

```
# IMPORTANT: RUN THIS CELL IN ORDER TO IMPORT YOUR KAGGLE DATA SOURCES,
# THEN FEEL FREE TO DELETE THIS CELL.
# NOTE: THIS NOTEBOOK ENVIRONMENT DIFFERS FROM KAGGLE'S PYTHON
# ENVIRONMENT SO THERE MAY BE MISSING LIBRARIES USED BY YOUR
# NOTEBOOK.
import kagglehub
kabhishm_adani_group_stock_prices_20162023_path = kagglehub.dataset_download('kabhishm/adani-group-stock-prices-20162023')
acarmehmet_nvidia_historical_stock_data_path = kagglehub.dataset_download('acarmehmet/nvidia-historical-stock-data')

print('Data source import complete.')
```

```
# This Python 3 environment comes with many helpful analytics libraries installed
# It is defined by the kaggle/python Docker image: https://github.com/kaggle/docker-python
# For example, here's several helpful packages to load
```

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

```
# Input data files are available in the read-only "../input/" directory
# For example, running this (by clicking run or pressing Shift+Enter) will list all files under the input directory
```

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

```
# You can write up to 20GB to the current directory (/kaggle/working/) that gets preserved as output when you create a version using "Sav
# You can also write temporary files to /kaggle/temp/, but they won't be saved outside of the current session
```

```
📄 /kaggle/input/adani-group-stock-prices-20162023/ADANIIENT.NS.csv
/kaggle/input/nvidia-historical-stock-data/nvidia_historical_stock_data.csv
```

✓ LIBRARY IMPORTATION

```
#Importing of libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
import warnings
warnings.filterwarnings('ignore')
from pandas.plotting import lag_plot
```

✓ INDEX

INITIAL READING AND CLEANING PROCESS

EDA

MODEL PREPARATION

PREDICTIVE ANALYSIS

QUESTIONS

1. CLOSE VALUE OF NVIDIA STOCKS VISUAL
2. OPEN VALUE OF NVIDIA STOCKS VISUAL
3. CUMULATIVE SUM PRESENTING OPEN, HIGH, VOLUME, LOW, CLOSE
4. YEAR WISE CLOSE GRAPH
5. CORRELATION CHART

✓ INITIAL READING AND CLEANING PROCESS

```
#Reading of file
df = pd.read_csv('/kaggle/input/nvidia-historical-stock-data/nvidia_historical_stock_data.csv')
```

df



| | Date | Open | High | Low | Close | Volume |
|------|---------------------|------------|------------|------------|------------|-----------|
| 0 | 1999-10-21 00:00:00 | 0.038334 | 0.040244 | 0.037259 | 0.040244 | 382128000 |
| 1 | 1999-10-22 00:00:00 | 0.040125 | 0.042394 | 0.040125 | 0.041200 | 286752000 |
| 2 | 1999-10-25 00:00:00 | 0.040961 | 0.041559 | 0.039170 | 0.039170 | 213840000 |
| 3 | 1999-10-26 00:00:00 | 0.040842 | 0.044425 | 0.040603 | 0.042991 | 481248000 |
| 4 | 1999-10-27 00:00:00 | 0.043230 | 0.043469 | 0.040364 | 0.041081 | 214128000 |
| ... | ... | ... | ... | ... | ... | ... |
| 6285 | 2024-10-15 00:00:00 | 137.869995 | 138.570007 | 128.740005 | 131.600006 | 377831000 |
| 6286 | 2024-10-16 00:00:00 | 133.979996 | 136.619995 | 131.580002 | 135.720001 | 264879700 |
| 6287 | 2024-10-17 00:00:00 | 139.339996 | 140.889999 | 136.869995 | 136.929993 | 306435900 |
| 6288 | 2024-10-18 00:00:00 | 138.669998 | 138.899994 | 137.279999 | 138.000000 | 176090200 |
| 6289 | 2024-10-21 00:00:00 | 138.130005 | 143.710007 | 138.000000 | 143.710007 | 264554500 |

6290 rows x 6 columns

```
#Getting info on data
df.info()
```



```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6290 entries, 0 to 6289
Data columns (total 6 columns):
#   Column  Non-Null Count  Dtype
---  ---
0    Date    6290 non-null    object
1    Open    6290 non-null    float64
2    High    6290 non-null    float64
3    Low     6290 non-null    float64
4    Close   6290 non-null    float64
5    Volume  6290 non-null    int64
dtypes: float64(4), int64(1), object(1)
memory usage: 295.0+ KB
```

```
#Getting stats on numeric data
df.describe()
```

