

```
In [38]: import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)

import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))
```

```
In [2]: import numpy as np
import pandas as pd
import seaborn as sns
```

```
In [3]: import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings('ignore')
sns.set_style('darkgrid')
```

```
In [10]: #Creating Data base and storing it

data = pd.read_csv("C:\\Users\\user\\Downloads\\nifty_500 (1).csv")
data.head()
```

Out[10]:

	Company Name	Symbol	Industry	Series	Open	High	Low	Previous Close	Last Traded Price	Change	Percentage Change	Share Volume	Va
0	3M India Ltd.	3MINDIA	Diversified	EQ	21950.00	21999.00	21126.05	21854.05	21575.00	-279.05	-1.28	4159	8.9
1	Aarti Drugs Ltd.	AARTIDRUGS	Healthcare	EQ	400.50	401.80	394.10	403.85	400.00	-3.85	-0.95	31782	1.2
2	Aavas Financiers Ltd.	AAVAS	Financial Services	EQ	1997.10	2004.05	1894.50	2015.45	1943.15	-72.30	-3.59	150704	2.9
3	ABB India Ltd.	ABB	Capital Goods	EQ	2260.35	2311.50	2260.35	2300.90	2280.00	-20.90	-0.91	97053	2.2
4	Abbott India Ltd.	ABBOTINDIA	Healthcare	EQ	18700.40	19200.00	18605.00	18760.40	19199.80	439.40	2.34	12396	2.3

```
In [11]: #Let's take a look at its various columns for further analysis.

#The Open and Previous Close columns indicate the opening and closing price of the stocks on a particular day
#Industry Sectors to which the stocks belongs
#The High and Low columns provide the highest and the lowest price for the stock on a particular day, respect
#The Share Volume column tells us the total volume of stocks traded on a particular day.
#52 Week High and Low columns Provides the yearly high and low face value of the stock
```

```
In [12]: data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 501 entries, 0 to 500
Data columns (total 17 columns):
#   Column              Non-Null Count  Dtype
---  -
0   Company Name        501 non-null    object
1   Symbol              501 non-null    object
2   Industry            501 non-null    object
3   Series              501 non-null    object
4   Open                501 non-null    float64
5   High                501 non-null    float64
6   Low                 501 non-null    float64
7   Previous Close      501 non-null    float64
8   Last Traded Price   501 non-null    float64
9   Change              501 non-null    object
10  Percentage Change    501 non-null    object
11  Share Volume        501 non-null    int64
12  Value (Indian Rupee) 501 non-null    float64
```

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13 52 Week High          501 non-null    float64
14 52 Week Low           501 non-null    float64
15 365 Day Percentage Change 501 non-null    object
16 30 Day Percentage Change 501 non-null    object
dtypes: float64(8), int64(1), object(8)
memory usage: 66.7+ KB

```

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In [13]: object_dtype = data.select_dtypes(include=object)
         object_dtype.columns

```

```

Out[13]: Index(['Company Name', 'Symbol', 'Industry', 'Series', 'Change',
               'Percentage Change', '365 Day Percentage Change',
               '30 Day Percentage Change'],
              dtype='object')

```

```

In [14]: num_dtype = data.select_dtypes(exclude=object)
         num_dtype.columns

```

```

Out[14]: Index(['Open', 'High', 'Low', 'Previous Close', 'Last Traded Price',
               'Share Volume', 'Value (Indian Rupee)', '52 Week High', '52 Week Low'],
              dtype='object')

```

```

In [15]: #Getting shapes (rows and column)
         data_shape = data.shape
         print(f"The dataframe has {data_shape[0]} records and {data_shape[1]} features")

