Sales Analysis

Libraries

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import os
import warnings

//matplotlib inline
warnings.filterwarnings('ignore')
```

Merging 12 Months Data

```
In [2]: files = [file for file in os.listdir('Sales_Data')]
    all_months_data = pd.DataFrame()
    for file in files:
        df = pd.read_csv('Sales_Data/'+file)
        all_months_data = pd.concat([all_months_data, df])
    all_months_data.to_csv('all_data.csv', index=False)
```

Read and Cleaning Updated Data Frame

```
In [3]: all_data = pd.read_csv('all_data.csv')
    all_data.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 08:46	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

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 186850 entries, 0 to 186849
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: 8.6+ MB

memory usage: 9.9+ MB

```
In [6]: #Cleaning Null Rows
        all data = all data.dropna()
In [7]: all_data.info()
        <class 'pandas.core.frame.DataFrame'>
        Int64Index: 186305 entries, 0 to 186849
        Data columns (total 6 columns):
         #
             Column
                              Non-Null Count
                                               Dtype
                              186305 non-null object
         0
             Order ID
             Product
                              186305 non-null object
         1
             Quantity Ordered 186305 non-null object
             Price Each
                              186305 non-null object
             Order Date
                              186305 non-null object
             Purchase Address 186305 non-null object
        dtypes: object(6)
```

In [8]: #Number column still read as object, means there are non null data need to be cleaning
#Check if the are any rows containing header data
all_data.loc[all_data['Order ID'] == 'Order ID']

Out[8]:

	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address
519	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address
1149	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address
1155	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address
2878	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address
2893	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address
185164	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address
185551	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address
186563	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address
186632	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address
186738	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address

355 rows × 6 columns

```
In [9]: #Deleting all rows that contain header string
```

```
all_data = all_data.loc[all_data['Order ID'] != 'Order ID']
```

```
In [10]: all data.info()
         <class 'pandas.core.frame.DataFrame'>
         Int64Index: 185950 entries, 0 to 186849
         Data columns (total 6 columns):
              Column
                               Non-Null Count
                                                Dtvpe
              -----
                                -----
              Order ID
                               185950 non-null object
          1
              Product
                               185950 non-null object
              Ouantity Ordered 185950 non-null object
              Price Each
                               185950 non-null object
                               185950 non-null object
              Order Date
              Purchase Address 185950 non-null object
         dtypes: object(6)
         memory usage: 9.9+ MB
In [11]: #Change Data Type to Selected Columns
         all data[['Order ID','Quantity Ordered']] = all data[['Order ID','Quantity Ordered']].astype('int')
         all data['Price Each'] = all data['Price Each'].astype('float')
In [12]: all data.info()
         <class 'pandas.core.frame.DataFrame'>
         Int64Index: 185950 entries, 0 to 186849
         Data columns (total 6 columns):
              Column
                               Non-Null Count
                                                Dtype
                                -----
              Order ID
                               185950 non-null int32
          1
              Product
                               185950 non-null object
              Quantity Ordered 185950 non-null int32
          2
                               185950 non-null float64
              Price Each
          4
              Order Date
                               185950 non-null object
              Purchase Address 185950 non-null object
         dtypes: float64(1), int32(2), object(3)
         memory usage: 8.5+ MB
In [13]: #Reindexing
         all data.index = range(len(all data))
```

```
In [14]: all_data.info()
    all_data.head()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 185950 entries, 0 to 185949
Data columns (total 6 columns):

#	Column	Non-Null Count	Dtype
0	Order ID	185950 non-null	int32
1	Product	185950 non-null	object
2	Quantity Ordered	185950 non-null	int32
3	Price Each	185950 non-null	float64
4	Order Date	185950 non-null	object
5	Purchase Address	185950 non-null	object
d+vn	oc. float64(1) in	+22/2\ obioc+/2\	

dtypes: float64(1), int32(2), object(3)

memory usage: 7.1+ MB

Out[14]:	Order ID		Product	Quantity Ordered	Price Each	Order Date	Purchase Address
	0	176558	USB-C Charging Cable	2	11.95	04/19/19 08:46	917 1st St, Dallas, TX 75001
	1	176559	Bose SoundSport Headphones	1	99.99	04/07/19 22:30	682 Chestnut St, Boston, MA 02215
	2	176560	Google Phone	1	600.00	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001
	3	176560	Wired Headphones	1	11.99	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001

1

11.99 04/30/19 09:27

Augment Data With Additionnal Columns

Wired Headphones

Add Month Column

176561

333 8th St, Los Angeles, CA 90001