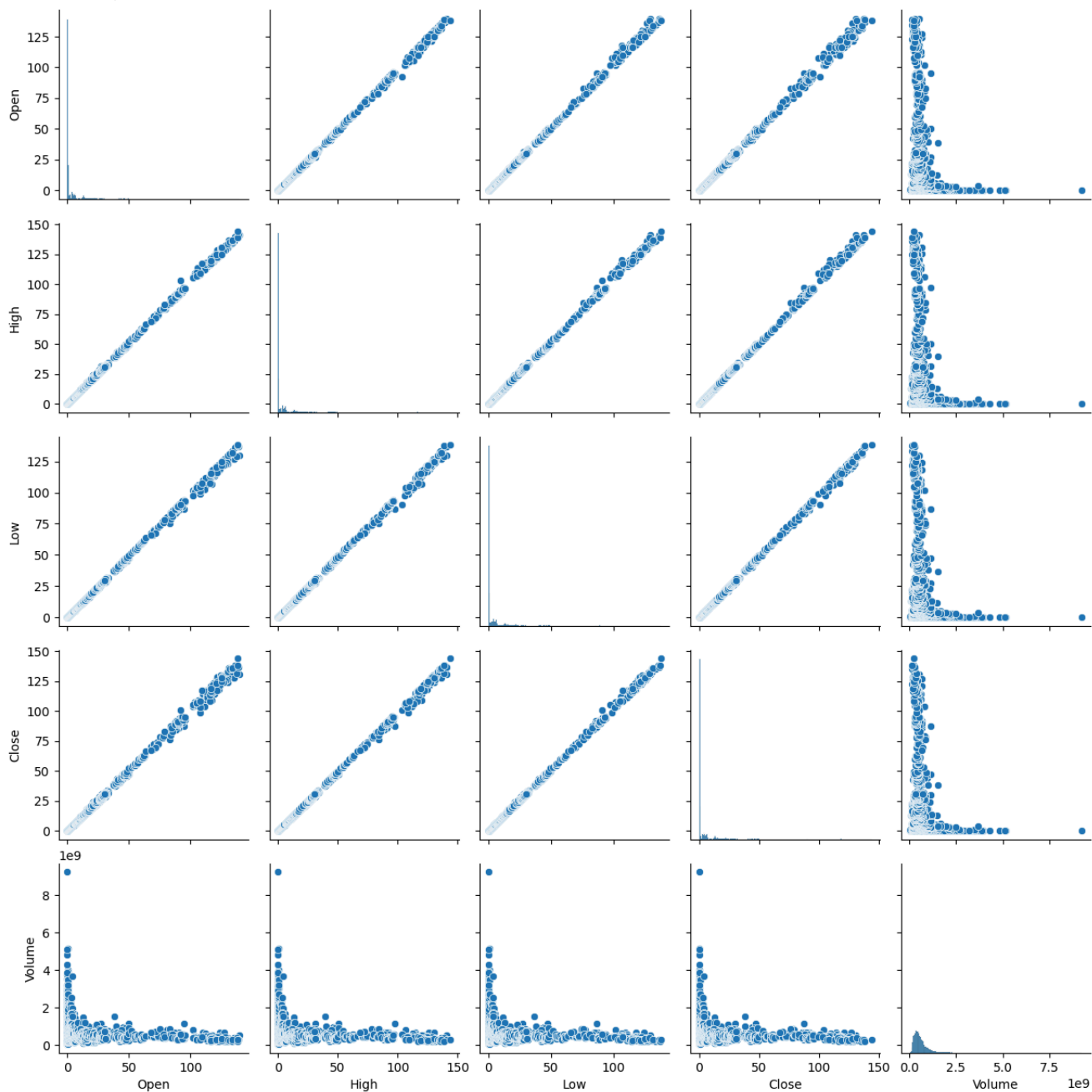


| | Open | High | Low | Close | Volume |
|-------|-------------|-------------|-------------|-------------|--------------|
| count | 6290.000000 | 6290.000000 | 6290.000000 | 6290.000000 | 6.290000e+03 |
| mean | 7.413215 | 7.552883 | 7.263899 | 7.415679 | 6.152477e+08 |
| std | 19.579643 | 19.955825 | 19.159134 | 19.577934 | 4.295346e+08 |
| min | 0.038334 | 0.040244 | 0.037259 | 0.039170 | 4.564400e+07 |
| 25% | 0.270100 | 0.277495 | 0.262916 | 0.269890 | 3.549901e+08 |
| 50% | 0.442333 | 0.448562 | 0.435627 | 0.442142 | 5.131920e+08 |
| 75% | 4.512542 | 4.586693 | 4.463839 | 4.539182 | 7.403648e+08 |
| max | 139.788047 | 143.710007 | 138.000000 | 143.710007 | 9.230856e+09 |

EDA

```
#PAIRPLOT
sns.pairplot(df)
```

<seaborn.axisgrid.PairGrid at 0x7879c6d42da0>



```
# ANSWER 1 NVIDIA CLOSE VALUE
```