```

The dataframe has 501 records and 17 features

```

In [16]: data.describe().T

```

```

Out[16]:
```

		count	mean	std	min	25%	50%	75%	max
	Open	501.0	1.525904e+03	4.466627e+03	6.75	215.30	5.511000e+02	1.404500e+03	7.030000e+04
	High	501.0	1.553805e+03	4.576378e+03	6.95	221.55	5.691000e+02	1.421250e+03	7.250000e+04
	Low	501.0	1.504042e+03	4.435492e+03	6.70	210.60	5.470000e+02	1.396850e+03	7.030000e+04
	Previous Close	501.0	1.528061e+03	4.477209e+03	6.85	217.20	5.547500e+02	1.411700e+03	7.080090e+04
	Last Traded Price	501.0	1.536925e+03	4.532005e+03	6.80	214.65	5.630000e+02	1.410000e+03	7.190000e+04
	Share Volume	501.0	2.580350e+06	9.407021e+06	1507.00	77405.00	3.296100e+05	1.235612e+06	1.257883e+08
	Value (Indian Rupee)	501.0	8.635146e+08	4.335973e+09	2587222.80	45022649.25	1.533133e+08	6.644570e+08	9.211987e+10
	52 Week High	501.0	2.182632e+03	5.728930e+03	13.10	328.70	8.180000e+02	2.096750e+03	8.755000e+04
	52 Week Low	501.0	1.281629e+03	3.850530e+03	4.55	166.80	4.380500e+02	1.128800e+03	6.300000e+04

```

In [17]: data.describe(include=object).T

```

```

Out[17]:
```

		count	unique		top	freq
	Company Name	501	501	Godfrey Phillips India Ltd.		1
	Symbol	501	501	BAJAJHLDNG		1
	Industry	501	21	Financial Services		88
	Series	501	2	EQ		500
	Change	501	372		0.10	7
	Percentage Change	501	354		-0.33	5
	365 Day Percentage Change	501	441		-	44
	30 Day Percentage Change	501	458		0.98	3

```

In [18]: #Coorealtion

```

```

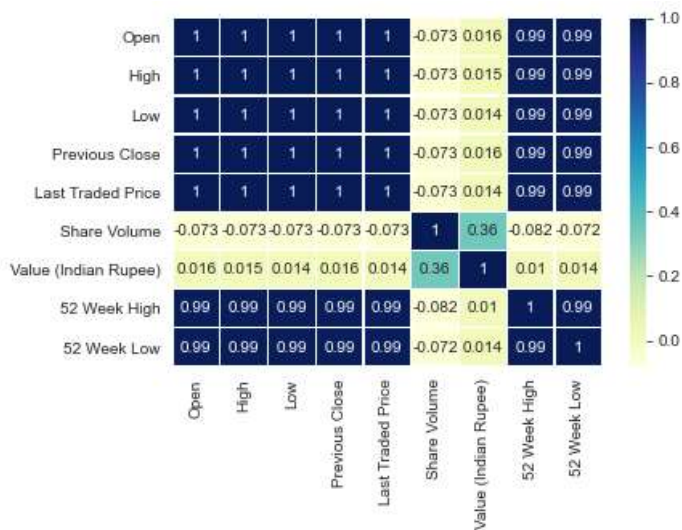
In [19]: sns.heatmap(data.corr(),annot=True,cmap='YlGnBu',linewidths=0.2)

```

```

Out[19]: <AxesSubplot:>

```



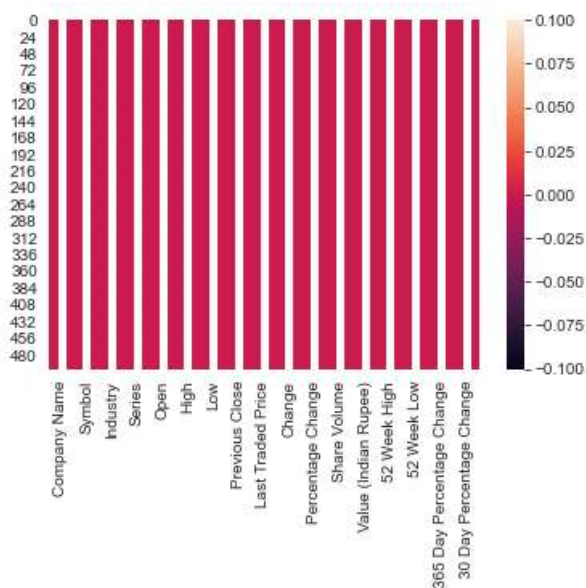
```
In [20]: #Data Cleaning
#Checking for Not available value(NULL/NA)
```

```
In [21]: data.isna().any()
```

```
Out[21]: Company Name      False
Symbol      False
Industry     False
Series       False
Open         False
High         False
Low          False
Previous Close False
Last Traded Price False
Change       False
Percentage Change False
Share Volume False
Value (Indian Rupee) False
52 Week High False
52 Week Low  False
365 Day Percentage Change False
30 Day Percentage Change False
dtype: bool
```

```
In [22]: sns.heatmap(data.isna(),annot=True)
```

```
Out[22]: <AxesSubplot:>
```



```
In [23]: #There is no null values in the dataset

#Getting unique values
```

```
In [24]: data.nunique()
```

```
Out[24]: Company Name      501
Symbol      501
Industry     21
Series       2
Open        492
High        495
Low         493
Previous Close 495
Last Traded Price 493
Change       372
Percentage Change 354
Share Volume  501
Value (Indian Rupee) 501
52 Week High  497
52 Week Low   494
365 Day Percentage Change 441
30 Day Percentage Change 458
dtype: int64
```

```
In [25]: #Dropping Symbol and Series Columns

data.drop(columns=['Symbol','Series'],inplace=True)
data.columns
```