```
In [15]: all_data['Month'] = all_data['Order Date'].str[0:2]
    all_data['Month'] = all_data['Month'].astype('int32')
    all_data.head()
```

Out	[15]]:

	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address	Month
0	176558	USB-C Charging Cable	2	11.95	04/19/19 08:46	917 1st St, Dallas, TX 75001	4
1	176559	Bose SoundSport Headphones	1	99.99	04/07/19 22:30	682 Chestnut St, Boston, MA 02215	4
2	176560	Google Phone	1	600.00	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001	4
3	176560	Wired Headphones	1	11.99	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001	4
4	176561	Wired Headphones	1	11.99	04/30/19 09:27	333 8th St, Los Angeles, CA 90001	4

Add Sales Column

```
In [16]: all_data['Sales'] = all_data['Quantity Ordered'] * all_data['Price Each']
all_data.head()
```

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	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 08:46	917 1st St, Dallas, TX 75001	4	23.90
1	176559	Bose SoundSport Headphones	1	99.99	04/07/19 22:30	682 Chestnut St, Boston, MA 02215	4	99.99
2	176560	Google Phone	1	600.00	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001	4	600.00
3	176560	Wired Headphones	1	11.99	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001	4	11.99
4	176561	Wired Headphones	1	11.99	04/30/19 09:27	333 8th St, Los Angeles, CA 90001	4	11.99

Add City Column

```
In [22]: def get_city(address):
    return address.split(',')[1]

def get_state(address):
    return address.split(',')[2].split(' ')[1]

all_data['City'] = all_data['Purchase Address'].apply(lambda x : f'{get_city(x)} ({get_state(x)})')

all_data.head()
```

Out[22]:

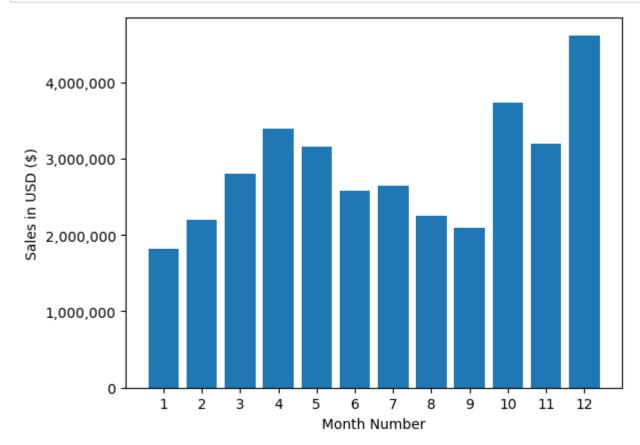
	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address	Month	Sales	City
0	176558	USB-C Charging Cable	2	11.95	04/19/19 08:46	917 1st St, Dallas, TX 75001	4	23.90	Dallas (TX)
1	176559	Bose SoundSport Headphones	1	99.99	04/07/19 22:30	682 Chestnut St, Boston, MA 02215	4	99.99	Boston (MA)
2	176560	Google Phone	1	600.00	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001	4	600.00	Los Angeles (CA)
3	176560	Wired Headphones	1	11.99	04/12/19 14:38	669 Spruce St, Los Angeles, CA 90001	4	11.99	Los Angeles (CA)
4	176561	Wired Headphones	1	11.99	04/30/19 09:27	333 8th St, Los Angeles, CA 90001	4	11.99	Los Angeles (CA)

Analysis

Question 1: What was the best month for sales? How much was earned the month?

In [23]: | all_data.groupby('Month').sum().sort_values(by='Sales', ascending=False) Out[23]: Order ID Quantity Ordered Price Each Sales Month **12** 7.685905e+09 28114 4588415.41 4613443.34 **10** 5.457110e+09 3715554.83 3736726.88 3.387347e+09 20558 3367671.02 3390670.24 **11** 5.047203e+09 19798 3180600.68 3199603.20 **5** 3.345872e+09 18667 3135125.13 3152606.75 2.564811e+09 2791207.83 2807100.38 17005 **7** 3.284140e+09 16072 2632539.56 2647775.76 6 2.932976e+09 15253 2562025.61 2577802.26 2.899374e+09 2230345.42 2244467.88 13448 **2** 1.871053e+09 13449 2188884.72 2202022.42 9 2.948727e+09 2084992.09 2097560.13 13109

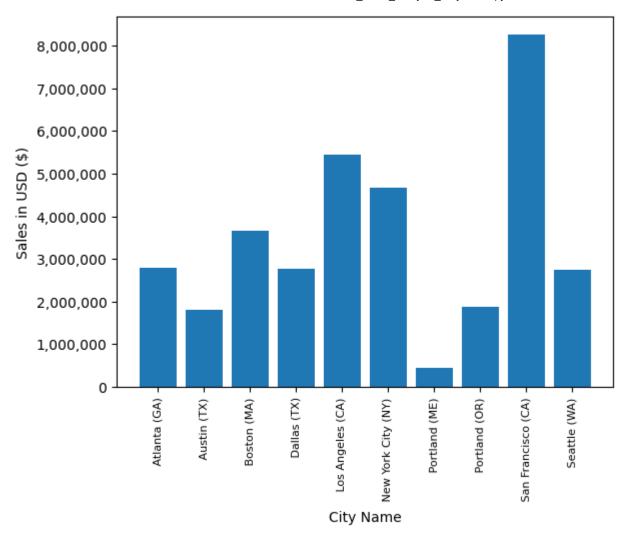
Answer 1: Month 12th (December)



Question 2: What city had the highest number of sales?

[34]:	all_data.groupby	(' <mark>City</mark> ').sum	n().sort_values	(by="Sales	", asce	nding= Fals
4]:		Order ID	Quantity Ordered	Price Each	Month	Sales
	City					
	San Francisco (CA)	1.030444e+10	50239	8211461.74	315520	8262203.91
	Los Angeles (CA)	6.811085e+09	33289	5421435.23	208325	5452570.80
	New York City (NY)	5.736334e+09	27932	4635370.83	175741	4664317.43
	Boston (MA)	4.598265e+09	22528	3637409.77	141112	3661642.01
	Atlanta (GA)	3.423838e+09	16602	2779908.20	104794	2795498.58
	Dallas (TX)	3.415644e+09	16730	2752627.82	104620	2767975.40
	Seattle (WA)	3.406694e+09	16553	2733296.01	104941	2747755.48
	Portland (OR)	2.305595e+09	11303	1860558.22	70621	1870732.34
	Austin (TX)	2.280982e+09	11153	1809873.61	69829	1819581.75
	Portland (ME)	5.632663e+08	2750	447189.25	17144	449758.27

Answer 2 : San Fransisco (CA)



Question 3: What time should we display advertisements to maximize likelihood of costumer's buying product?