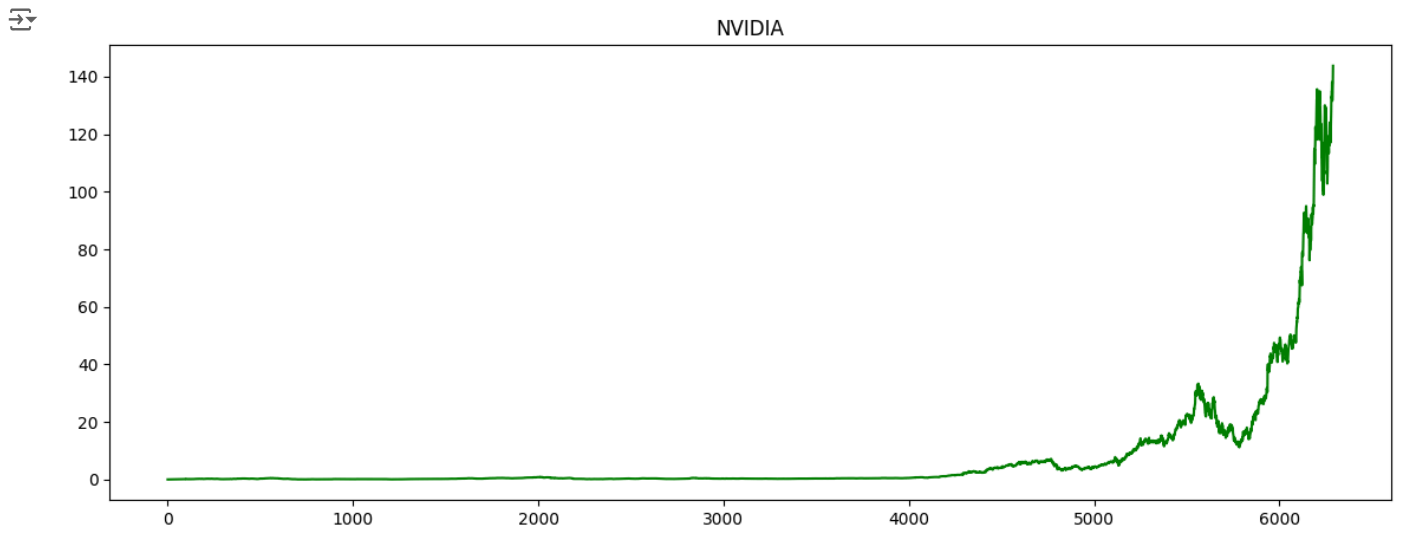
```
#Close plot of nvidia
```

```
plt.figure(figsize=(14,5))
```

```
plt.plot(df[['Close']],color="green")
```

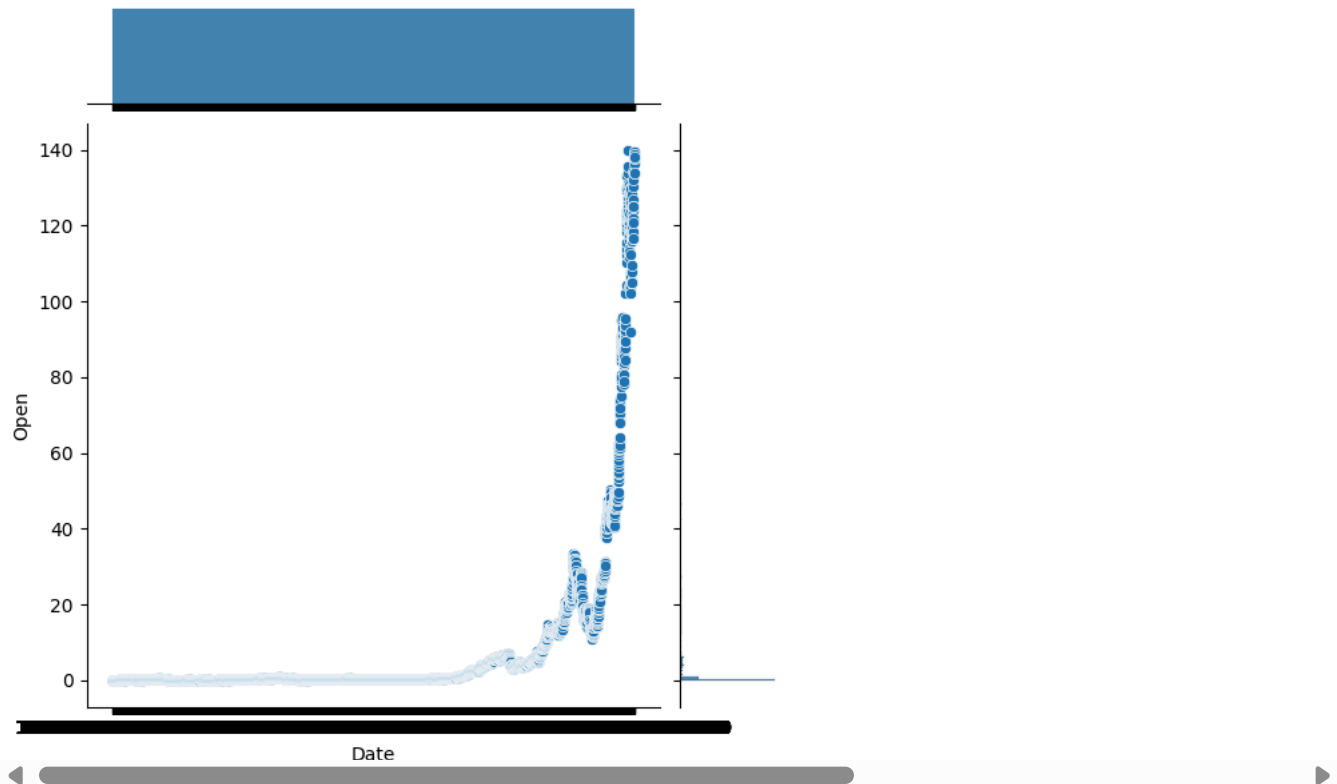
```
plt.title('NVIDIA')
```

```
plt.show()
```