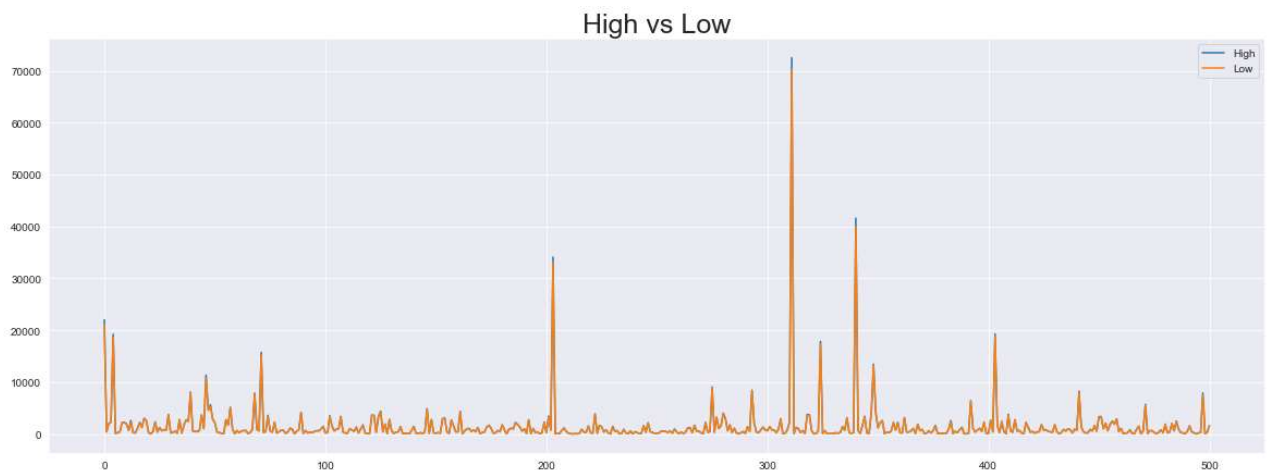
```
Out[25]: Index(['Company Name', 'Industry', 'Open', 'High', 'Low', 'Previous Close',
               'Last Traded Price', 'Change', 'Percentage Change', 'Share Volume',
               'Value (Indian Rupee)', '52 Week High', '52 Week Low',
               '365 Day Percentage Change', '30 Day Percentage Change'],
              dtype='object')
```

```
In [26]: #Data Visualization
#In this phase,we will witness a lot of line graphs which can help us to understand the trend

#Analysing High vs Low
```

```
In [27]: fx = data[['High','Low']].plot(figsize=(20,7))
fx.set_title("High vs Low",fontsize=25)
```

