```
In [1]:
import os
                                                                                    In [2]:
import pandas as pd
from pandas import DataFrame
import numpy as np
                                                                                    In [3]:
import urllib
from urllib.request import urlretrieve
                                                                                    In [4]:
url="https://github.com/swapniljariwala/nsepy"
                                                                                    In [5]:
from nsepy import get history
from datetime import date
                                                                                    In [6]:
nse tcs=get history(symbol="TCS",start=date(2015,1,1),end=date(2015,12,31))
                                                                                    In [7]:
nse_tcs.columns
                                                                                    Out[7]:
Index(['Symbol', 'Series', 'Prev Close', 'Open', 'High', 'Low', 'Last',
       'Close', 'VWAP', 'Volume', 'Turnover', 'Trades', 'Deliverable Volume',
      '%Deliverble'],
     dtype='object')
                                                                                    In [8]:
nse_infy=get_history(symbol="INFY",start=date(2015,1,1),end=date(2015,12,31))
                                                                                    In [9]:
nse_infy.columns
                                                                                    Out[9]:
Index(['Symbol', 'Series', 'Prev Close', 'Open', 'High', 'Low', 'Last',
      'Close', 'VWAP', 'Volume', 'Turnover', 'Trades', 'Deliverable Volume',
      '%Deliverble'],
```

dtype='object')

```
In [10]:
```

```
nifty_it=get_history(symbol="NIFTY IT",start=date(2015,1,1),end=date(2015,12,31))
```

In [11]:

nifty_it.columns

Out[11]:

In [12]:

nse_infy.head()

Out[12]:

	Symbol	Series	Prev Close	Open	High	Low	Last	Close	VWAP	Volume	^
Date											
2015- 01-01	INFY	EQ	1972.55	1968.95	1982.00	1956.9	1971.00	1974.40	1971.34	500691	9.8
2015- 01-02	INFY	EQ	1974.40	1972.00	2019.05	1972.0	2017.95	2013.20	2003.25	1694580	3.3
2015- 01-05	INFY	EQ	2013.20	2009.90	2030.00	1977.5	1996.00	1995.90	2004.59	2484256	4.9
2015- 01-06	INFY	EQ	1995.90	1980.00	1985.00	1934.1	1965.10	1954.20	1954.82	2416829	4.7
2015- 01-07	INFY	EQ	1954.20	1965.00	1974.75	1950.0	1966.05	1963.55	1962.59	1812479	3.5

In [13]:

Out[13]:

	Symbol	Series	Prev Close	Open	High	Low	Last	Close	VWAP	Volume	^
Date											
2015- 01-01	TCS	EQ	2558.25	2567.0	2567.00	2541.00	2550.00	2545.55	2548.51	183415	4.6
2015- 01-02	TCS	EQ	2545.55	2551.0	2590.95	2550.60	2588.40	2579.45	2568.19	462870	1.1
2015- 01-05	TCS	EQ	2579.45	2581.0	2599.90	2524.65	2538.10	2540.25	2563.94	877121	2.2
2015- 01-06	TCS	EQ	2540.25	2529.1	2529.10	2440.00	2450.05	2446.60	2466.90	1211892	2.9
2015-	TCS	EQ	2446.60	2470.0	2479.15	2407.45	2426.90	2417.70	2433.96	1318166	3.2

In [14]:

nifty_it.head()

Out[14]:

	Symbol	Series	Prev Close	Open	High	Low	Last	Close	VWAP	Volume	Turnover	Trades	C ^
Date													
<					-						-	>	

In [15]:

nse_infy.shape

Out[15]:

(248, 14)

In [16]:

nse_tcs.shape

Out[16]:

(248, 14)

In [17]:

Out[17]:

	1							٠,
	Prev Close	Open	High	Low	Last	Close	V	_
count	248.000000	248.000000	248.000000	248.000000	248.000000	248.000000	248.000	(
mean	1551.474798	1550.506855	1566.266532	1530.085887	1548.084879	1547.978226	1548.133	
std	529.396894	530.578342	534.714088	524.194873	529.493276	529.468189	528.861	Į
min	937.500000	941.000000	952.100000	932.650000	935.500000	937.500000	941.1800	(
25%	1085.912500	1088.000000	1099.975000	1067.150000	1086.875000	1085.912500	1085.90	
50%	1149.650000	1150.000000	1159.725000	1131.150000	1145.625000	1149.325000	1146.24	Į
75%	2125.312500	2136.137500	2150.000000	2104.500000	2125.250000	2125.312500	2125.082	
max	2324 700000	2328 500000	2336 000000	2292 050000	2323 200000	2324 700000	2322 170	ſ

In [18]:

nse_tcs.describe()

Out[18]:

	ı	ı	1	I	I	1		
	Prev Close	Open	High	Low	Last	Close	V	
count	248.000000	248.000000	248.000000	248.000000	248.000000	248.000000	248.0000	
mean	2538.207460	2542.172782	2563.580444	2514.408468	2538.039718	2537.717944	2538.432	
std	86.829359	87.605699	90.598368	82.952778	86.849305	87.057814	86.8130	;
min	2319.800000	2319.400000	2343.900000	2315.250000	2321.000000	2319.800000	2322.270	
25%	2495.312500	2499.500000	2518.900000	2472.100000	2497.500000	2495.150000	2496.668	
50%	2543.050000	2548.500000	2566.000000	2520.000000	2540.150000	2541.475000	2540.44	
75%	2592.000000	2594.250000	2615.750000	2567.300000	2593.425000	2592.000000	2592.607	
max	2776.000000	2788.000000	2812.100000	2721.900000	2785.100000	2776.000000	2763.040	

In [19]:

Out[19]:

Symbol	0
Series	0
Prev Close	0
Open	0
High	0
Low	0
Last	0
Close	0
VWAP	0
Volume	0
Turnover	0
Trades	0
Deliverable Volume	0
%Deliverble	0
dtype: int64	

In [20]:

nse tcs.isnull().sum()
115C_CC3.1511U11().5U11()

Out[20]:

Symbol	0
Series	0
Prev Close	0
Open	0
High	0
Low	0
Last	0
Close	0
VWAP	0
Volume	0
Turnover	0
Trades	0
Deliverable Volume	0
%Deliverble	0
dtype: int64	

In [21]:

```
<class 'pandas.core.frame.DataFrame'>
Index: 248 entries, 2015-01-01 to 2015-12-31
Data columns (total 14 columns):
Symbol
                      248 non-null object
Series
                     248 non-null object
Prev Close
                     248 non-null float64
Open
                     248 non-null float64
                     248 non-null float64
High
Low
                     248 non-null float64
                     248 non-null float64
Last
Close
                     248 non-null float64
VWAP
                     248 non-null float64
                     248 non-null int64
Volume
Turnover
                     248 non-null float64
Trades
                     248 non-null int64
Deliverable Volume
                     248 non-null int64
%Deliverble
                     248 non-null float64
dtypes: float64(9), int64(3), object(2)
memory usage: 29.1+ KB
```

In [22]:

```
nse_tcs.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 248 entries, 2015-01-01 to 2015-12-31
Data columns (total 14 columns):
Symbol
                      248 non-null object
Series
                      248 non-null object
Prev Close
                     248 non-null float64
Open
                      248 non-null float64
                      248 non-null float64
High
Low
                      248 non-null float64
Last
                     248 non-null float64
                      248 non-null float64
Close
VWAP
                     248 non-null float64
                      248 non-null int64
Volume
Turnover
                     248 non-null float64
Trades
                     248 non-null int64
Deliverable Volume
                     248 non-null int64
%Deliverble
                      248 non-null float64
dtypes: float64(9), int64(3), object(2)
memory usage: 29.1+ KB
```