```
In [55]: all_data['Order Date'] = pd.to_datetime(all_data['Order Date']) #Convert data type to date time
```

```
In [61]: all_data['Hour'] = all_data['Order Date'].dt.hour
    all_data['Minute'] = all_data['Order Date'].dt.minute
    all_data.head()
```

Out[61]:

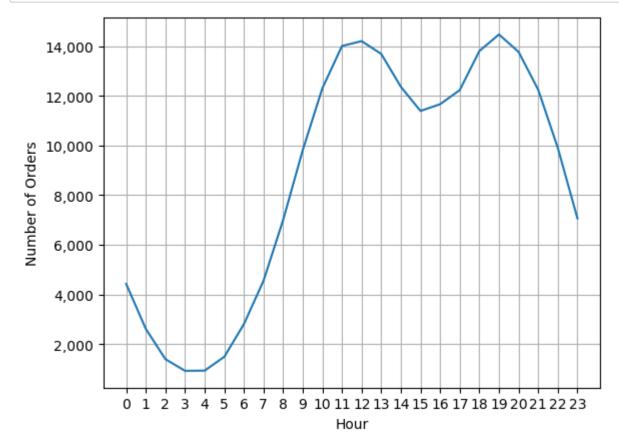
•		Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address	Month	Sales	City	Hour	Minute
	0	176558	USB-C Charging Cable	2	11.95	2019-04-19 08:46:00	917 1st St, Dallas, TX 75001	4	23.90	Dallas (TX)	8	46
	1	176559	Bose SoundSport Headphones	1	99.99	2019-04-07 22:30:00	682 Chestnut St, Boston, MA 02215	4	99.99	Boston (MA)	22	30
	2	176560	Google Phone	1	600.00	2019-04-12 14:38:00	669 Spruce St, Los Angeles, CA 90001	4	600.00	Los Angeles (CA)	14	38
	3	176560	Wired Headphones	1	11.99	2019-04-12 14:38:00	669 Spruce St, Los Angeles, CA 90001	4	11.99	Los Angeles (CA)	14	38
	4	176561	Wired Headphones	1	11.99	2019-04-30 09:27:00	333 8th St, Los Angeles, CA 90001	4	11.99	Los Angeles (CA)	9	27

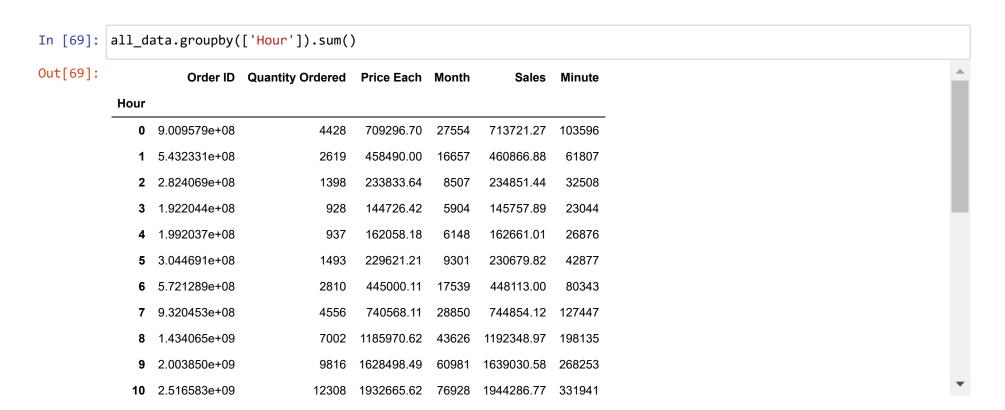
Answer 3 : Peak of sales happened in 12am (12:00) and 7pm (19:00), recommendation sales effective time is from 9am (9:00) to 10pm (22:00)

```
In [80]: hours = [hour for hour, x in all_data.groupby('Hour')]
    #Using Line chart instead of bar chart

plt.plot(hours, all_data.groupby('Hour').sum()['Quantity Ordered'])
    plt.xticks(hours)
    current_values = plt.gca().get_yticks()
    plt.gca().set_yticklabels(['{:,.0f}'.format(x) for x in current_values]) #To add separator comma in thosands
    plt.xlabel('Hour')
    plt.ylabel('Number of Orders')
    plt.grid()

plt.show()
```





Question 4: What products are most often sold together?

In [190]: all_data.head()

Out[190]:

	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address	Month	Sales	City	Hour	Minute