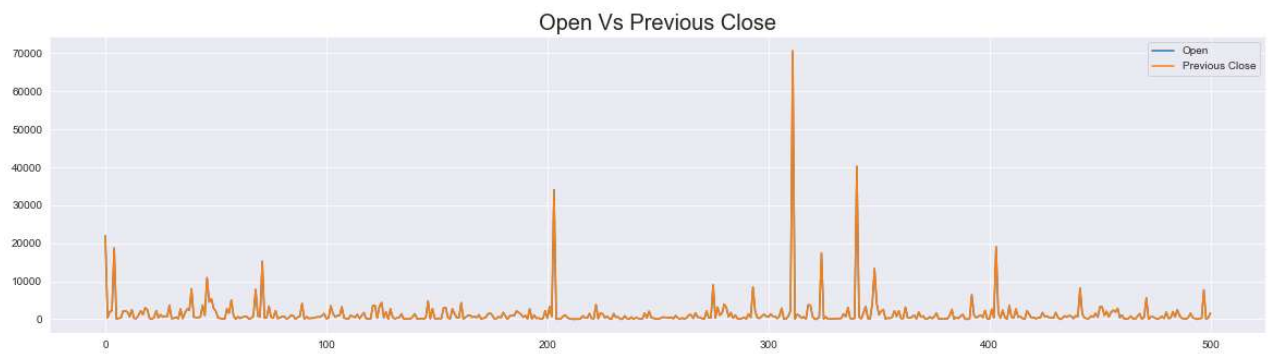
```
Out[27]: Text(0.5, 1.0, 'High vs Low')
```



```
In [28]: #Analysing Closing price vs Open Price

fx = data[['Open','Previous Close']].plot(figsize=(20,5))
fx.set_title("Open Vs Previous Close",fontsize=20)
```

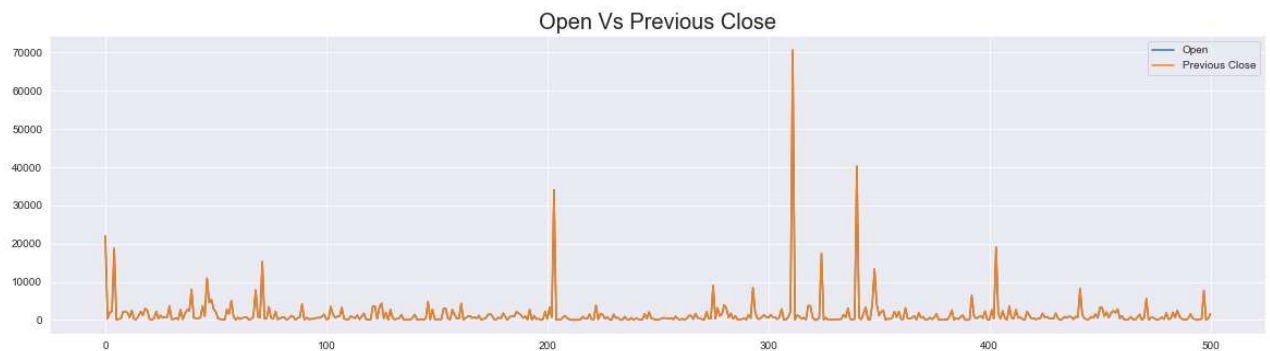
```
Out[28]: Text(0.5, 1.0, 'Open Vs Previous Close')
```



```
In [29]: #Analysing Closing price vs Open Price

fx = data[['Open', 'Previous Close']].plot(figsize=(20,5))
fx.set_title("Open Vs Previous Close", fontsize=20)
```

Out[29]: Text(0.5, 1.0, 'Open Vs Previous Close')



