

Import the necessary libraries

```
In [ ]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from mlxtend.frequent_patterns import apriori
from mlxtend.frequent_patterns import association_rules
```

Importing the data

```
In [ ]: df = pd.read_csv("/home/owekitiibwa/Desktop/dataAnalysis/Pandas-Data-Science-Tasks/AssociationRuleMining/Orders.csv")
df.head()
```

```
Out[ ]:
```

	Order ID	Product	Quantity Ordered	Price Each	Order Date	Purchase Address
0	194095	Wired Headphones	1	11.99	05/16/19 17:14	669 2nd St, New York City, NY 10001
1	194096	AA Batteries (4-pack)	1	3.84	05/19/19 14:43	844 Walnut St, Dallas, TX 75001
2	194097	27in FHD Monitor	1	149.99	05/24/19 11:36	164 Madison St, New York City, NY 10001
3	194098	Wired Headphones	1	11.99	05/02/19 20:40	622 Meadow St, Dallas, TX 75001
4	194099	AAA Batteries (4-pack)	2	2.99	05/11/19 22:55	17 Church St, Seattle, WA 98101

Data Cleaning

Count missing values for each column and sort in descending order

```
In [ ]: missing_values = df.isnull().sum().sort_values(ascending=False)
missing_values
```

```
Out[ ]: Order ID      1090
Product      1090
Quantity Ordered  1090
Price Each    1090
Order Date    1090
Purchase Address  1090
dtype: int64
```

Deleting missing values from the dataset

```
In [ ]: df = df.dropna()
missing_values = df.isnull().sum().sort_values(ascending=False)
missing_values
```

```
Out[ ]: Order ID      0
        Product      0
        Quantity Ordered  0
        Price Each    0
        Order Date    0
        Purchase Address  0
        dtype: int64
```

Checking for duplicates and delete them

```
In [ ]: # Use the `duplicated` method to check for duplicate rows
        duplicates = df[df.duplicated()]
        df = df.drop_duplicates()
        #Checking whether duplicates have been removed
        duplicates = df[df.duplicated()]
        duplicates
```

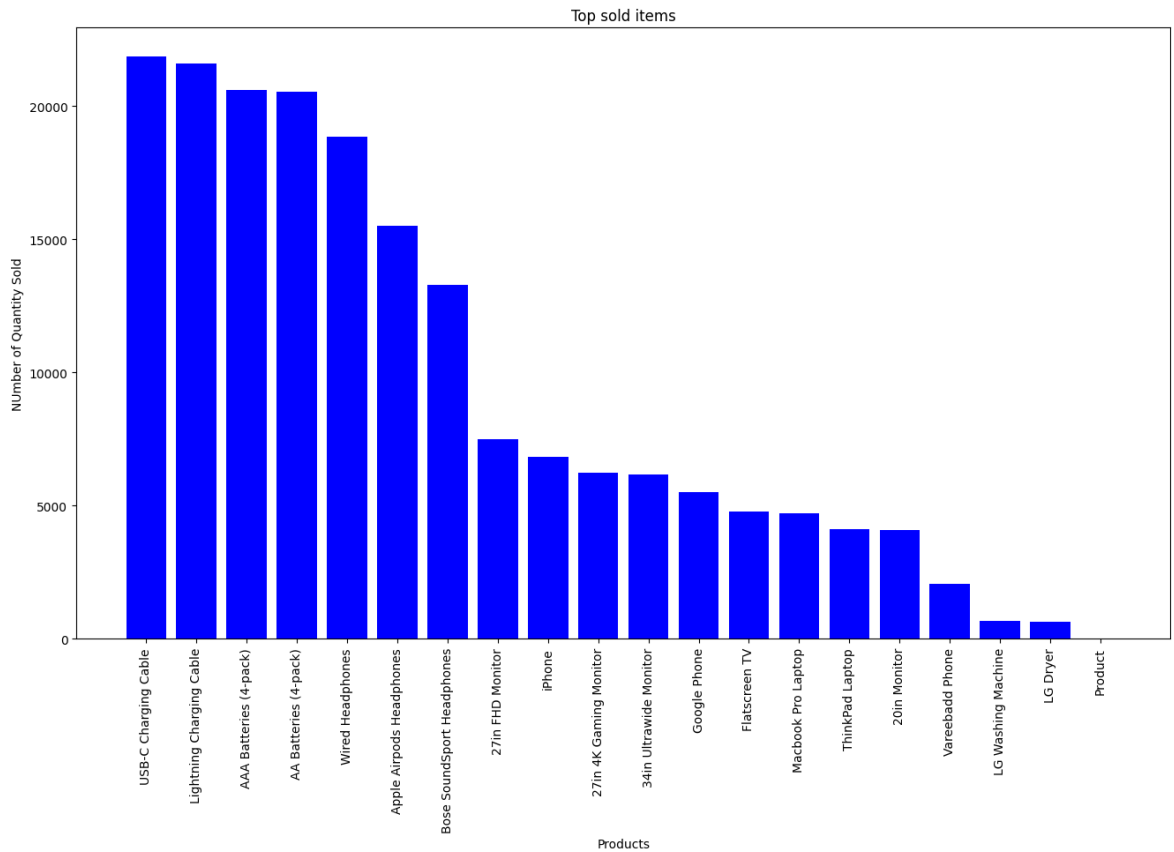
```
Out[ ]: Order ID  Product  Quantity Ordered  Price Each  Order Date  Purchase Address
```