In [23]:

nifty it.shape

Out[23]:

```
In [24]:
nse infy['Series'].unique()
                                                                                    Out[24]:
array(['EQ'], dtype=object)
                                                                                    In [25]:
nse_tcs['Series'].unique()
                                                                                    Out[25]:
array(['EQ'], dtype=object)
                                                                                    In [26]:
def movingaverage(x, w):
   return pd.Series(x.rolling(window=w,min periods=0).mean())
                                                                                    In [27]:
nse infy['4weeks']=movingaverage(nse infy['Close'],20)
nse infy['16weeks']=movingaverage(nse infy['Close'],80)
nse infy['28weeks']=movingaverage(nse infy['Close'],140)
nse infy['40weeks']=movingaverage(nse infy['Close'],200)
nse_infy['52weeks']=movingaverage(nse_infy['Close'],260)
nse tcs['4weeks']=movingaverage(nse tcs['Close'],20)
nse tcs['16weeks']=movingaverage(nse tcs['Close'],80)
nse tcs['28weeks']=movingaverage(nse tcs['Close'],140)
nse tcs['40weeks']=movingaverage(nse tcs['Close'],200)
nse tcs['52weeks']=movingaverage(nse tcs['Close'],260)
                                                                                    In [28]:
nse infy[['Close','4weeks','16weeks','28weeks','40weeks','52weeks']].tail()
                                                                                    Out[28]:
             Close
                                 16weeks
                                              28weeks
                                                          40weeks
                                                                       52weeks
                     4weeks
      Date
2015-12-24
           1096.35
                   1071.865 1099.681250 1124.790357
                                                       1418.08950
                                                                   1555.317418
2015-12-28 | 1104.75 | 1073.785
                              1099.955625
                                          1118.195357
                                                       1412.37525
                                                                   1553,478367
2015-12-29 | 1104.70 | 1074.625 | 1099.876250 | 1111.717143
                                                       1406.94850
                                                                  1551.654065
2015-12-30 | 1086.30
                   1074.990 1099.775000 1105.190357
                                                       1401.39000
                                                                   1549.770040
```

2015-12-31 | 1105.40

1077.225

1099.980000

1396.00600

1547.978226

1098.856786

```
nse_tcs[['Close','4weeks','16weeks','28weeks','40weeks','52weeks']].tail()
```

Out[29]:

	Close	4weeks	16weeks	28weeks	40weeks	52weeks
Date						
2015-12-24	2434.25	2382.1025	2489.293750	2523.807500	2534.58675	2539.263934
2015-12-28	2462.70	2387.5700	2487.894375	2522.786429	2533.41800	2538.951429
2015-12-29	2455.80	2392.1250	2486.456250	2521.928929	2532.46300	2538.613415
2015-12-30	2418.30	2394.9450	2484.634375	2520.792143	2531.34100	2538.126316
2015-12-31	2436.85	2398.5275	2483.377500	2519.900714	2530.49250	2537.717944

In [30]:

nse_infy.tail()

Out[30]:

	Symbol	Series	Prev Close	Open	High	Low	Last	Close	VWAP	Volume	^
Date											
2015- 12-24	INFY	EQ	1100.85	1102.05	1104.45	1090.1	1095.8	1096.35	1095.84	615027	6.73
2015- 12-28	INFY	EQ	1096.35	1090.00	1110.00	1090.0	1103.8	1104.75	1105.51	2763476	3.05
2015- 12-29	INFY	EQ	1104.75	1101.25	1110.00	1097.3	1103.0	1104.70	1104.14	1672531	1.84
2015- 12-30	INFY	EQ	1104.70	1101.10	1106.60	1083.0	1088.0	1086.30	1094.45	2576985	2.82
2015- 12-31	INFY	EQ	1086.30	1090.15	1109.95	1087.0	1107.0	1105.40	1102.09	3971969	4.37

In [31]:

Out[31]:

	Symbol	Series	Prev Close	Open	High	Low	Last	Close	VWAP	Volume	^
Date											
2015- 12-24	TCS	EQ	2425.80	2421.00	2438.2	2404.00	2435.00	2434.25	2424.03	421580	1.C
2015- 12-28	TCS	EQ	2434.25	2428.00	2466.4	2420.20	2456.95	2462.70	2455.00	1852099	4.5
2015- 12-29	TCS	EQ	2462.70	2458.35	2465.3	2445.75	2449.35	2455.80	2452.26	854262	2.0
2015- 12-30	TCS	EQ	2455.80	2453.05	2459.0	2412.30	2421.70	2418.30	2442.06	802881	1.9
2015- 12-31	TCS	EQ	2418.30	2415.75	2448.5	2407.50	2430.00	2436.85	2428.90	620159	1.5

In [32]:

```
def volumeshocks(data):
    data['PreviousVolume']=data['Volume'].shift(1)
    data['VolumeShocks'] = (data['Volume']>data['PreviousVolume']*0.1+data['PreviousVolume']).map({True:0,False:1})
    return data
```

In [33]:

```
nse_infy=volumeshocks(nse_infy)
nse_tcs=volumeshocks(nse_tcs)
```

In [34]:

```
nse_infy[['Volume','PreviousVolume','VolumeShocks']].head()
```

Out[34]:

	Volume	PreviousVolume	VolumeShocks
Date			
2015-01-01	500691	NaN	1
2015-01-02	1694580	500691.0	0
2015-01-05	2484256	1694580.0	0
2015-01-06	2416829	2484256.0	1
2015-01-07	1812479	2416829.0	1

```
nse_tcs[['Volume','PreviousVolume','VolumeShocks']].head()
```

Out[35]:

	Volume	PreviousVolume	VolumeShocks
Date			
2015-01-01	183415	NaN	1
2015-01-02	462870	183415.0	0
2015-01-05	877121	462870.0	0
2015-01-06	1211892	877121.0	0
2015-01-07	1318166	1211892.0	1

In [36]:

```
def priceshocks(data):
    data['T']=data['Close'].shift(1)
    data['PriceShocks'] = (data['Close']-data['T']>0.20*(data['Close']-data['T'])).map({Tru
e:0,False:1})
    return data
```

In [37]:

```
nse_infy=priceshocks(nse_infy)
nse_tcs=priceshocks(nse_tcs)
```

In [38]:

```
nse_infy[['Close','PriceShocks']].head()
```

Out[38]:

	Close	PriceShocks
Date		
2015-01-01	1974.40	1
2015-01-02	2013.20	0
2015-01-05	1995.90	1
2015-01-06	1954.20	1
2015-01-07	1963.55	0

```
nse_infy[['Close','PriceShocks']].head()
```

Out[39]:

	Close	PriceShocks
Date		
2015-01-01	1974.40	1
2015-01-02	2013.20	0
2015-01-05	1995.90	1
2015-01-06	1954.20	1
2015-01-07	1963.55	0

In [40]:

```
def blackswan(data):
    data['T1']=data['Prev Close'].shift(1)
    data['BlackSwanPrice'] = (data['Prev Close']-data['T1']>0.20*(data['Prev Close']-data
['T1'])).map({True:0,False:1})
    return data
```

In [41]:

```
nse_infy=blackswan(nse_infy)
nse_tcs=blackswan(nse_tcs)
```

In [42]:

```
nse_infy[['Prev Close','BlackSwanPrice']].head()
```

Out[42]:

	Prev Close	BlackSwanPrice
Date		
2015-01-01	1972.55	1
2015-01-02	1974.40	0
2015-01-05	2013.20	0
2015-01-06	1995.90	1
2015-01-07	1954.20	1

In [43]:

```
nse_tcs[['Prev Close','BlackSwanPrice']].head()
```

	Prev Close	BlackSwanPrice
Date		
2015-01-01	2558.25	1
2015-01-02	2545.55	1
2015-01-05	2579.45	0
2015-01-06	2540.25	1
2015-01-07	2446.60	1

In [44]:

def priceshocknovolshock(data):
 data['notvolshock'] = (~(data['VolumeShocks'].astype(bool))).astype(int)
 data['PriceshockNovolumeshocks'] = data['notvolshock'] & data['PriceShocks']
 return data

In [45]:

nse_infy=priceshocknovolshock(nse_infy)
nse_tcs=priceshocknovolshock(nse_tcs)