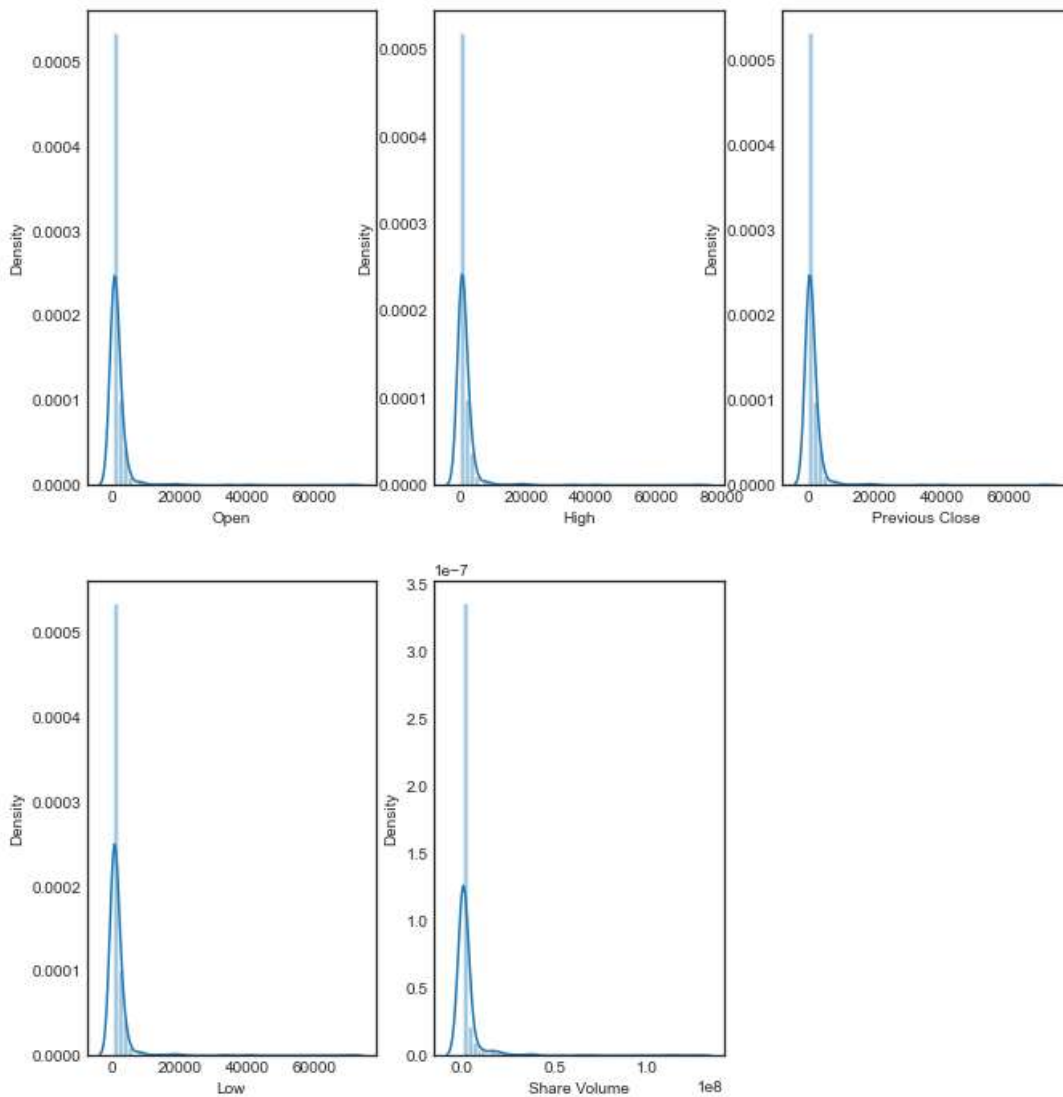
```
In [30]: #Distribution of Stock Measures

#Let witness the histogram distribution of the stock measures such as open,close,high,Low and volume
```

```
In [31]: columns = ['Open', 'High', 'Previous Close', 'Low', 'Share Volume']
di = {}
plt.figure(figsize=(20,7))
plt.style.use('seaborn-white')
d_x = 231
for i in columns:
    plt.subplot(d_x)
    sns.distplot(data[i])
    a_x = plt.gcf()
    a_x.set_size_inches(11,12)
    d_x+=1
    di[i] = data[i].skew()
print("Skewness of each measures")
print(di)
```

Skewness of each measures

```
{'Open': 10.0834494701177, 'High': 10.202033718688162, 'Previous Close': 10.146385305666179, 'Low': 10.201970935326168, 'Share Volume': 8.938483021288945}
```

