

## ✓ 1\_TimeSeriesData

A time series is the series of data points listed in time order.

A time series is a sequence of successive equal interval points in time.


A time-series analysis consists of methods for analyzing time series data in order to extract meaningful insights and other useful characteristics of data.

For performing time series analysis download stock\_data.csv


```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

```
# reading the dataset using read_csv
df = pd.read_csv("/content/stock_data.csv",
                 parse_dates=True,
                 index_col="Date")
```

```
# displaying the first five rows of dataset
df.head()
```



	Open	High	Low	Close	Volume	Name
Date						
2006-01-03	39.69	41.22	38.79	40.91	24232729	AABA
2006-01-04	41.22	41.90	40.77	40.97	20553479	AABA
2006-01-05	40.93	41.73	40.85	41.53	12829610	AABA
2006-01-06	42.88	43.57	42.80	43.21	29422828	AABA
2006-01-09	43.10	43.66	42.82	43.42	16268338	AABA



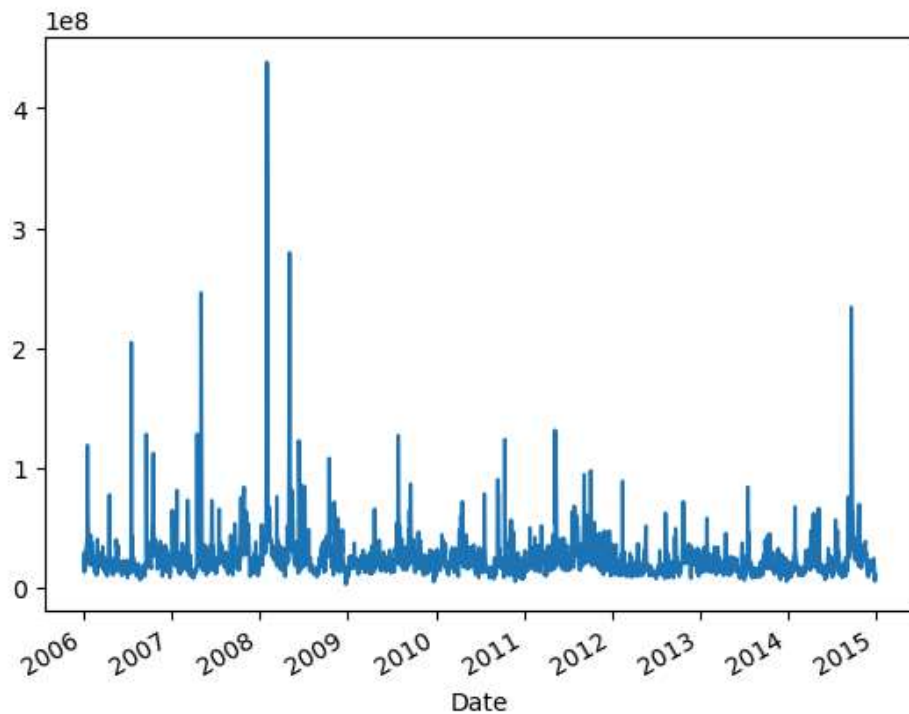
Next steps:

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```
# deleting column  
df=df.drop(columns='Name',axis=1)
```