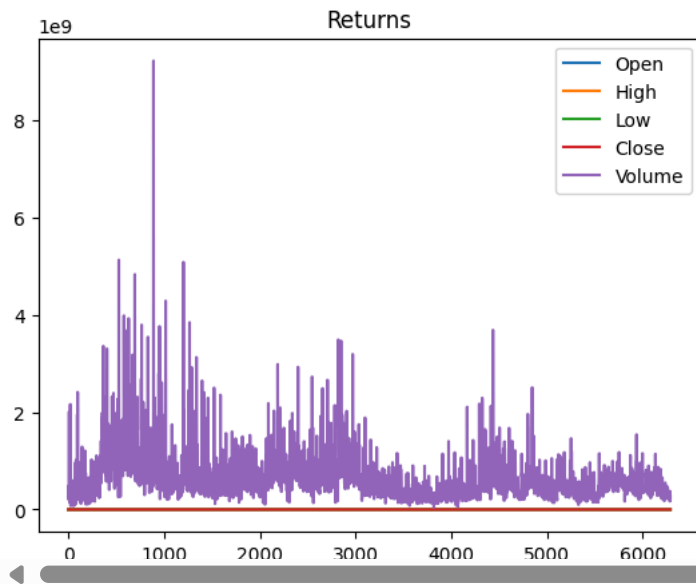
```
# ANSWER 1 NVIDIA OPEN VALUE VISUALIZATION  
#JOIN PLOT  
sns.jointplot(data=df, x='Date', y="Open")
```

```
<seaborn.axisgrid.JointGrid at 0x7879c2e85000>
```



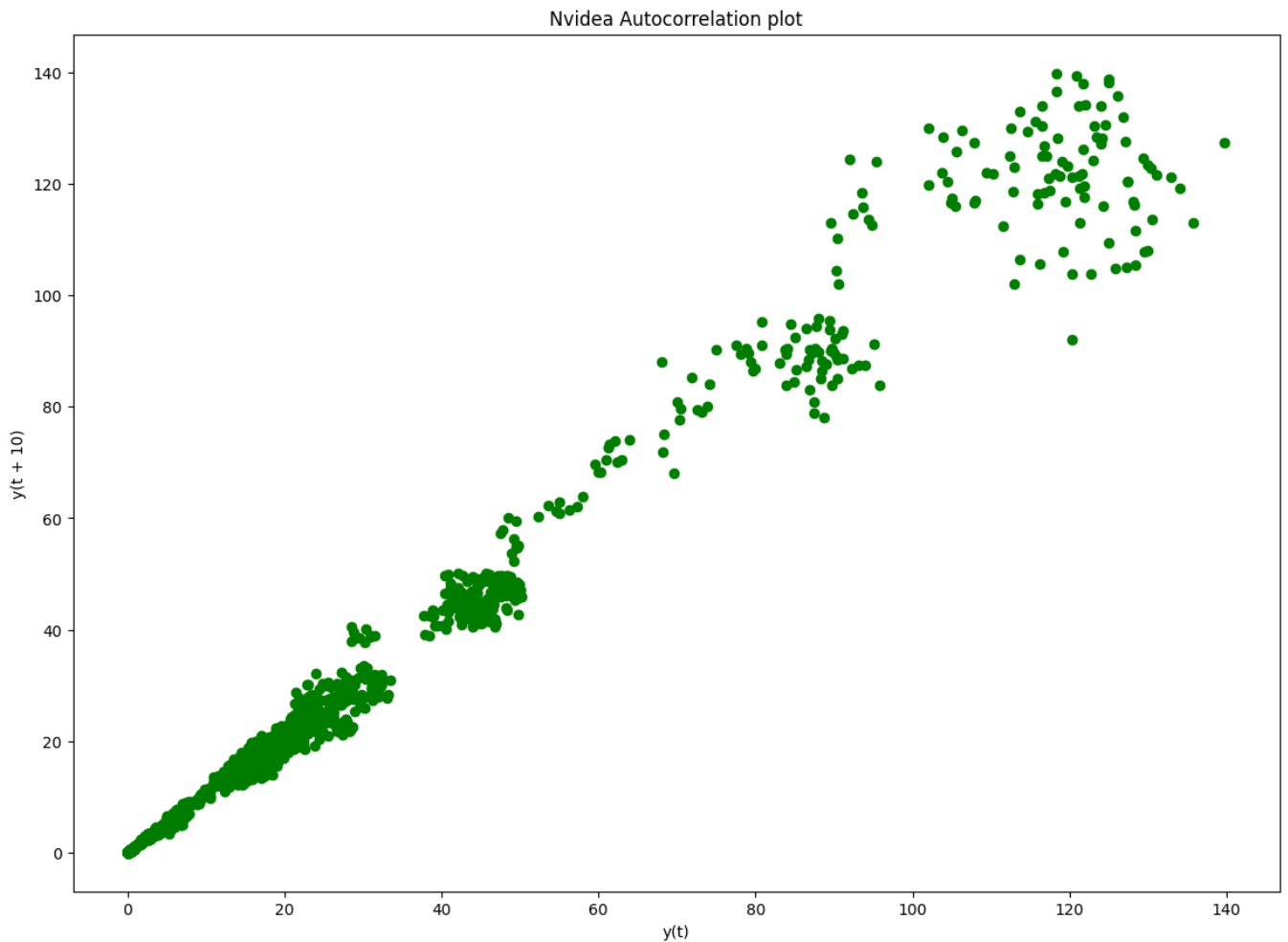
```
# ANSWER 3 CUMULATIVE SUM PLOT  
dret = df.cumsum()  
df.plot()  
plt.title('Returns')
```

