')
In [18]: sales_df.info()
         <class 'pandas.core.frame.DataFrame'>
         Int64Index: 182735 entries, 0 to 11685
         Data columns (total 6 columns):
              Column
                               Non-Null Count Dtype
         ---
             -----
                               -----
                                               ----
             Order ID 182735 non-null int64
Product 182735 non-null object
          0
            Product
          1
          2
              Quantity Ordered 182735 non-null int32
             Price Each 182735 non-null float64
          3
             Order Date 182735 non-null object
              Purchase Address 182735 non-null object
          5
         dtypes: float64(1), int32(1), int64(1), object(3)
         memory usage: 9.1+ MB
```

4.Get the month value from the order date?

```
In [19]:
         import datetime
         month=pd.to_datetime(sales_df['Order Date']).dt.month
         month
Out[19]: 0
                  4
         2
                   4
         3
                   4
         4
         5
                  4
         11681
         11682
                  9
                  9
         11683
         11684
                  9
                  9
         11685
         Name: Order Date, Length: 182735, dtype: int64
```

In [20]: sales_df['month']=month
 sales_df

Out[20]:

	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address	month
0	176558	USB-C Charging Cable	2	11.95	04/19/19 8:46	917 1st St, Dallas, TX 75001	4
2	176559	Bose SoundSport Headphones	1	99.99	04/07/19 22:30	682 Chestnut St, Boston, MA 02215	4
3	176560	Google Phone	1	600.00	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001	4
4	176560	Wired Headphones	1	11.99	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001	4
5	176561	Wired Headphones	1	11.99	04/30/19 9:27	333 8th St, Los Angeles, CA 90001	4
11681	259353	AAA Batteries (4- pack)	3	2.99	09/17/19 20:56	840 Highland St, Los Angeles, CA 90001	9
11682	259354	iPhone	1	700.00	09/01/19 16:00	216 Dogwood St, San Francisco, CA 94016	9
11683	259355	iPhone	1	700.00	09/23/19 7:39	220 12th St, San Francisco, CA 94016	9
11684	259356	34in Ultrawide Monitor	1	379.99	09/19/19 17:30	511 Forest St, San Francisco, CA 94016	9
11685	259357	USB-C Charging Cable	1	11.95	09/30/19 0:18	250 Meadow St, San Francisco, CA 94016	9

182735 rows × 7 columns

5. Which was the most productive month in terms of sales?

In [21]: sales_df['Sales']=sales_df['Quantity Ordered']*sales_df['Price Each']

```
In [22]: sales_df.head()
```

Out[22]:

	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address	month	Sales
0	176558	USB-C Charging Cable	2	11.95	04/19/19 8:46	917 1st St, Dallas, TX 75001	4	23.90
2	176559	Bose SoundSport Headphones	1	99.99	04/07/19 22:30	682 Chestnut St, Boston, MA 02215	4	99.99
3	176560	Google Phone	1	600.00	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001	4	600.00
4	176560	Wired Headphones	1	11.99	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001	4	11.99
5	176561	Wired Headphones	1	11.99	04/30/19 9:27	333 8th St, Los Angeles, CA 90001	4	11.99

```
In [23]: a=sales_df.groupby('month')["Sales"].sum().sort_values(ascending=False)
a
```

Out[23]: month

- 12 4557905.42
- 10 3679254.16
- 4 3336376.42
- 11 3149785.09
- 5 3101881.04
- 3 2755969.40
- 7 2587444.91
- 6 2524464.99
- 8 2191698.31
- 2 2158127.48
- 9 2050361.26
- 1 1786511.29

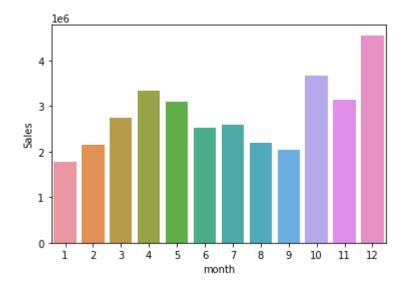
Name: Sales, dtype: float64

Out[24]:

	month	Sales
11	12	4557905.42
9	10	3679254.16
3	4	3336376.42
10	11	3149785.09
4	5	3101881.04