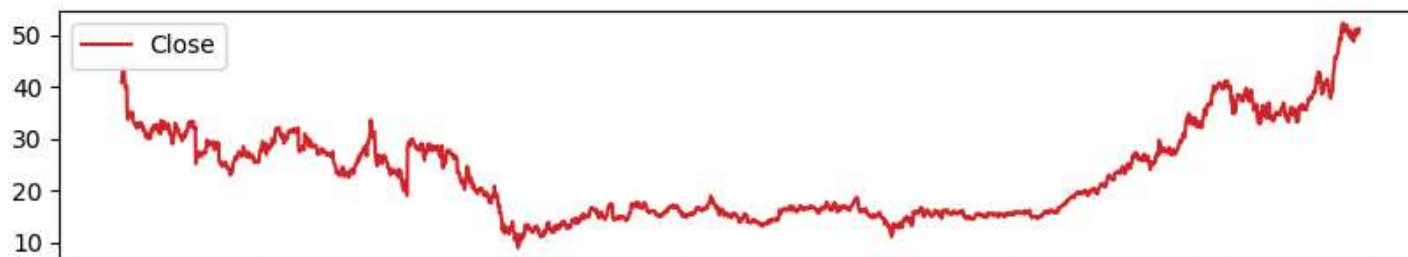
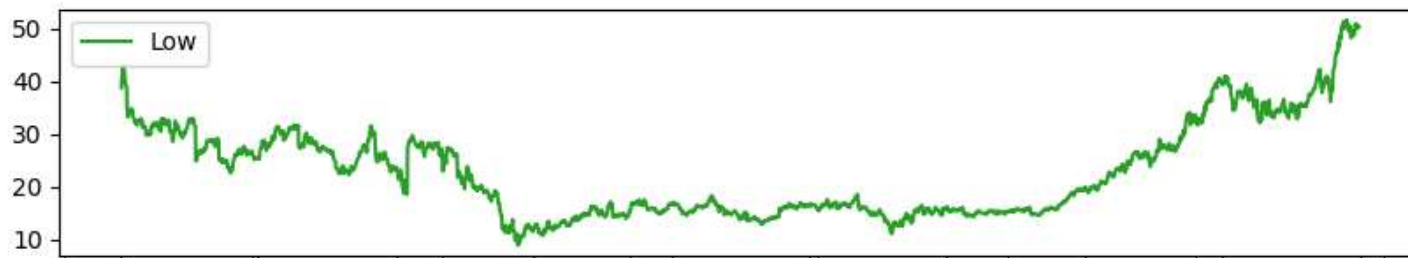
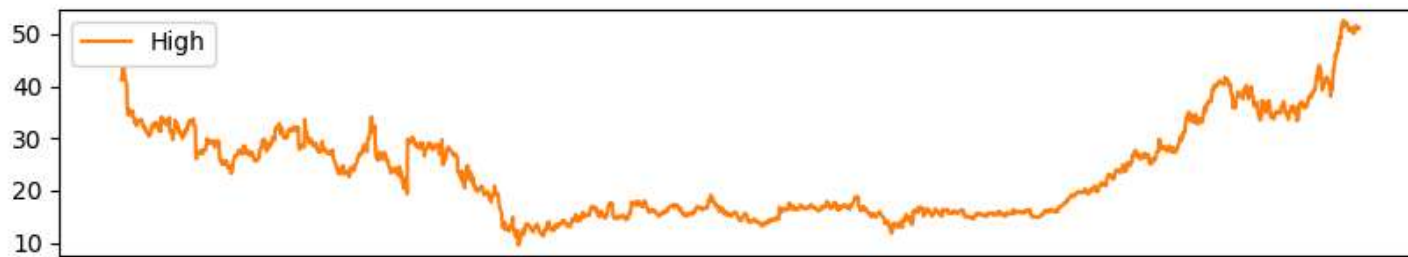
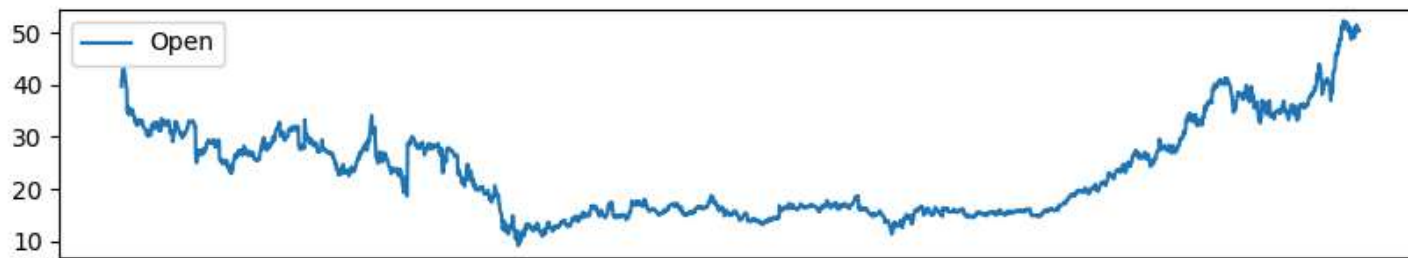
```
df['Volume'].plot()
```