Text(0.5, 1.0, 'Returns')



```
#LAG Plot
plt.figure(figsize=(14,10))
lag_plot(df['Open'], lag=10,c="green")
plt.title('Nvidea Autocorrelation plot')
```

Text(0.5, 1.0, 'Nvidea Autocorrelation plot')



✓ MODEL PREPRATION

Splitting of Date column into splits of year month day and Day of week

```
#Converting date as object to date as date time
df['Date'] = pd.to_datetime(df['Date'])
```

```
#Converting object to year,month ,day and day of week one by one
df['Year'] = df['Date'].dt.year
df['Month'] = df['Date'].dt.month
df['Day'] = df['Date'].dt.day
df['DayOfWeek'] = df['Date'].dt.dayofweek
```

```
#Simplifying more to quarter and day of year
df['Quarter'] = df['Date'].dt.quarter
df['DayOfYear'] = df['Date'].dt.dayofyear
```

```
#Dropping of initial dtae column
df.drop(columns=['Date'], inplace=True)
```

df



| | Open | High | Low | Close | Volume | Year | Month | Day | DayOfWeek | Quarter | DayOfYear |
|------|------------|------------|------------|------------|-----------|------|-------|-----|-----------|---------|-----------|
| 0 | 0.038334 | 0.040244 | 0.037259 | 0.040244 | 382128000 | 1999 | 10 | 21 | 3 | 4 | 294 |
| 1 | 0.040125 | 0.042394 | 0.040125 | 0.041200 | 286752000 | 1999 | 10 | 22 | 4 | 4 | 295 |
| 2 | 0.040961 | 0.041559 | 0.039170 | 0.039170 | 213840000 | 1999 | 10 | 25 | 0 | 4 | 298 |
| 3 | 0.040842 | 0.044425 | 0.040603 | 0.042991 | 481248000 | 1999 | 10 | 26 | 1 | 4 | 299 |
| 4 | 0.043230 | 0.043469 | 0.040364 | 0.041081 | 214128000 | 1999 | 10 | 27 | 2 | 4 | 300 |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 6285 | 137.869995 | 138.570007 | 128.740005 | 131.600006 | 377831000 | 2024 | 10 | 15 | 1 | 4 | 289 |
| 6286 | 133.979996 | 136.619995 | 131.580002 | 135.720001 | 264879700 | 2024 | 10 | 16 | 2 | 4 | 290 |
| 6287 | 139.339996 | 140.889999 | 136.869995 | 136.929993 | 306435900 | 2024 | 10 | 17 | 3 | 4 | 291 |
| 6288 | 138.669998 | 138.899994 | 137.279999 | 138.000000 | 176090200 | 2024 | 10 | 18 | 4 | 4 | 292 |
| 6289 | 138.130005 | 143.710007 | 138.000000 | 143.710007 | 264554500 | 2024 | 10 | 21 | 0 | 4 | 295 |

6290 rows x 11 columns

df.info()



```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6290 entries, 0 to 6289
Data columns (total 11 columns):
#   Column      Non-Null Count  Dtype
---  ---
0   Open        6290 non-null   float64
1   High        6290 non-null   float64
2   Low         6290 non-null   float64
3   Close       6290 non-null   float64
4   Volume      6290 non-null   int64
5   Year        6290 non-null   int32
6   Month       6290 non-null   int32
7   Day         6290 non-null   int32
8   DayOfWeek   6290 non-null   int32
9   Quarter     6290 non-null   int32
10  DayOfYear   6290 non-null   int32
dtypes: float64(4), int32(6), int64(1)
memory usage: 393.2 KB
```