```
In [25]: sns.barplot(x='month',y='Sales',data=c_data,ci=None)
```

Out[25]: <AxesSubplot:xlabel='month', ylabel='Sales'>



The most productive month in terms of sales is December.

6. Which city had the highest number of sales?

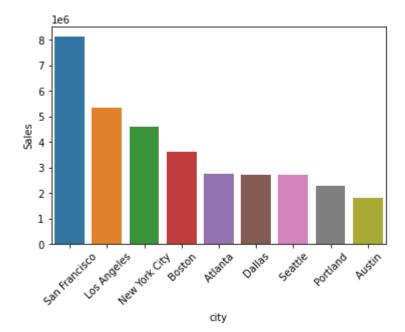
Out[26]:

	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address	month	Sales	city
0	176558	USB-C Charging Cable	2	11.95	04/19/19 8:46	917 1st St, Dallas, TX 75001	4	23.90	Dallas
2	176559	Bose SoundSport Headphones	1	99.99	04/07/19 22:30	682 Chestnut St, Boston, MA 02215	4	99.99	Boston
3	176560	Google Phone	1	600.00	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001	4	600.00	Los Angeles
4	176560	Wired Headphones	1	11.99	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001	4	11.99	Los Angeles
5	176561	Wired Headphones	1	11.99	04/30/19 9:27	333 8th St, Los Angeles, CA 90001	4	11.99	Los Angeles

Out[27]:

```
citySales7San Francisco8124120.944Los Angeles5354039.935New York City4581658.912Boston3604080.860Atlanta2741642.05
```

```
In [28]: sns.barplot(x='city',y='Sales',data=c_data,ci=None)
plt.xticks(rotation=45);
```



The city had the highest number of sales is: San Francisco

7.At what time people mostly purchase the product?

```
In [29]: hour= pd.to_datetime(sales_df['Order Date']).dt.hour
         hour
Out[29]: 0
                   8
                  22
         2
         3
                  14
         4
                  14
         5
                   9
         11681
                  20
         11682
                  16
         11683
                  7
                  17
         11684
         11685
                   0
         Name: Order Date, Length: 182735, dtype: int64
In [30]: sales_df['hour']=hour
         sales_df.head()
```

Out[30]:

	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address	month	Sales	city	hour
0	176558	USB-C Charging Cable	2	11.95	04/19/19 8:46	917 1st St, Dallas, TX 75001	4	23.90	Dallas	8
2	176559	Bose SoundSport Headphones	1	99.99	04/07/19 22:30	682 Chestnut St, Boston, MA 02215	4	99.99	Boston	22
3	176560	Google Phone	1	600.00	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001	4	600.00	Los Angeles	14
4	176560	Wired Headphones	1	11.99	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001	4	11.99	Los Angeles	14
5	176561	Wired Headphones	1	11.99	04/30/19 9:27	333 8th St, Los Angeles, CA 90001	4	11.99	Los Angeles	9

```
In [37]: k=sales_df.groupby(by='hour' ,as_index=False)['Quantity Ordered'].sum()
    k.sort_values(by='Quantity Ordered' ,ascending=False)
```

Out[37]:

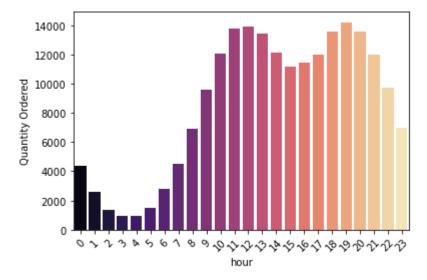
	hour	Quantity Ordered
19	19	14228
12	12	13946
11	11	13763
18	18	13576
20	20	13565
13	13	13437
14	14	12163
10	10	12058
21	21	12030
17	17	12016
16	16	11420
15	15	11175
22	22	9715
9	9	9628
23	23	6955
8	8	6917
7	7	4483
0	0	4355
6	6	2767
1	1	2579
5	5	1463
2	2	1379
4	4	925
3	3	912