↔ <Axes: xlabel='Date'>



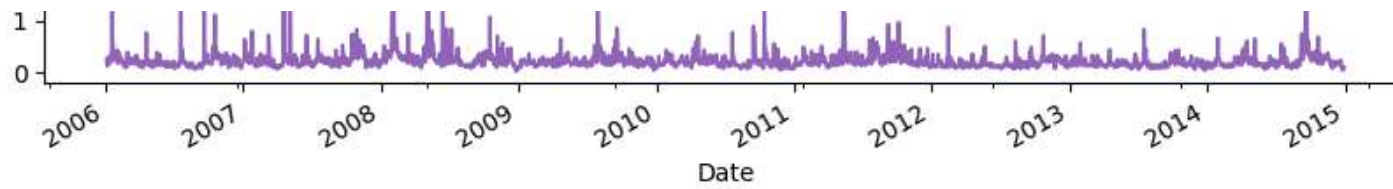
```
df.plot(subplots=True, figsize=(10, 12))
```

```
array([<Axes: xlabel='Date'>, <Axes: xlabel='Date'>,
      <Axes: xlabel='Date'>, <Axes: xlabel='Date'>,
      <Axes: xlabel='Date'>], dtype=object)
```



1e8



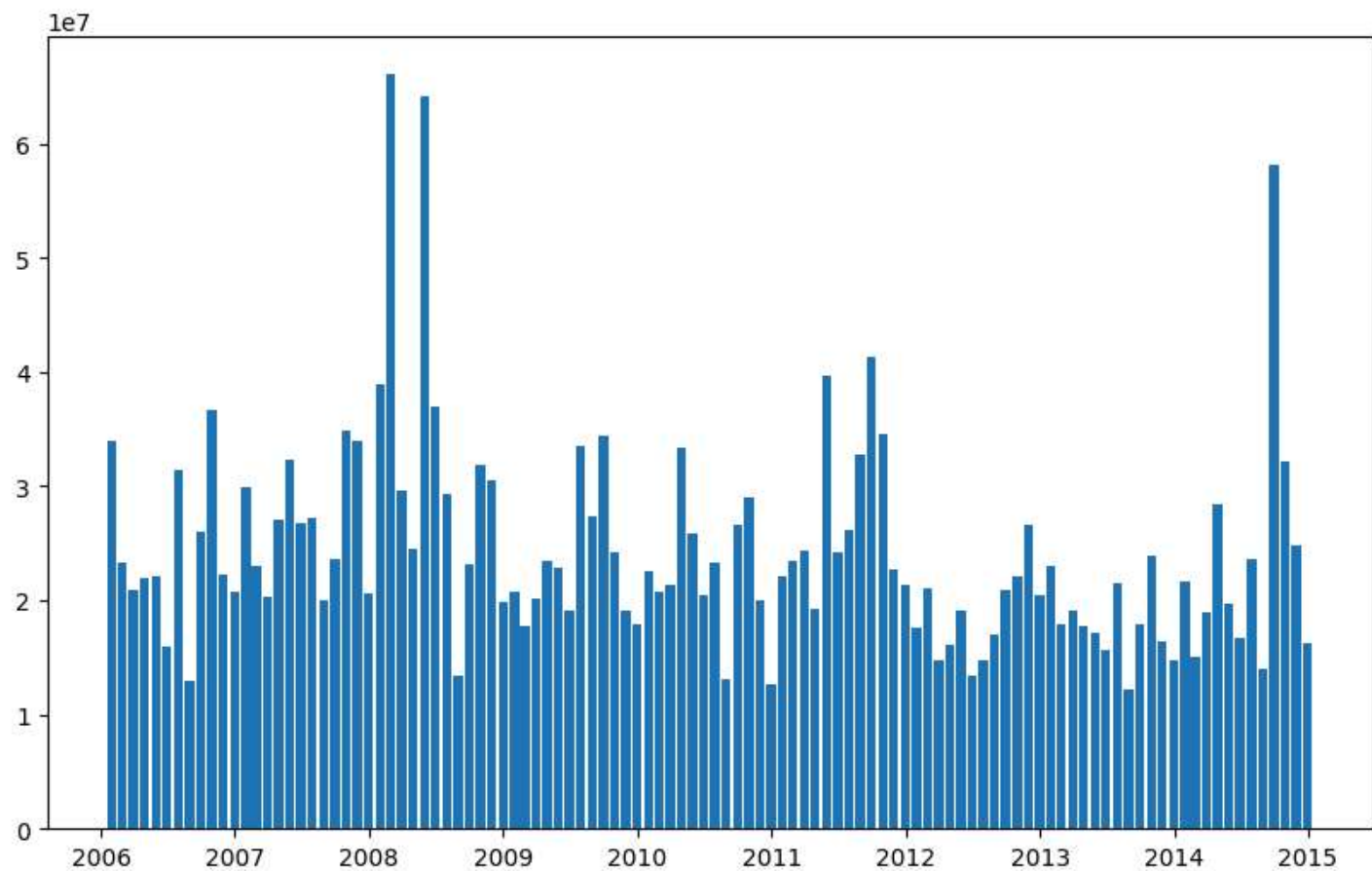