FLITER DATA FROM 2018-2024

```
yn =df[(df['Year'] > 2017)]
```

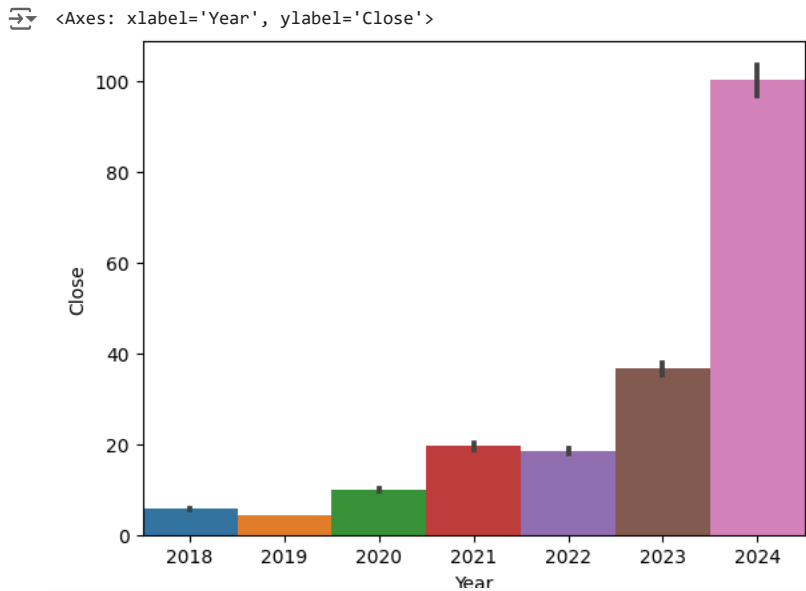
yn

| | Open | High | Low | Close | Volume | Year | Month | Day | DayOfWeek | Quarter | DayOfYear |
|------|------------|------------|------------|------------|-----------|------|-------|-----|-----------|---------|-----------|
| 4578 | 4.841929 | 4.933930 | 4.810273 | 4.930220 | 355616000 | 2018 | 1 | 2 | 1 | 1 | 2 |
| 4579 | 5.047694 | 5.285116 | 5.039038 | 5.254696 | 914704000 | 2018 | 1 | 3 | 2 | 1 | 3 |
| 4580 | 5.336064 | 5.392699 | 5.260138 | 5.282397 | 583268000 | 2018 | 1 | 4 | 3 | 1 | 4 |
| 4581 | 5.297236 | 5.364505 | 5.220321 | 5.327161 | 580124000 | 2018 | 1 | 5 | 4 | 1 | 5 |
| 4582 | 5.450818 | 5.564582 | 5.405806 | 5.490388 | 881216000 | 2018 | 1 | 8 | 0 | 1 | 8 |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 6285 | 137.869995 | 138.570007 | 128.740005 | 131.600006 | 377831000 | 2024 | 10 | 15 | 1 | 4 | 289 |
| 6286 | 133.979996 | 136.619995 | 131.580002 | 135.720001 | 264879700 | 2024 | 10 | 16 | 2 | 4 | 290 |
| 6287 | 139.339996 | 140.889999 | 136.869995 | 136.929993 | 306435900 | 2024 | 10 | 17 | 3 | 4 | 291 |
| 6288 | 138.669998 | 138.899994 | 137.279999 | 138.000000 | 176090200 | 2024 | 10 | 18 | 4 | 4 | 292 |
| 6289 | 138.130005 | 143.710007 | 138.000000 | 143.710007 | 264554500 | 2024 | 10 | 21 | 0 | 4 | 295 |

1712 rows x 11 columns

#ANSWER 4

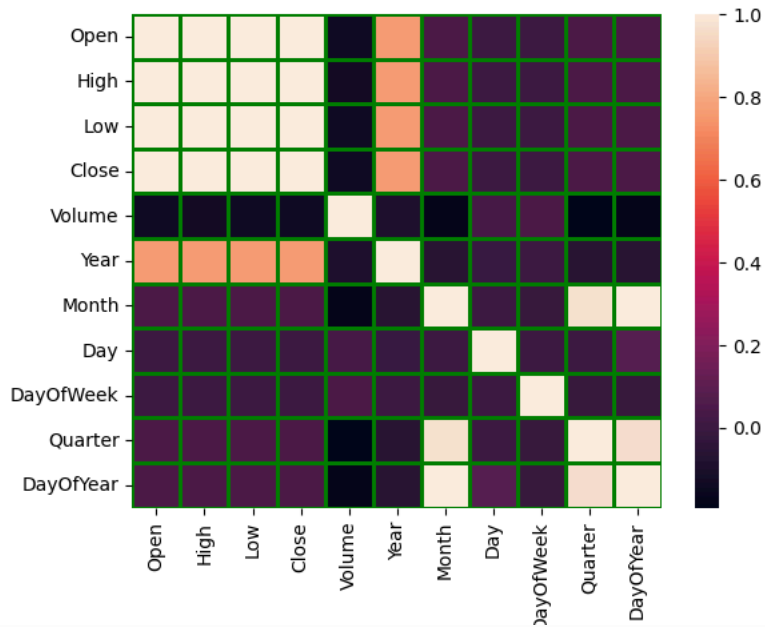
```
#Creating of bar plot as per year and close
sns.barplot(x='Year',y='Close',data=yn,width=1.0)
```



#ANSWER 5

```
#Correlation chart
coreal=yn.corr()
linewidths = 2
linecolor = "green"
sns.heatmap(coreal,linewidths=linewidths,
            linecolor=linecolor)
```