```
In [31]: sales_df['hour'].value_counts().head()
```

```
Out[31]: 19 12685
12 12360
11 12202
18 12074
20 12040
```

Name: hour, dtype: int64

```
In [42]: sns.barplot(x='hour',y='Quantity Ordered',data=k,palette='magma')
plt.xticks(rotation=45);
```



The Time in which people mostly purchase the product is between (18-21)

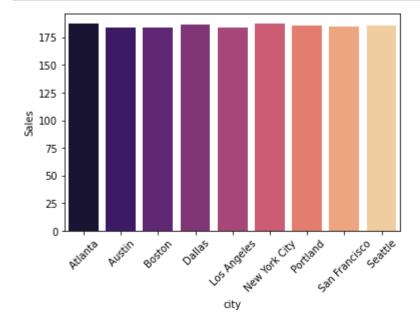
8What is the average purchase by city?

```
In [48]: g=sales_df.groupby(by='city' ,as_index=False)['Sales'].mean()
g.sort_values(by='Sales' ,ascending=False)
```

Out[48]:

	city	Sales
0	Atlanta	187.578137
5	New York City	187.342939
3	Dallas	186.520741
8	Seattle	185.894153
6	Portland	185.682183
7	San Francisco	184.857580
2	Boston	184.116519
4	Los Angeles	183.943379
1	Austin	183.935096

```
In [50]: sns.barplot(x='city',y='Sales',data=g,palette='magma')
plt.xticks(rotation=45);
```



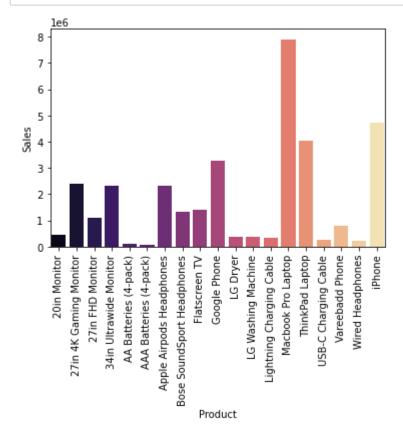
9. Which product has the highest sales?

```
In [53]: g=sales_df.groupby(by='Product' ,as_index=False)['Sales'].sum()
g.sort_values(by='Sales' ,ascending=False)
```

Out[53]:

	Product	Sales
13	Macbook Pro Laptop	7896500.00
18	iPhone	4712400.00
14	ThinkPad Laptop	4053959.46
9	Google Phone	3264000.00
1	27in 4K Gaming Monitor	2392198.66
3	34in Ultrawide Monitor	2308819.24
6	Apple Airpods Headphones	2307450.00
8	Flatscreen TV	1417200.00
7	Bose SoundSport Headphones	1323467.64
2	27in FHD Monitor	1114275.71
16	Vareebadd Phone	809200.00
0	20in Monitor	446339.42
11	LG Washing Machine	389400.00
10	LG Dryer	384000.00
12	Lightning Charging Cable	341472.95
15	USB-C Charging Cable	281482.25
17	Wired Headphones	242209.99
4	AA Batteries (4-pack)	104248.32
5	AAA Batteries (4-pack)	91156.13

```
In [58]: sns.barplot(x='Product',y='Sales',data=g,palette='magma');
plt.xticks(rotation=90);
```



Macbook Pro Laptop product has the highest sales

10. In Month of September, which product has the lowest sales?