0	176558	USB-C Charging Cable	2	11.95	2019-04-19 08:46:00	917 1st St, Dallas, TX 75001	4	23.90	Dallas (TX)	8	46
1	176559	Bose SoundSport Headphones	1	99.99	2019-04-07 22:30:00	682 Chestnut St, Boston, MA 02215	4	99.99	Boston (MA)	22	30
2	176560	Google Phone	1	600.00	2019-04-12 14:38:00	669 Spruce St, Los Angeles, CA 90001	4	600.00	Los Angeles (CA)	14	38
3	176560	Wired Headphones	1	11.99	2019-04-12 14:38:00	669 Spruce St, Los Angeles, CA 90001	4	11.99	Los Angeles (CA)	14	38
4	176561	Wired Headphones	1	11.99	2019-04-30 09:27:00	333 8th St, Los Angeles, CA 90001	4	11.99	Los Angeles (CA)	9	27

```
In [213]: #find duplicated order id to check which product ordered together

df = all_data[all_data['Order ID'].duplicated(keep=False)]

df['Grouped Products'] = df.groupby('Order ID')['Product'].transform(lambda x: ' , '.join(x))

df = df[['Order ID','Grouped Products']].drop_duplicates()

df.head(20)
```

Out[213]:

	Order ID	Grouped Products
2	176560	Google Phone , Wired Headphones
17	176574	Google Phone , USB-C Charging Cable
29	176585	Bose SoundSport Headphones , Bose SoundSport H
31	176586	AAA Batteries (4-pack) , Google Phone
118	176672	Lightning Charging Cable , USB-C Charging Cable
128	176681	Apple Airpods Headphones , ThinkPad Laptop
137	176689	Bose SoundSport Headphones , AAA Batteries (4
188	176739	34in Ultrawide Monitor , Google Phone
224	176774	Lightning Charging Cable , USB-C Charging Cable
232	176781	iPhone , Lightning Charging Cable
249	176797	Google Phone , Bose SoundSport Headphones , Wi
259	176805	Google Phone , USB-C Charging Cable
263	176808	Google Phone , Wired Headphones
269	176813	Google Phone , Wired Headphones
392	176935	AAA Batteries (4-pack) , 27in FHD Monitor
433	176975	USB-C Charging Cable , AAA Batteries (4-pack)
448	176989	Google Phone , USB-C Charging Cable
453	176993	iPhone , Wired Headphones
483	177022	iPhone , Wired Headphones
564	177102	iPhone , 27in 4K Gaming Monitor