```
# Resampling the time series data based on monthly 'M' frequency
df_month = df.resample("M").mean()

# using subplot
fig, ax = plt.subplots(figsize=(10, 6))

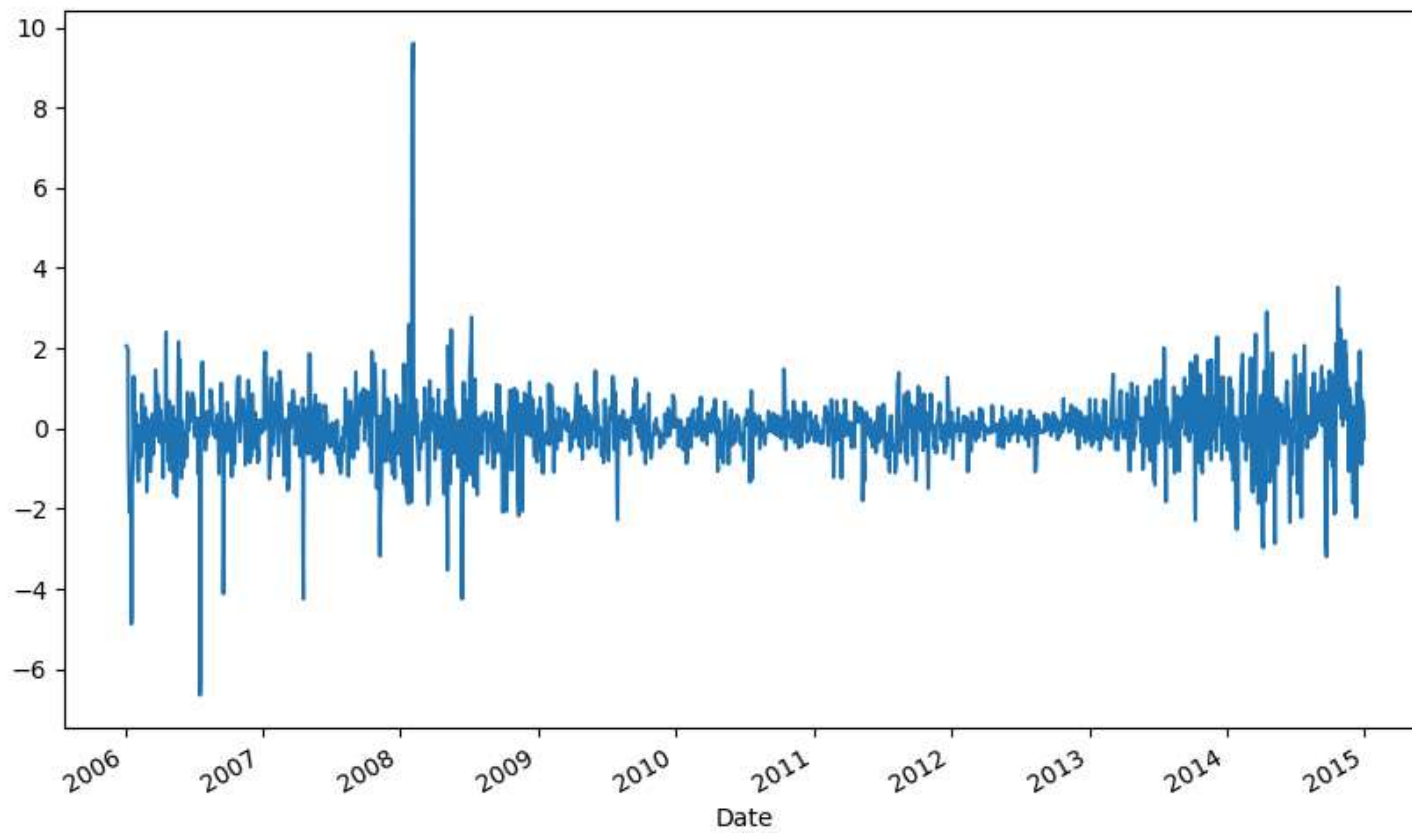
# plotting bar graph
ax.bar(df_month['2006:'].index,
      df_month.loc['2006:', "Volume"],
      width=25, align='center')
```