```
In [63]: sept=sales_df[sales_df['month']== 9]
sept
```

	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address	month	Sales	city	hοι
2251	238834	Apple Airpods Headphones	1	150.00	09/01/19 4:13	761 Forest St, San Francisco, CA 94016	9	150.00	San Francisco	
2731	239285	34in Ultrawide Monitor	1	379.99	09/01/19 1:09	373 1st St, San Francisco, CA 94016	9	379.99	San Francisco	
4124	240636	Lightning Charging Cable	1	14.95	09/01/19 2:07	63 1st St, Seattle, WA 98101	9	14.95	Seattle	
4569	241054	AAA Batteries (4- pack)	1	2.99	09/01/19 0:25	175 South St, San Francisco, CA 94016	9	2.99	San Francisco	
5914	242343	ThinkPad Laptop	1	999.99	09/01/19 2:44	510 Park St, Boston, MA 02215	9	999.99	Boston	
11681	259353	AAA Batteries (4- pack)	3	2.99	09/17/19 20:56	840 Highland St, Los Angeles, CA 90001	9	8.97	Los Angeles	2
11682	259354	iPhone	1	700.00	09/01/19 16:00	216 Dogwood St, San Francisco, CA 94016	9	700.00	San Francisco	1
11683	259355	iPhone	1	700.00	09/23/19 7:39	220 12th St, San Francisco, CA 94016	9	700.00	San Francisco	
11684	259356	34in Ultrawide Monitor	1	379.99	09/19/19 17:30	511 Forest St, San Francisco, CA 94016	9	379.99	San Francisco	1
11685	259357	USB-C Charging Cable	1	11.95	09/30/19 0:18	250 Meadow St, San Francisco, CA 94016	9	11.95	San Francisco	
44275 rawa w 40 aakurana										

11375 rows × 10 columns

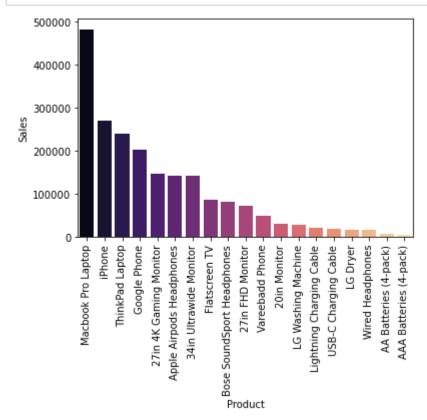
4

```
In [64]: g=sept.groupby(by='Product' ,as_index=False)['Sales'].sum()
t=g.sort_values(by='Sales' ,ascending=False)
t
```

Out[64]:

	Product	Sales
13	Macbook Pro Laptop	482800.00
18	iPhone	269500.00
14	ThinkPad Laptop	239997.60
9	Google Phone	201600.00
1	27in 4K Gaming Monitor	146246.25
6	Apple Airpods Headphones	142500.00
3	34in Ultrawide Monitor	140976.29
8	Flatscreen TV	85500.00
7	Bose SoundSport Headphones	80891.91
2	27in FHD Monitor	71095.26
16	Vareebadd Phone	48000.00
0	20in Monitor	29587.31
11	LG Washing Machine	27600.00
12	Lightning Charging Cable	21049.60
15	USB-C Charging Cable	18713.70
10	LG Dryer	16800.00
17	Wired Headphones	15251.28
4	AA Batteries (4-pack)	6600.96
5	AAA Batteries (4-pack)	5651.10

```
In [66]: sns.barplot(x='Product',y='Sales',data=t,palette='magma')
plt.xticks(rotation=90);
```



In month of sept, AAA Batteries (4-Pack) has lowest sale

```
In [ ]: Question -Answers
```

5. Which was the most productive month in terms of sales?

December was the most productive month in terms of sales

6. Which city had the highest number of sales?

San Francisco has the highest number of sales

7. At what time people mostly purchase the product?

At 19:00 people mostly purchase the product

8. What is the average purchase by city?

Atlanta has the maximum average purchase by the cities.

9. Which product has the highest sales?

Macbook Pro Laptop has the highest sales.

10. In Month of September, which product has the lowest sales?

In month of september AAA Batteries (4-pack) has the lowest sale.

In []:	
In []:	
In []:	
In []:	
In []:	