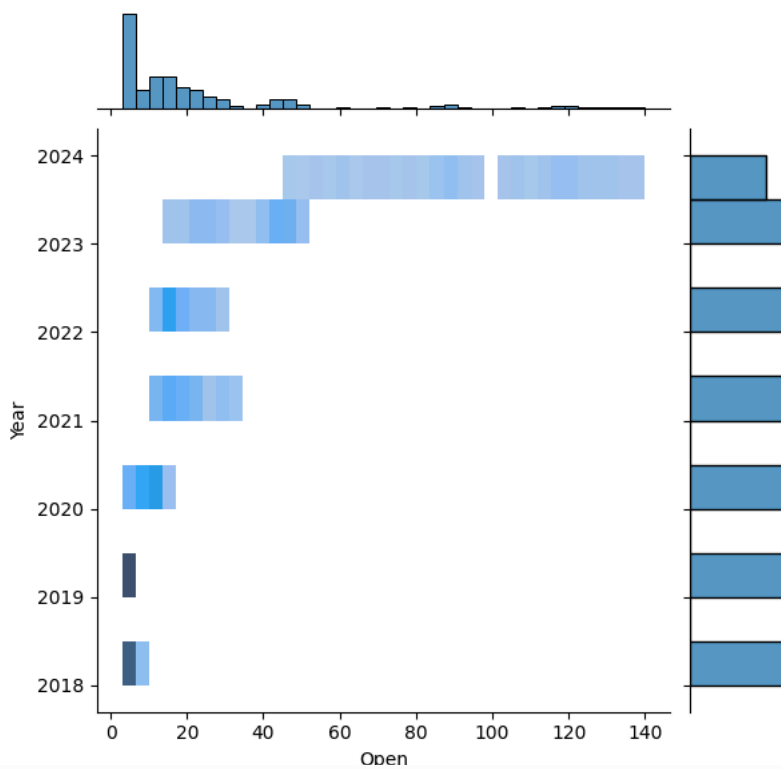
<Axes: >




#

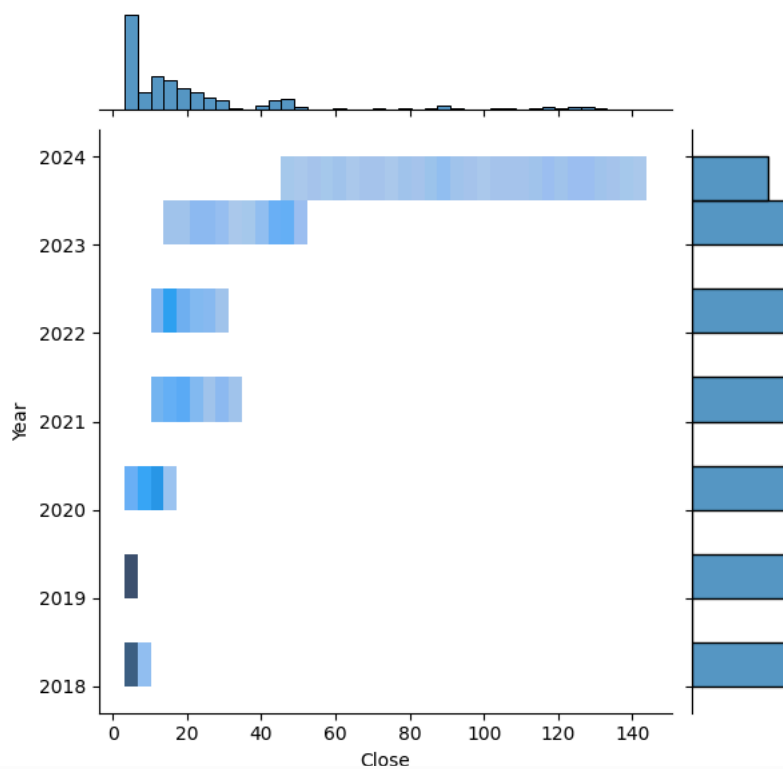
```
#Join plot for open and year  
sns.jointplot(data=yn, x="Open", y="Year", kind="hist")
```

<seaborn.axisgrid.JointGrid at 0x7879b02e9930>




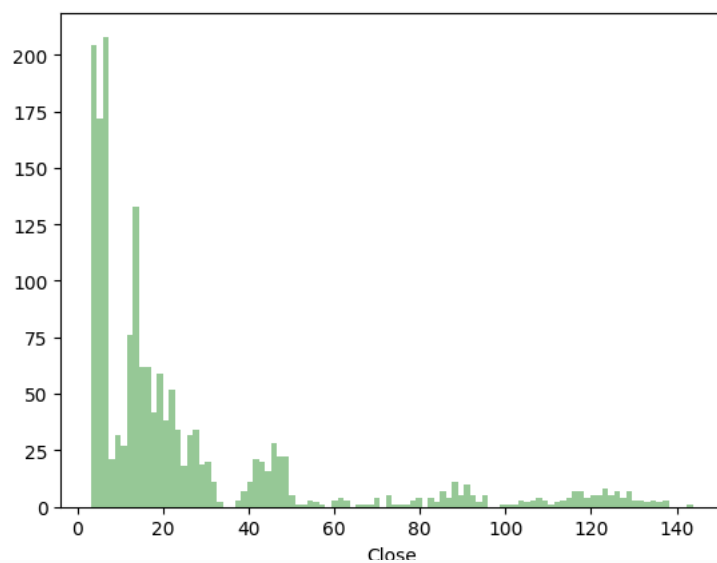
```
#Join plot for close and year  
sns.jointplot(data=yn, x="Close", y="Year", kind="hist")
```