↗ <BarContainer object of 108 artists>



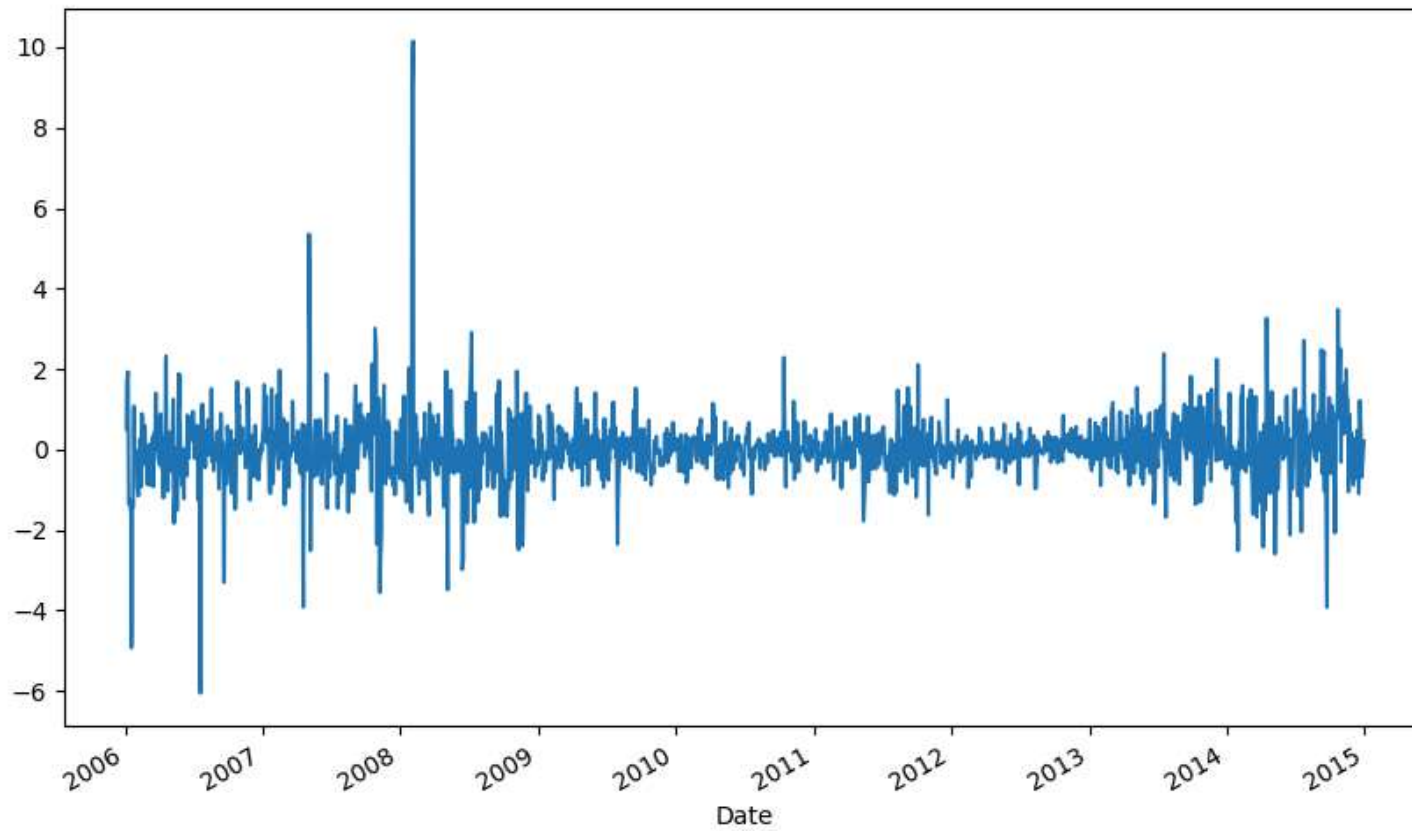
```
df.Low.diff(2).plot(figsize=(10, 6))
```