In [46]:

nse_infy[['VolumeShocks','PriceShocks','PriceshockNovolumeshocks']].head()

Out[46]:

	VolumeShocks	PriceShocks	PriceshockNovolumeshocks
Date			
2015-01-01	1	1	0
2015-01-02	0	0	0
2015-01-05	0	1	1
2015-01-06	1	1	0
2015-01-07	1	0	0

In [47]:

Out[47]:

	VolumeShocks	PriceShocks	PriceshockNovolumeshocks
Date			
2015-01-01	1	1	0
2015-01-02	0	0	0
2015-01-05	0	1	1
2015-01-06	0	1	1
2015-01-07	1	1	0

In [48]:

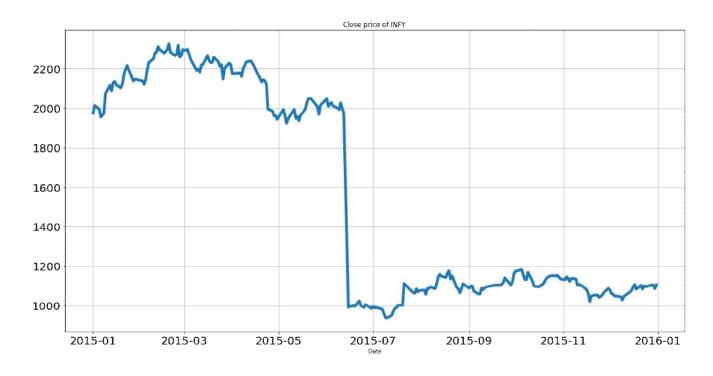
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns

In [49]:

nse_infy.Close.plot(figsize=(20,10),linewidth=5,fontsize=20,grid=True)
plt.title("Close price of INFY")

Out[49]:

Text(0.5,1,'Close price of INFY')

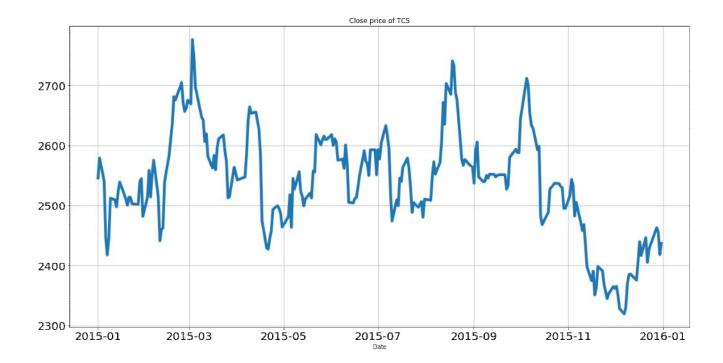


In [50]:

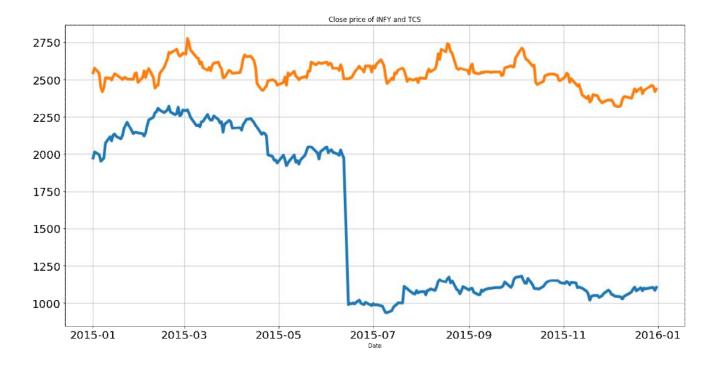
 $\label{lem:nse_tcs.Close.plot(figsize=(20,10),linewidth=5,fontsize=20,grid=\mathbf{True})} \\ \text{plt.title("Close price of TCS")}$

Out[50]:

Text(0.5,1,'Close price of TCS')



In [51]:

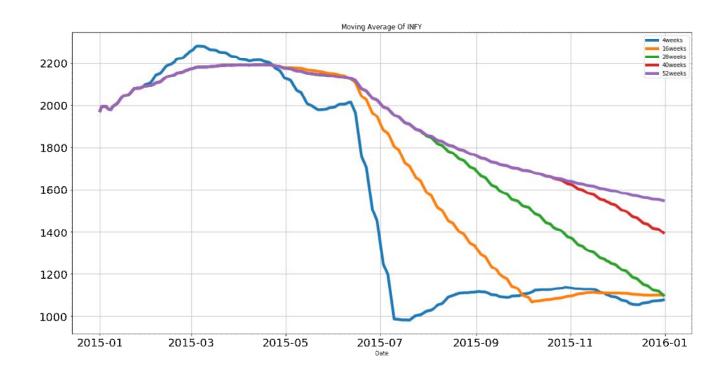


In [52]:

nse_infy[['4weeks','16weeks','28weeks','40weeks','52weeks']].plot(grid=True,figsize=(20,1
0),linewidth=5,fontsize=20)
plt.title("Moving Average Of INFY")

Out[52]:

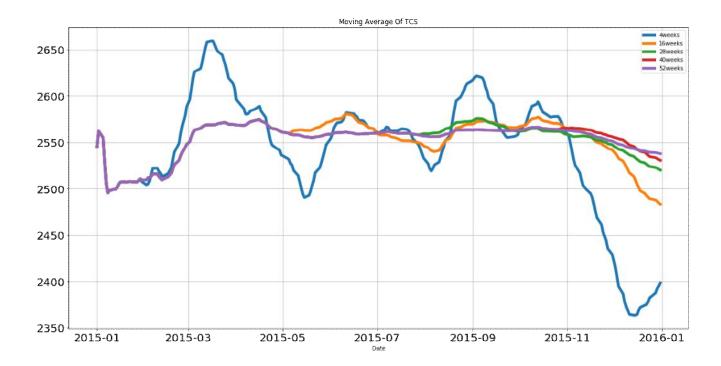
Text(0.5,1,'Moving Average Of INFY')



nse_tcs[['4weeks','16weeks','28weeks','40weeks','52weeks']].plot(grid=True, figsize=(20,10),
linewidth=5, fontsize=20)
plt.title("Moving Average Of TCS")

Out[53]:

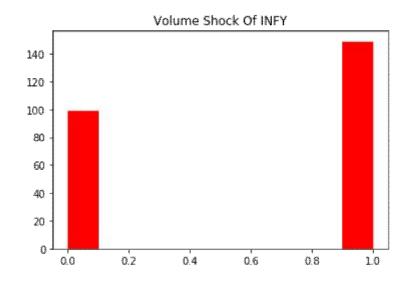
Text(0.5,1,'Moving Average Of TCS')



In [54]:

Out[54]:

Text(0.5,1,'Volume Shock Of INFY')

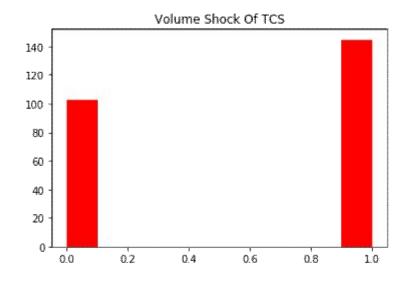


In [55]:

plt.hist(nse_tcs.VolumeShocks,color='red')
plt.title("Volume Shock Of TCS")

Out[55]:

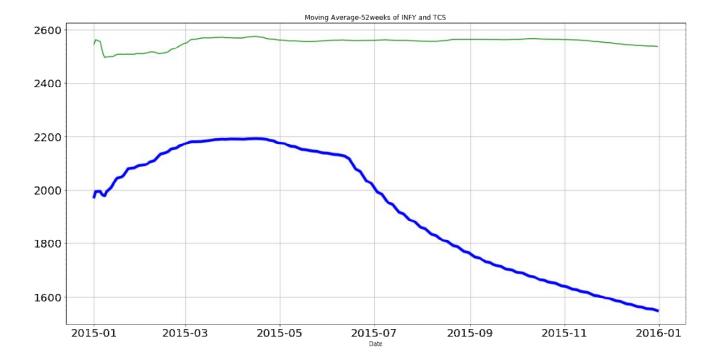
Text(0.5,1,'Volume Shock Of TCS')