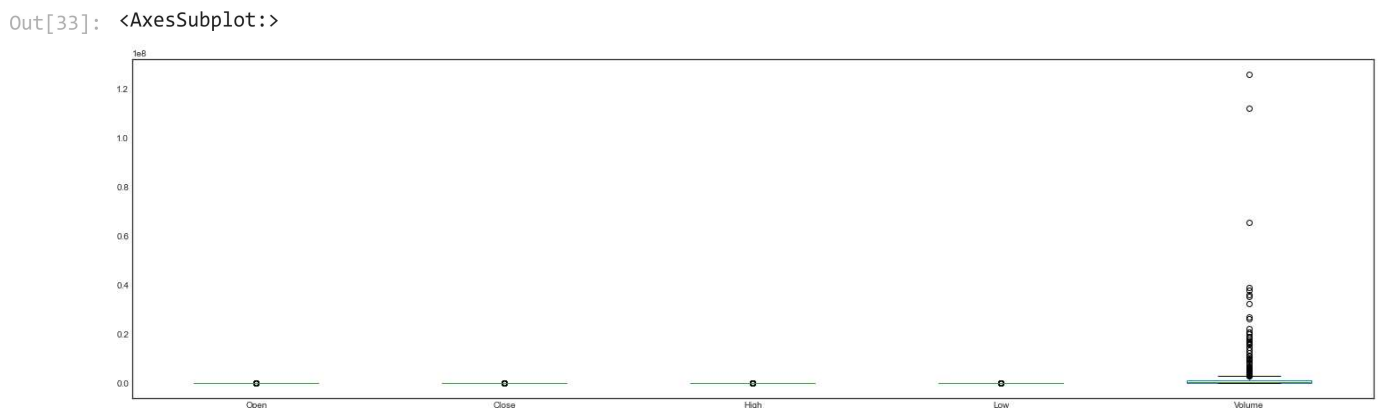


```
In [32]: #Insights:

#ALL the measures are exhibit equal distribution property
#ALL the Distributions are positivly Skewed
```

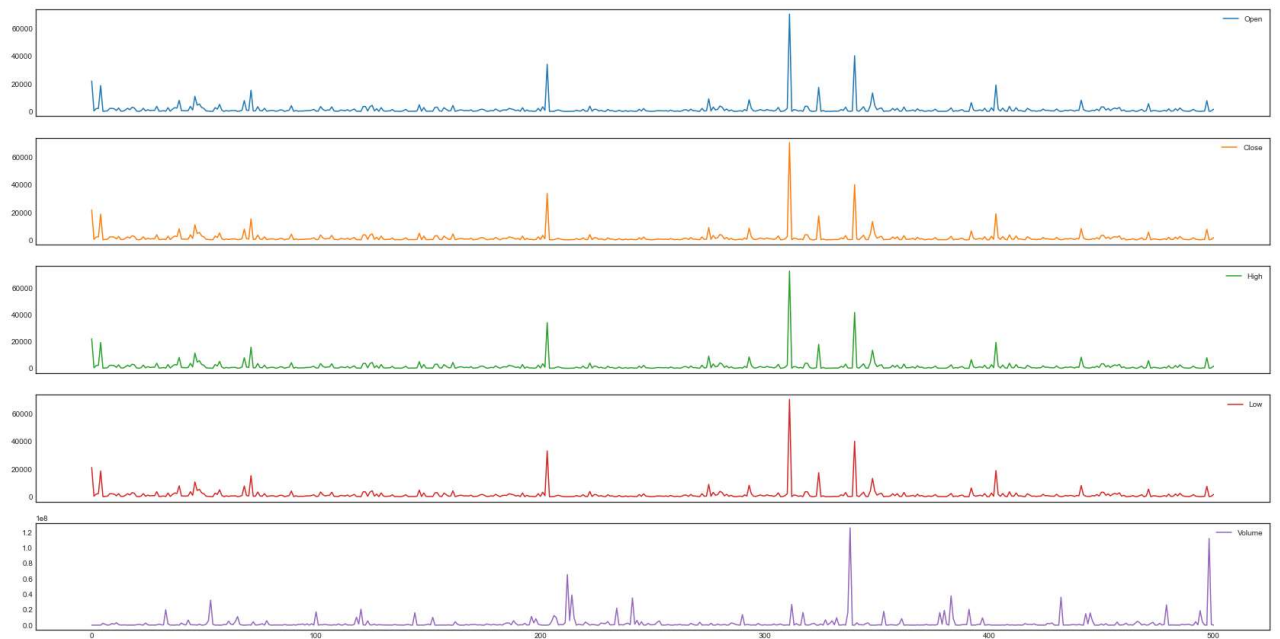
```
In [33]: #Box Plot of Features

df = pd.DataFrame({'Open':data['Open'],'Close':data['Previous Close'],'High':data['High'],'Low':data['Low'],'Volume':data['Share Volume']})
df.plot.box(figsize=(25,7))
```



```
In [34]: df.plot(subplots=True,figsize=(29,15))
```

Out[34]: array([<AxesSubplot:>, <AxesSubplot:>, <AxesSubplot:>, <AxesSubplot:>, <AxesSubplot:>], dtype=object)