↔ <Axes: xlabel='Date'>




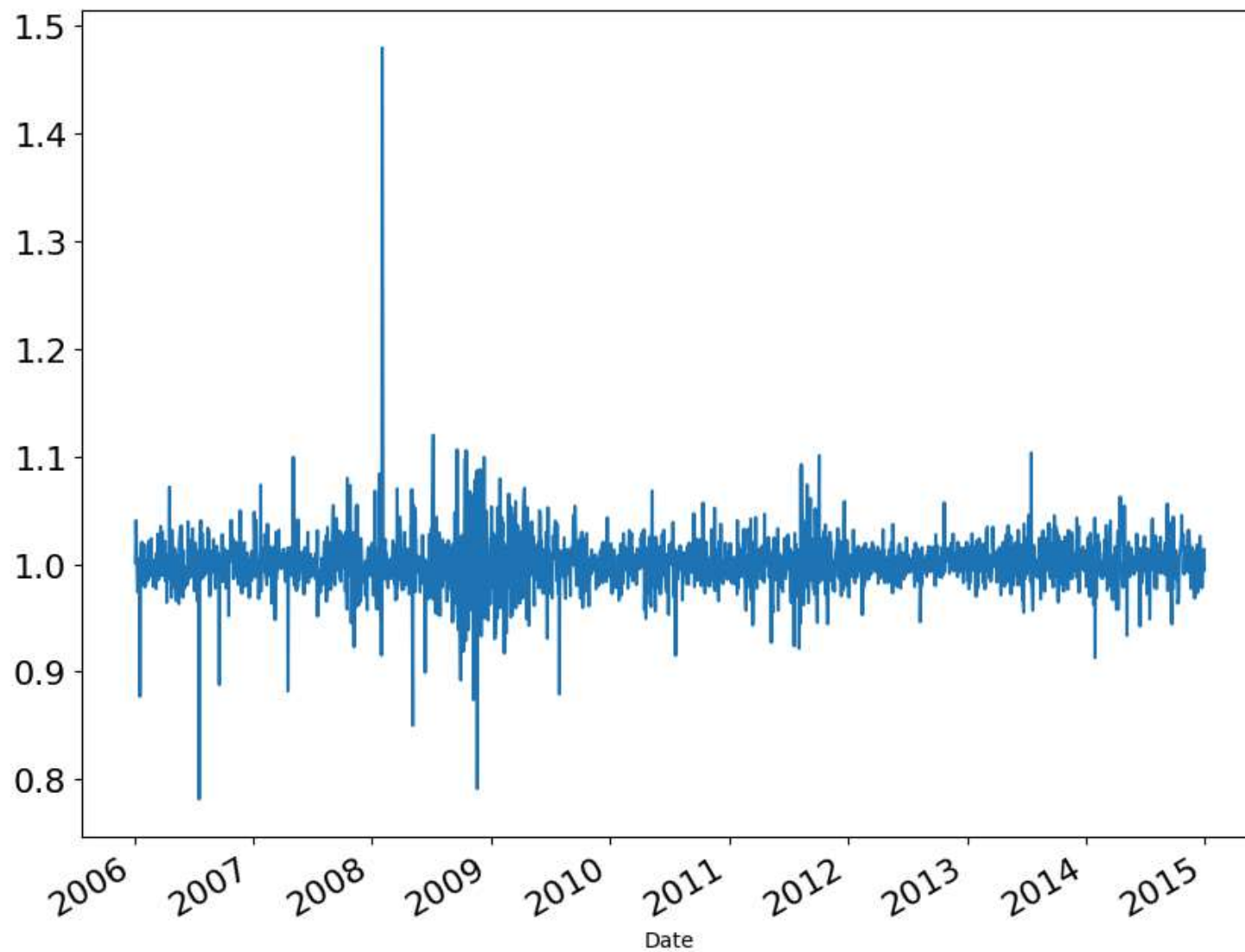
```
df.High.diff(2).plot(figsize=(10, 6))
```

↔ <Axes: xlabel='Date'>



```
df['Change'] = df.Close.div(df.Close.shift())  
df['Change'].plot(figsize=(10, 8), fontsize=16)
```