```
In [234]: # Import other necessary libraries
from itertools import combinations
from collections import Counter

count = Counter()

#Counting most pairing product in selected chosen number of pairing

for row in df['Grouped Products']:
    row_list = row.split(' , ')
    count.update(Counter(combinations(row_list, 2))) #2 way of pairing, can be 3 or 4 any chosen number

for key, value in count.most_common(10):
    print(key, value)

('iPhone' 'Lightning Changing Cable') 1805
```

```
('iPhone', 'Lightning Charging Cable') 1005
('Google Phone', 'USB-C Charging Cable') 987
('iPhone', 'Wired Headphones') 447
('Google Phone', 'Wired Headphones') 414
('Vareebadd Phone', 'USB-C Charging Cable') 361
('iPhone', 'Apple Airpods Headphones') 360
('Google Phone', 'Bose SoundSport Headphones') 220
('USB-C Charging Cable', 'Wired Headphones') 160
('Vareebadd Phone', 'Wired Headphones') 92
```

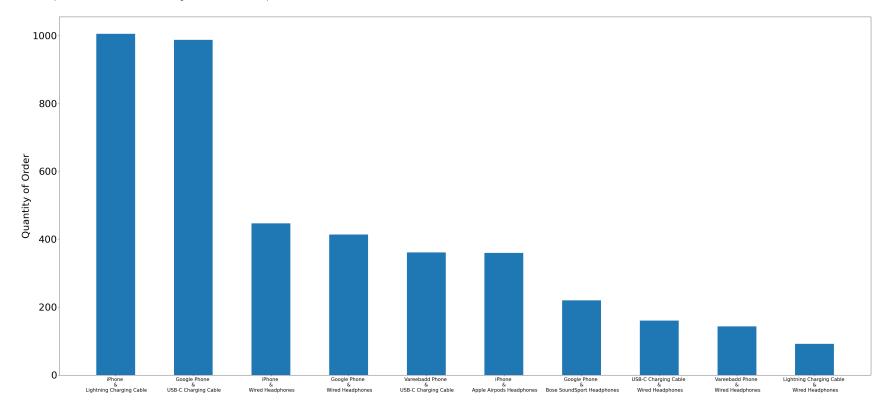
Answer 4: Iphone and Lightning Charging Cable are the most being sold together

```
In [315]: #make empty list as container for graphic data
products = list()

#insert data to list
for key, value in count.most_common(10):
    products.append(' \n&\n '.join(key)) #This data is tuple need to be convert join
    sold.append(value)