Plotting the item distribution

```
In [ ]: Itemdistro = df.groupby(by= "Product").size().reset_index(name= 'Frequency')
        bars = Itemdistro['Product']
        height = Itemdistro['Frequency']
        x_pos = np.arange(len(bars))

        plt.figure(figsize=(16,9))
        plt.bar(x_pos,height,color="blue")
        plt.title("Top sold items")
        plt.xlabel("Products")
        plt.ylabel("NUmber of Quantity Sold")
        plt.xticks(x_pos, bars, rotation=90)

        plt.show()
```



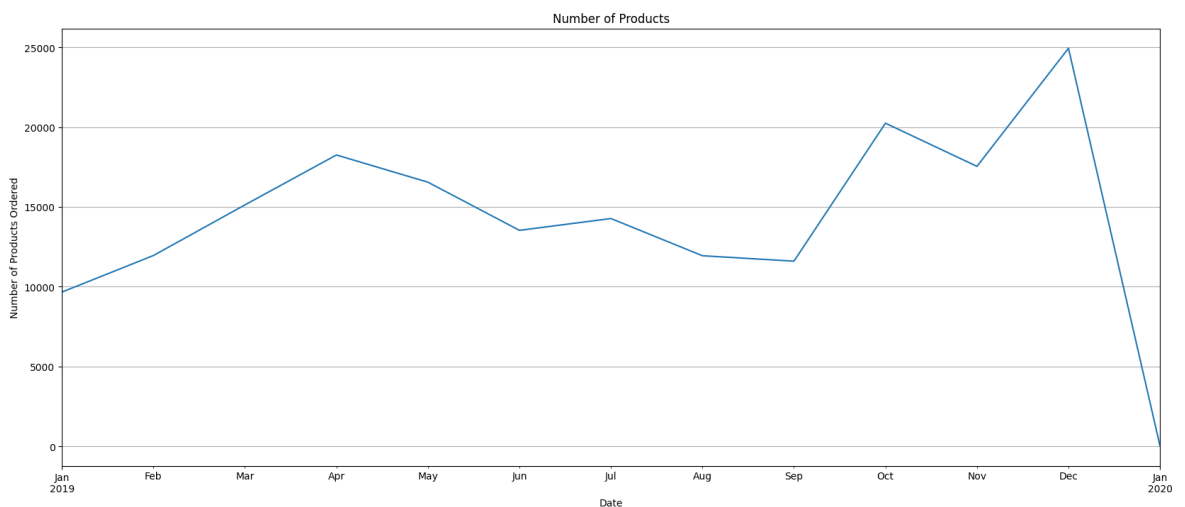
Lets us create a chart showing trends the products ordered

```
In [ ]: df['Order Date'] = pd.to_datetime(df['Order Date'], errors='coerce')
df_date = df.set_index(df["Order Date"])
df_date.resample('M')['Product'].count().plot(figsize = (20,8),grid= True)
```

/tmp/ipykernel_258289/1884926306.py:1: UserWarning: Could not infer format, so each element will be parsed individually, falling back to `dateutil`. To ensure parsing is consistent and as-expected, please specify a format.

```
df['Order Date'] = pd.to_datetime(df['Order Date'], errors='coerce')
```

```
Out[ ]: [Text(0.5, 0, 'Date'), Text(0, 0.5, 'Number of Products Ordered')]
```



Now, let's pivot this table to convert the items into columns and the transaction into rows:

```
In [ ]: df2 = pd.crosstab(df['Order ID'], df['Product'])
df2.head()
```

```
Out[ ]:
```

Product	20in Monitor	27in 4K Gaming Monitor	27in FHD Monitor	34in Ultrawide Monitor	AA Batteries (4-pack)	AAA Batteries (4-pack)	Apple AirPods Headphones	S
Order ID								
141275	0	0	0	0	0	0	0	
141290	0	0	0	0	1	0	1	
141365	0	0	0	0	0	0	0	
141384	0	0	0	0	0	0	0	
141450	0	0	0	0	0	0	0	