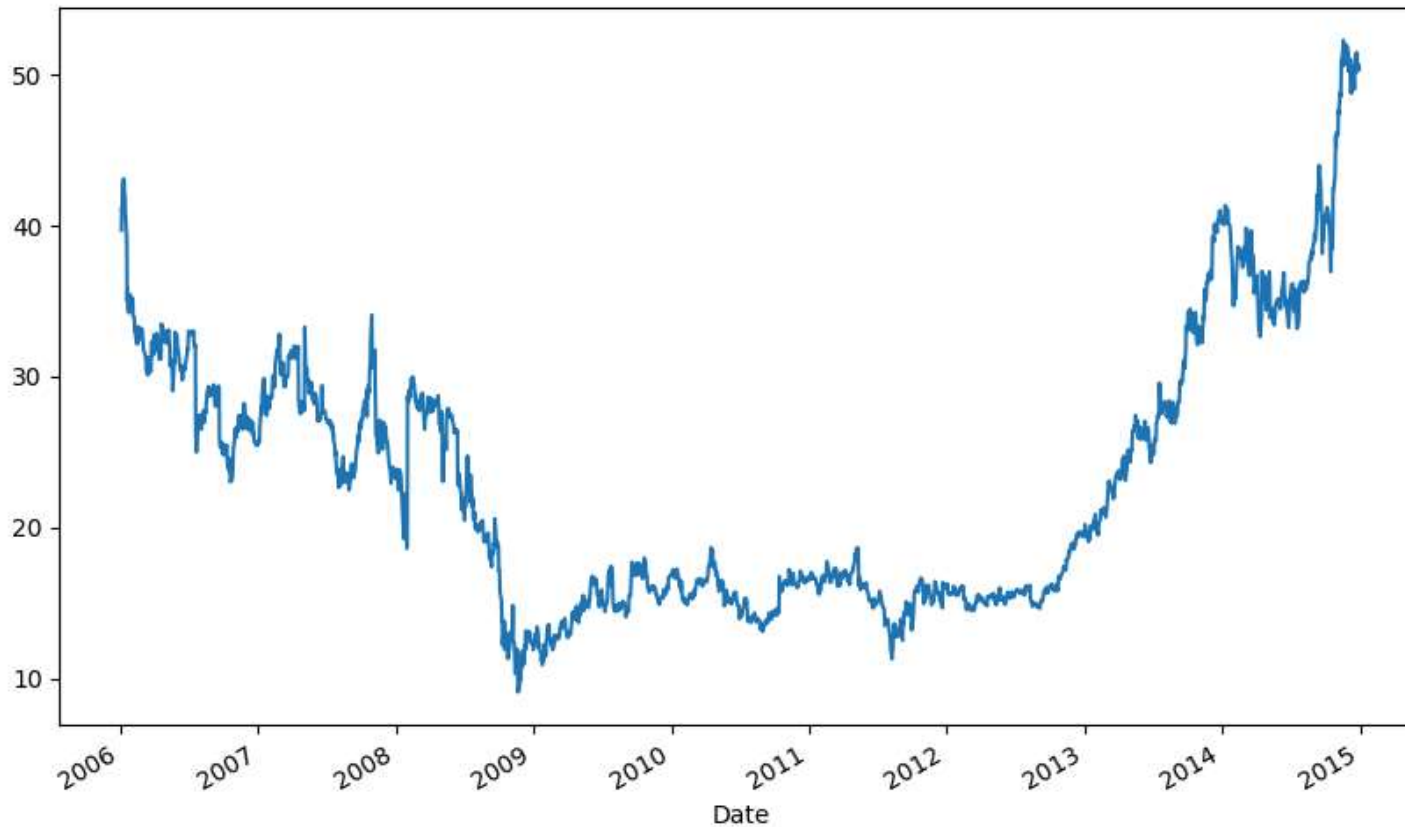
 <Axes: xlabel='Date'>



```
df['Open'].plot(figsize=(10, 6))
```



↔ <Axes: xlabel='Date'>



## ✓ Market\_Basket

Market basket analysis is used by companies to identify items that are frequently purchased together.

### **How Does Market Basket Analysis Work?**

Market basket analysis is frequently used by restaurants, retail stores, and online shopping platforms to encourage customers to make more purchases in a single visit. This is a use-case of data science in marketing that increases company sales and drives business growth and commonly utilizes the Apriori algorithm.

### **What is the Apriori Algorithm?**

The Apriori algorithm is the most common technique for performing market basket analysis.

It is used for association rule mining, which is a rule-based process used to identify correlations between items purchased by users.

### What Are the Components of the Apriori Algorithm?

The Apriori algorithm has three main components:

- Support
- Lift
- Confidence

Here is a tabular representation of this purchase data:

	Milk	Beer	Eggs	Bread	Bananas	Apples
--	------	------	------	-------	---------	--------

Basket1 1 1 1 1 0 0

Basket2 1 0 0 1 0 0

Basket3 1 0 0 1 0 1

Basket4 0 0 0 1 1 1

Let's calculate the support, confidence, and lift.