 <seaborn.axisgrid.JointGrid at 0x7879ae425e40>



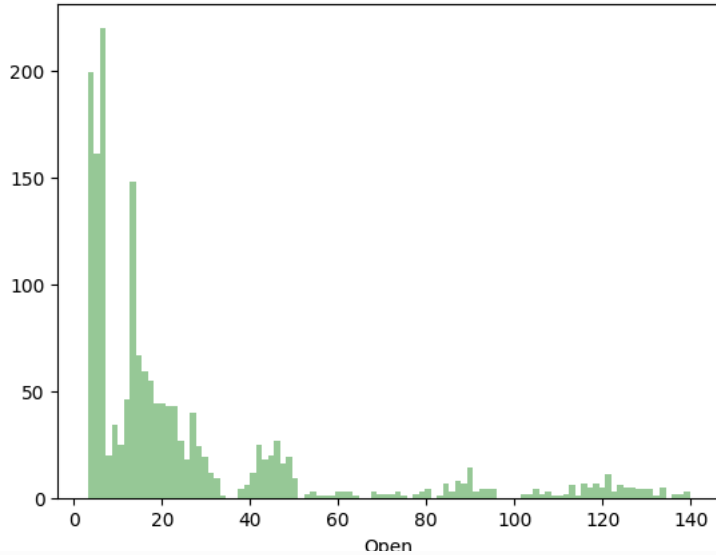
```
#Distplot for univariate distribution  
sns.distplot(yn['Close'],bins=100,kde=False,color='green')
```

 <Axes: xlabel='Close'>



```
#Distplot for univariate distribution  
sns.distplot(yn['Open'],bins=100,kde=False,color='green')
```

<Axes: xlabel='Open'>



✓ PREDICTIVE ANALYSIS MODEL

yn

| | Open | High | Low | Close | Volume | Year | Month | Day | DayOfWeek | Quarter | DayOfYear |
|------|------------|------------|------------|------------|-----------|------|-------|-----|-----------|---------|-----------|
| 4578 | 4.841929 | 4.933930 | 4.810273 | 4.930220 | 355616000 | 2018 | 1 | 2 | 1 | 1 | 2 |
| 4579 | 5.047694 | 5.285116 | 5.039038 | 5.254696 | 914704000 | 2018 | 1 | 3 | 2 | 1 | 3 |
| 4580 | 5.336064 | 5.392699 | 5.260138 | 5.282397 | 583268000 | 2018 | 1 | 4 | 3 | 1 | 4 |
| 4581 | 5.297236 | 5.364505 | 5.220321 | 5.327161 | 580124000 | 2018 | 1 | 5 | 4 | 1 | 5 |
| 4582 | 5.450818 | 5.564582 | 5.405806 | 5.490388 | 881216000 | 2018 | 1 | 8 | 0 | 1 | 8 |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 6285 | 137.869995 | 138.570007 | 128.740005 | 131.600006 | 377831000 | 2024 | 10 | 15 | 1 | 4 | 289 |
| 6286 | 133.979996 | 136.619995 | 131.580002 | 135.720001 | 264879700 | 2024 | 10 | 16 | 2 | 4 | 290 |
| 6287 | 139.339996 | 140.889999 | 136.869995 | 136.929993 | 306435900 | 2024 | 10 | 17 | 3 | 4 | 291 |
| 6288 | 138.669998 | 138.899994 | 137.279999 | 138.000000 | 176090200 | 2024 | 10 | 18 | 4 | 4 | 292 |
| 6289 | 138.130005 | 143.710007 | 138.000000 | 143.710007 | 264554500 | 2024 | 10 | 21 | 0 | 4 | 295 |

1712 rows x 11 columns

#TRAIN AND TEST SPLIT

#Importing of necessary library

```
from sklearn.model_selection import train_test_split
```

```
x = yn[['Open', 'High', 'Low', 'Close', 'Volume', 'Year', 'Month', 'Day', 'DayOfWeek',  
        'Quarter', 'DayOfYear']]
```

```
y = yn['Close']
```

x.shape

(1712, 11)

y.shape


(1712,)

```
x_train ,x_test ,y_train ,y_test = train_test_split(x,y,random_state=40,test_size=0.2)
```


x_train.shape

(1369, 11)


x_test.shape

 (343, 11)

y_train.shape

 (1369,)

y_test.shape

 (343,)

x_train.head()



| | Open | High | Low | Close | Volume | Year | Month | Day | DayOfWeek | Quarter | DayOfYear |
|------|-----------|-----------|-----------|-----------|------------|------|-------|-----|-----------|---------|-----------|
| 4847 | 3.376587 | 3.424948 | 3.249112 | 3.263745 | 1153932000 | 2019 | 1 | 29 | 1 | 1 | 29 |
| 5883 | 23.397570 | 23.615464 | 22.714903 | 22.953787 | 474866000 | 2023 | 3 | 10 | 4 | 1 | 69 |
| 4885 | 4.365909 | 4.430208 | 4.247985 | 4.314271 | 525212000 | 2019 | 3 | 25 | 0 | 1 | 84 |
| 4676 | 5.949849 | 6.130861 | 5.949106 | 6.129622 | 442956000 | 2018 | 5 | 23 | 2 | 2 | 143 |
| 6077 | 48.183650 | 49.303300 | 48.109674 | 48.879509 | 4709480000 | 2023 | 12 | 15 | 4 | 4 | 340 |

Building of Model

```
#Importing of necessary liabrary
from sklearn.ensemble import RandomForestRegressor
rfr=RandomForestRegressor(n_estimators=100, random_state=40)
```

```
rfr.fit(x_train,y_train)
```



```
RandomForestRegressor
RandomForestRegressor(random state=40)
```

```
y_pred=rfr.predict(x_test)
```