Now lets encode the data for the Apriori Algorithm

```
In [ ]: def encode(item_freq):
    res = 0
    if item_freq > 0:
        res = 1
    return res

basket_input = df2.applymap(encode)
```

/tmp/ipykernel_258289/230850535.py:7: FutureWarning: DataFrame.applymap has been deprecated. Use DataFrame.map instead.
 basket_input = df2.applymap(encode)

Build the Apriori Algorithm for Market Basket Analysis

```
In [ ]: from mlxtend.frequent_patterns import apriori
from mlxtend.frequent_patterns import association_rules
frequent_itemsets = apriori(basket_input, min_support=0.001, use_colnames=True)

rules = association_rules(frequent_itemsets, metric="lift")

rules.sort_values(["support", "confidence", "lift"], axis = 0, ascending = False)
```

/home/owekitiibwa/.local/lib/python3.10/site-packages/mlxtend/frequent_patterns/fpcommon.py:110: DeprecationWarning: DataFrames with non-bool types result in worse computational performance and their support might be discontinued in the future. Please use a DataFrame with bool type
 warnings.warn(

Out[]:

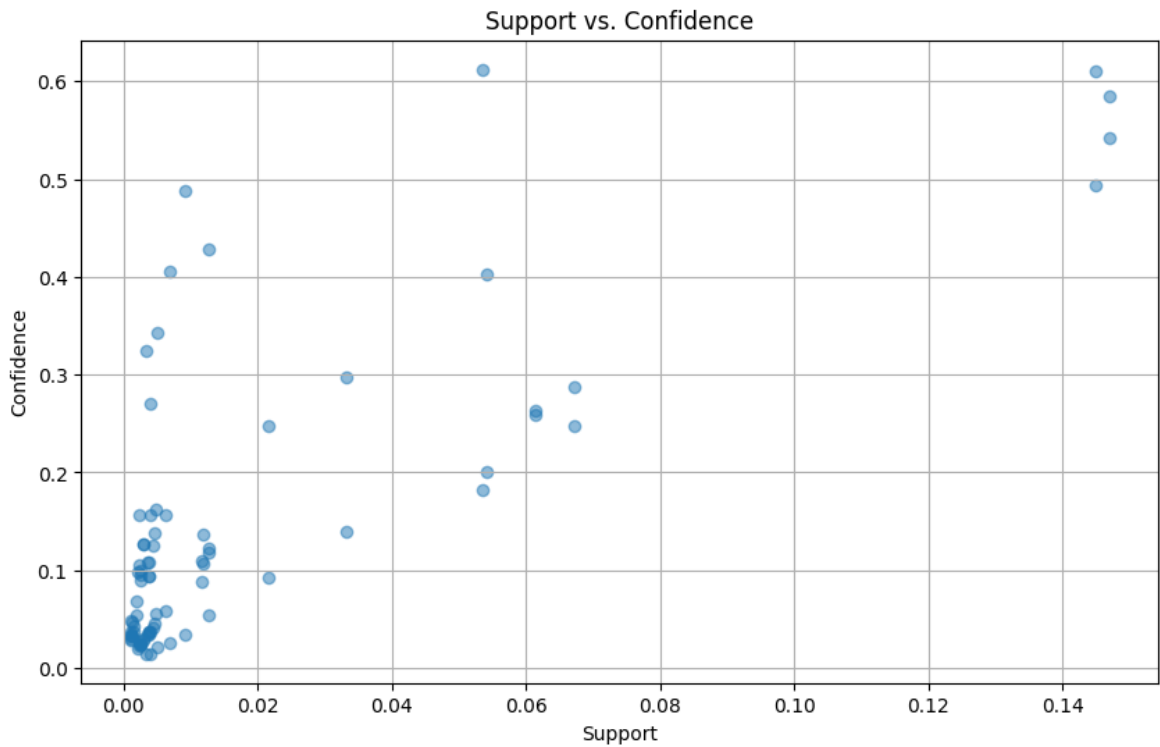
	antecedents	consequents	antecedent support	consequent support	support	confidence	li
57	(Lightning Charging Cable)	(iPhone)	0.251199	0.270970	0.146969	0.585069	2.1591
56	(iPhone)	(Lightning Charging Cable)	0.270970	0.251199	0.146969	0.542382	2.1591
53	(Google Phone)	(USB-C Charging Cable)	0.237389	0.293793	0.144934	0.610533	2.0781
52	(USB-C Charging Cable)	(Google Phone)	0.293793	0.237389	0.144934	0.493320	2.0781
62	(Wired Headphones)	(iPhone)	0.233610	0.270970	0.067161	0.287492	1.0609
63	(iPhone)	(Wired Headphones)	0.270970	0.233610	0.067161	0.247854	1.0609
54	(Wired Headphones)	(Google Phone)	0.233610	0.237389	0.061346	0.262601	1.1062
55	(Google Phone)	(Wired Headphones)	0.237389	0.233610	0.061346	0.258420	1.1062
47	(Apple AirPods Headphones)	(iPhone)	0.134613	0.270970	0.054223	0.402808	1.4865
46	(iPhone)	(Apple AirPods Headphones)	0.270970	0.134613	0.054223	0.200107	1.4865