plt.figure(figsize=(45,20))
plt.bar(products,sold, width = 0.5)
plt.xticks(size=15)
plt.yticks(size=30)
plt.ylabel('Quantity of Order', size = 30)
```

Out[315]: Text(0, 0.5, 'Quantity of Order')



Question 5: What product sold the most? Why?

In [316]: all_data.head()

Out[316]:

:		Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address	Month	Sales	City	Hour	Minute
	0 1	176558	USB-C Charging Cable	2	11.95	2019-04-19 08:46:00	917 1st St, Dallas, TX 75001	4	23.90	Dallas (TX)	8	46
	1 1	176559	Bose SoundSport Headphones	1	99.99	2019-04-07 22:30:00	682 Chestnut St, Boston, MA 02215	4	99.99	Boston (MA)	22	30
	2 1	176560	Google Phone	1	600.00	2019-04-12 14:38:00	669 Spruce St, Los Angeles, CA 90001	4	600.00	Los Angeles (CA)	14	38
	3 1	176560	Wired Headphones	1	11.99	2019-04-12 14:38:00	669 Spruce St, Los Angeles, CA 90001	4	11.99	Los Angeles (CA)	14	38
	4 1	176561	Wired Headphones	1	11.99	2019-04-30 09:27:00	333 8th St, Los Angeles, CA 90001	4	11.99	Los Angeles (CA)	9	27

In [321]: all_data.groupby('Product').sum().sort_values(by='Quantity Ordered', ascending=False)

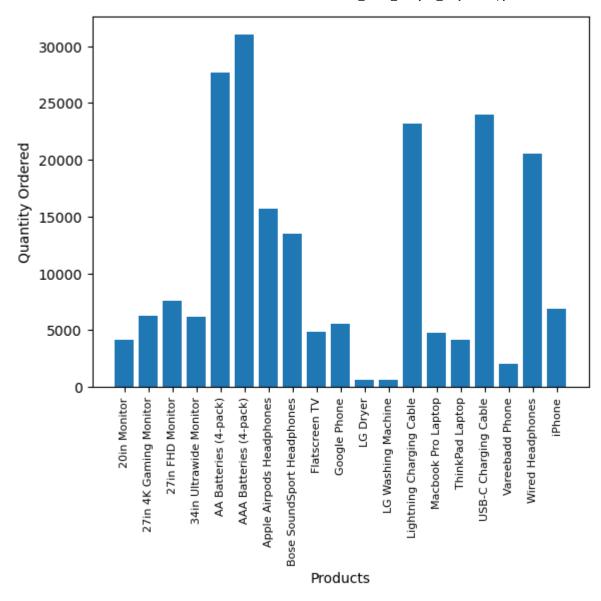
Out[321]:

	Order ID	Quantity Ordered	Price Each	Month	Sales	Hour	Minute
Product							
AAA Batteries (4-pack)	4.764959e+09	31017.0	61716.59	146370	92740.83	297332	612113
AA Batteries (4-pack)	4.744174e+09	27635.0	79015.68	145558	106118.40	298342	609039
USB-C Charging Cable	5.049538e+09	23975.0	261740.85	154819	286501.25	314645	647586
Lightning Charging Cable	4.994091e+09	23217.0	323787.10	153092	347094.15	312529	634442
Wired Headphones	4.350952e+09	20557.0	226395.18	133397	246478.43	271720	554023
Apple Airpods Headphones	3.579120e+09	15661.0	2332350.00	109477	2349150.00	223304	455570
Bose SoundSport Headphones	3.071496e+09	13457.0	1332366.75	94113	1345565.43	192445	392603
27in FHD Monitor	1.724224e+09	7550.0	1125974.93	52558	1132424.50	107540	219948
iPhone	1.571390e+09	6849.0	4789400.00	47941	4794300.00	98657	201688
27in 4K Gaming Monitor	1.442589e+09	6244.0	2429637.70	44440	2435097.56	90916	184331
34in Ultrawide Monitor	1.418986e+09	6199.0	2348718.19	43304	2355558.01	89076	183480
Google Phone	1.262237e+09	5532.0	3315000.00	38305	3319200.00	79479	162773
Flatscreen TV	1.110943e+09	4819.0	1440000.00	34224	1445700.00	68815	142789
Macbook Pro Laptop	1.091958e+09	4728.0	8030800.00	33548	8037600.00	68261	137574
ThinkPad Laptop	9.487932e+08	4130.0	4127958.72	28950	4129958.70	59746	121508
20in Monitor	9.508897e+08	4129.0	451068.99	29336	454148.71	58764	122252
Vareebadd Phone	4.725325e+08	2068.0	826000.00	14309	827200.00	29472	61835
LG Washing Machine	1.507187e+08	666.0	399600.00	4523	399600.00	9785	19462
LG Dryer	1.465563e+08	646.0	387600.00	4383	387600.00	9326	19043

```
In [327]: x = [hour for hour, x in all_data.groupby('Product')]
y = all_data.groupby('Product').sum()['Quantity Ordered']

plt.bar(x,y)
plt.xlabel('Products')
plt.ylabel('Quantity Ordered')
plt.xticks(rotation='vertical', size=8)

plt.show()
```



```
In [379]: prices= all_data.groupby('Product').mean()['Price Each']

fig, ax1 = plt.subplots()

ax2 = ax1.twinx()
    ax1.bar(x,y, color = 'blue')
    ax2.plot(x,prices, 'red')

ax1.set_xlabel('Product Name')
    ax1.set_ylabel('Quantity Ordered', color = 'blue')
    ax2.set_ylabel('Price ($)', color = 'red')
    ax1.set_xticklabels(products, rotation='vertical', size=8)

plt.show()
```