#### Support

The first component of the Apriori algorithm is support – we use it to assess the overall popularity of a given product with the following formula:

$\text{Support}(\text{item}) = \text{Transactions comprising the item} / \text{Total transactions}$

A high support value indicates that the item is present in most purchases, therefore marketers should focus on it more.

#### Confidence

Confidence tells us the likelihood of different purchase combinations. We calculate that using the following formula:

$\text{Confidence}(\text{Bread} \rightarrow \text{Milk}) = \text{Transactions comprising bread and milk} / \text{Transactions comprising bread}$

#### Lift

Finally, lift refers to the increase in the ratio of the sale of milk when you sell bread:

$\text{Lift} = \text{Confidence}(\text{Bread} \rightarrow \text{Milk}) / \text{Support}(\text{Bread}) = 0.75/1 = 1.3.$

This means that customers are 1.3 times more likely to buy milk if you also sell bread.

### Step 1: Pre-Requisites for Performing Market Basket Analysis

Download the dataset "groceries\_dataset.csv"

### Step 2: Reading the Dataset

```
import pandas as pd
from google.colab import drive
drive.mount('/content/drive')
df = pd.read_csv('content/drive/My Drive/Data/Groceries_dataset.csv')
df.head()
```

### Step 3: Data Preparation for Market Basket Analysis

Before we perform market basket analysis, we need to convert this data into a format that can easily be ingested into the Apriori algorithm. In other words, we need to turn it into a tabular structure comprising ones and zeros, as displayed in the bread and milk example above.

To achieve this, the first group items that have the same member number and date:

```
df['single_transaction'] = df['Member_number'].astype(str)+'_'+df['Date'].astype(str)
df.head()
```

```
df2 = pd.crosstab(df['single_transaction'], df['itemDescription'])
df2.head()
```

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

basket_input = df2.applymap(encode)
```

### Step 4: Build the Apriori Algorithm for Market Basket Analysis

Now, let's import the Apriori algorithm from the MLXtend Python package and use it to discover frequently-bought-together item combinations:

```
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.head()

rules.sort_values(["support", "confidence","lift"],axis = 0, ascending = False).head(8)
```

## ✓ 3\_TextVisualization


### Load the Pacakges

To get started, open a Colab notebook and load the Pandas, Matplotlib, and Wordcloud packages.

```
import pandas as pd
import matplotlib.pyplot as plt
from wordcloud import WordCloud
from wordcloud import STOPWORDS
```

```
from google.colab import drive
```

```
drive.mount('/content/drive/')
```

 Mounted at /content/drive/

```
df=pd.read_csv('/content/netflix_titles.csv', usecols=['cast'])
df.head()
```



cast



0

NaN



1 Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban...

2 Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi...

3 NaN

4 Mayur More, Jitendra Kumar, Ranjan Raj, Alam K...

Next steps:

[Generate code with df](#)

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```
ndf=df.dropna()  
ndf.head()
```



cast



1 Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban...



2 Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi...

4 Mayur More, Jitendra Kumar, Ranjan Raj, Alam K...

5 Kate Siegel, Zach Gilford, Hamish Linklater, H...

6 Vanessa Hudgens, Kimiko Glenn, James Marsden, ...

Next steps:

[Generate code with ndf](#)

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```
text = " ".join(item for item in ndf['cast'])  
print(text)
```



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```
stopwords = set(STOPWORDS)
```





## What is Clustering?

Clustering is the process of separating different parts of data based on common characteristics. Disparate industries including retail, finance and healthcare use clustering techniques for various analytical tasks

```
from google.colab import drive
drive.mount('/content/drive')
```

➦ Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force\_remount=True).

```
import matplotlib.pyplot as plt
import seaborn as sns
import pandas as pd
df = pd.read_csv('/content/Mall_Customers.csv')
print(df.head(15))
```

➦

	CustomerID	Gender	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	Male	19	15	39
1	2	Male	21	15	81
2	3	Female	20	16	6
3	4	Female	23	16	77
4	5	Female	31	17	40
5	6	Female	22	17	76
6	7	Female	35	18	6
7	8	Female	23	18	94
8	9	Male	64	19	3
9	10	Female	30	19	72
10	11	Male	67	19	14
11	12	Female	35	19	99
12	13	Female	58	20	15
13	14	Female	24	20	77
14	15	Male	37	20	13

```
from sklearn.cluster import KMeans
```