In [56]:

Out[56]:

Text(0.5,1,'Moving Average-52weeks of INFY and TCS')

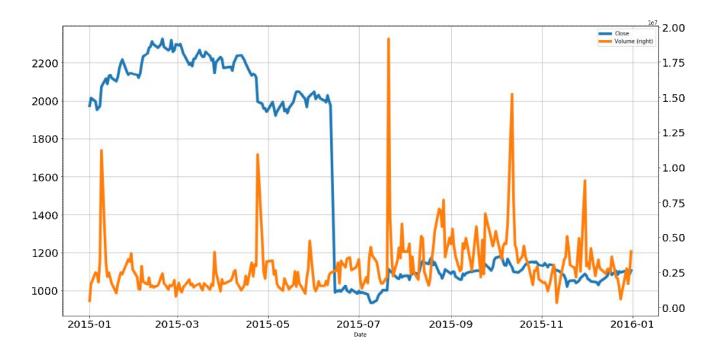


In [57]:

nse_infy[['Close','Volume']].plot(secondary_y=['Volume'],grid=True,figsize=(20,10),linewidt
h=5,fontsize=20)

Out[57]:

<matplotlib.axes._subplots.AxesSubplot at 0x204f2472e10>

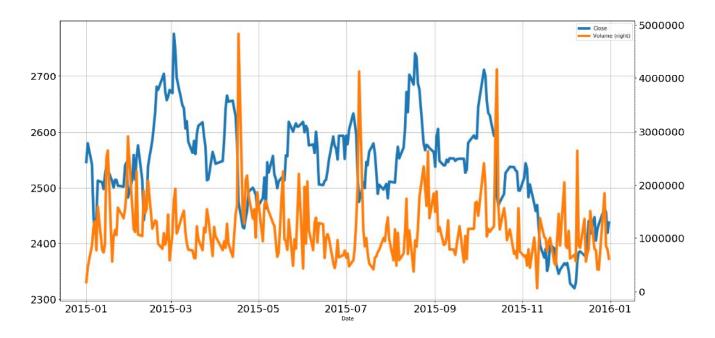


In [58]:

nse_tcs[['Close','Volume']].plot(secondary_y=['Volume'],grid=True,figsize=(20,10),linewidth
=5,fontsize=20)

Out[58]:

<matplotlib.axes._subplots.AxesSubplot at 0x204f3057e80>

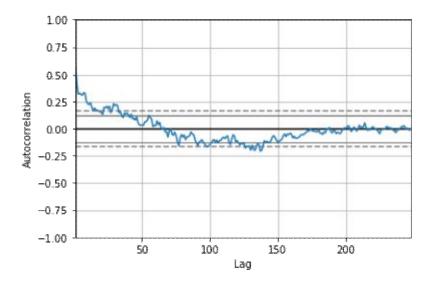


In [59]:

C:\Users\TINU ROHITH\Anaconda3\lib\site-packages\ipykernel_launcher.py:2: FutureWarning: 'p andas.tools.plotting.autocorrelation_plot' is deprecated, import 'pandas.plotting.autocorre lation_plot' instead.

Out[59]:

<matplotlib.axes._subplots.AxesSubplot at 0x204f27b6630>



In [60]:

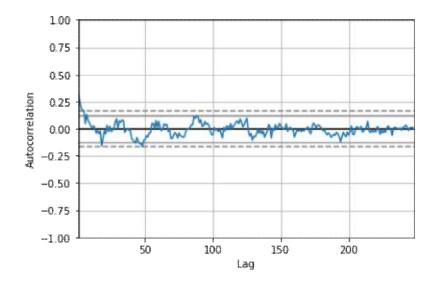
autocorrelation_plot(np.log(nse_tcs['Volume']))

C:\Users\TINU ROHITH\Anaconda3\lib\site-packages\ipykernel_launcher.py:1: FutureWarning: 'p andas.tools.plotting.autocorrelation_plot' is deprecated, import 'pandas.plotting.autocorre lation_plot' instead.