Support vs. Confidence Scatter Plot

```

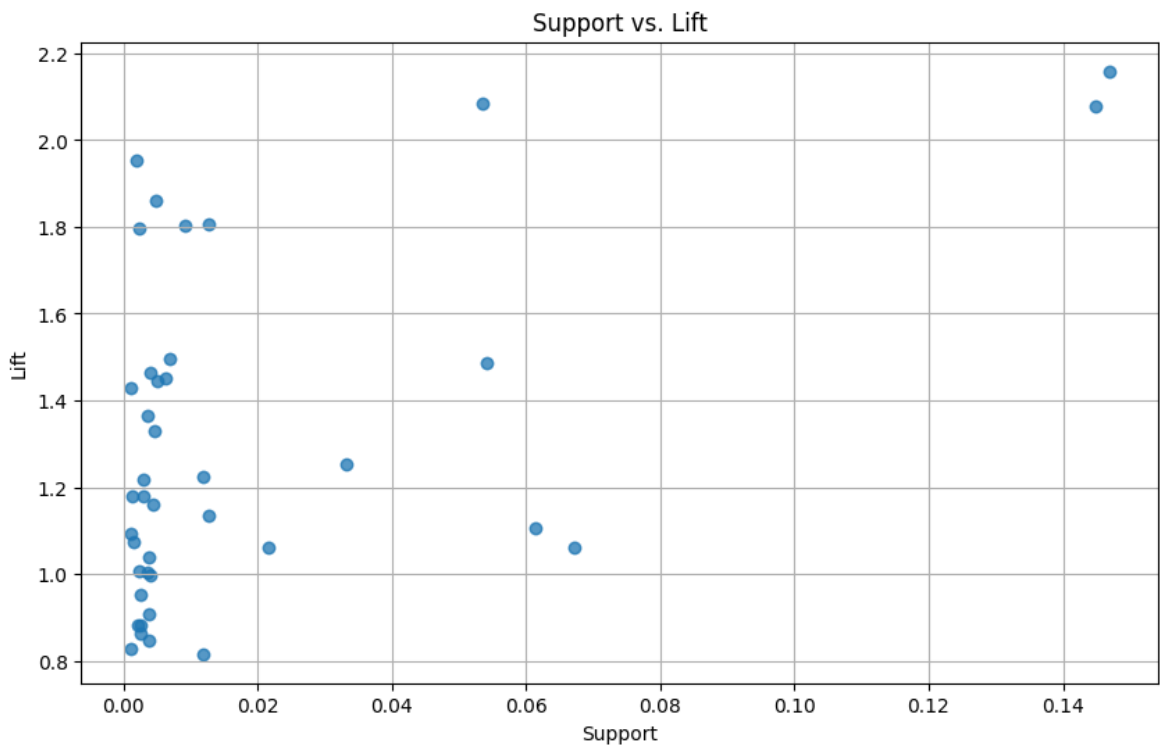
In [ ]: plt.figure(figsize=(10, 6))
plt.scatter(rules['support'], rules['confidence'], alpha=0.5)
plt.xlabel('Support')
plt.ylabel('Confidence')
plt.title('Support vs. Confidence')
plt.grid()
plt.show()

```



Support vs Lift Scatter Plot

```
In [ ]: plt.figure(figsize=(10, 6))
plt.scatter(rules['support'], rules['lift'], alpha=0.5)
plt.xlabel('Support')
plt.ylabel('Lift')
plt.title('Support vs. Lift')
plt.grid()
plt.show()
```



Confidence vs. Lift Scatter Plot

```
In [ ]: plt.figure(figsize=(10, 6))
plt.scatter(rules['confidence'], rules['lift'], alpha=0.5)
plt.xlabel('Confidence')
plt.ylabel('Lift')
plt.title('Confidence vs. Lift')
plt.grid()
plt.show()
```