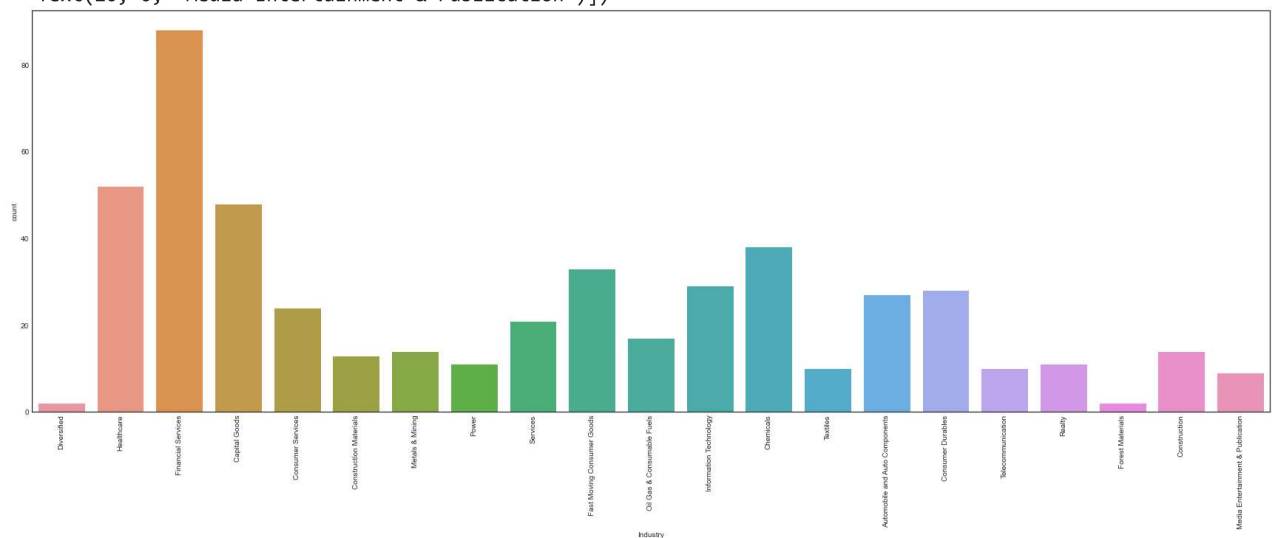
In [35]:

```
#Distribution of Industries

plt.figure(figsize=(30,10))
sns.countplot(data['Industry'])
plt.xticks(rotation=90)
```

Out[35]: (array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16,

```
17, 18, 19, 20]),
[Text(0, 0, 'Diversified'),
Text(1, 0, 'Healthcare'),
Text(2, 0, 'Financial Services'),
Text(3, 0, 'Capital Goods'),
Text(4, 0, 'Consumer Services'),
Text(5, 0, 'Construction Materials'),
Text(6, 0, 'Metals & Mining'),
Text(7, 0, 'Power'),
Text(8, 0, 'Services'),
Text(9, 0, 'Fast Moving Consumer Goods'),
Text(10, 0, 'Oil Gas & Consumable Fuels'),
Text(11, 0, 'Information Technology'),
Text(12, 0, 'Chemicals'),
Text(13, 0, 'Textiles'),
Text(14, 0, 'Automobile and Auto Components'),
Text(15, 0, 'Consumer Durables'),
Text(16, 0, 'Telecommunication'),
Text(17, 0, 'Realty'),
Text(18, 0, 'Forest Materials'),
Text(19, 0, 'Construction'),
Text(20, 0, 'Media Entertainment & Publication')])
```



In [36]:

```
#Observations
```

```
#Financial Services Contributes more value counts
#Forest Materials and Diversified are having low value counts
```

```
In [37]: data['Industry'].value_counts()
```

```
Out[37]: Financial Services      88
Healthcare                     52
Capital Goods                  48
Chemicals                      38
Fast Moving Consumer Goods    33
Information Technology         29
Consumer Durables             28
Automobile and Auto Components 27
Consumer Services             24
Services                      21
Oil Gas & Consumable Fuels    17
Metals & Mining                14
Construction                  14
Construction Materials        13
Power                         11
Realty                        11
Telecommunication             10
Textiles                      10
Media Entertainment & Publication 9
Forest Materials              2
Diversified                   2
Name: Industry, dtype: int64
```

```
In [ ]:
```