"""Entry point for launching an IPython kernel.

Out[60]:

<matplotlib.axes._subplots.AxesSubplot at 0x204ef8d1c50>



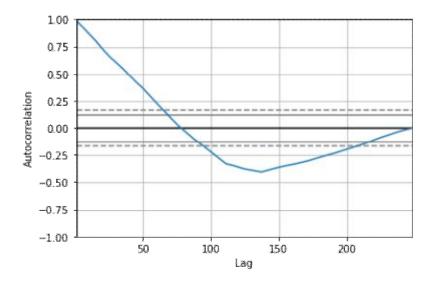
In [61]:

C:\Users\TINU ROHITH\Anaconda3\lib\site-packages\ipykernel_launcher.py:1: FutureWarning: 'p andas.tools.plotting.autocorrelation_plot' is deprecated, import 'pandas.plotting.autocorre lation plot' instead.

"""Entry point for launching an IPython kernel.

Out[61]:

<matplotlib.axes._subplots.AxesSubplot at 0x204f2888d68>



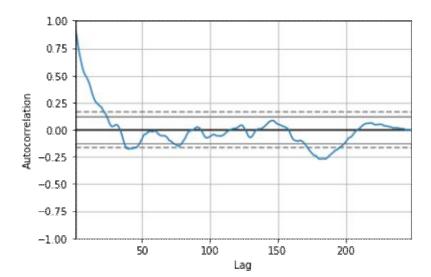
In [62]:

C:\Users\TINU ROHITH\Anaconda3\lib\site-packages\ipykernel_launcher.py:1: FutureWarning: 'p andas.tools.plotting.autocorrelation_plot' is deprecated, import 'pandas.plotting.autocorre lation plot' instead.

"""Entry point for launching an IPython kernel.

Out[62]:

<matplotlib.axes. subplots.AxesSubplot at 0x204f28d5c88>



In [63]:

 $\verb|nse_infy=nse_infy.drop(['T','notvolshock','PreviousVolume','T1'],axis=1).head()|$

In [64]:

nse_tcs=nse_tcs.drop(['T','notvolshock','PreviousVolume','T1'],axis=1).head()

In [65]:

x=nse_infy.drop(['Close','Symbol','Series'],axis=1)
y=nse_infy['Close']

In [66]:

import sklearn.model selection as model selection

x_train,x_test,y_train,y_test=model_selection.train_test_split(x,y,test_size=0.20,random_st
ate=200)

In [67]:

import sklearn.linear_model as linear_model

In [68]:

reg=linear_model.Ridge(normalize=True, fit_intercept=True)
reg=reg.fit(x_train, y_train)

In [69]:

```
{'alpha': 99.1}
                                                                                 In [70]:
reg=linear_model.Ridge(normalize=True, fit_intercept=True, alpha=99.1)
reg=reg.fit(x train,y train)
reg.coef
                                                                                 Out[70]:
array([ 3.95348801e-03, 3.44343167e-03, 6.40849456e-03, 1.36441625e-02,
       8.41724573e-03, 8.69453162e-03, 7.38063198e-08, 3.88842417e-16,
       1.72158077e-06, 9.74845159e-08, 6.91166204e-01, 1.65754154e-02,
       1.65754154e-02, 1.65754154e-02, 1.65754154e-02, 1.65754154e-02,
       -3.20000526e-01, -3.20385487e-02, -3.20000526e-01, 9.29126429e-02])
                                                                                 In [71]:
reg.intercept
                                                                                 Out[71]:
1733.3273414111247
                                                                                 In [72]:
import sklearn.metrics as metrics
import sklearn.preprocessing as preprocessing
                                                                                 In [73]:
x test=preprocessing.normalize(x test)
                                                                                 In [76]:
print('MAE:',metrics.mean squared error(y test,greg.predict(x test)))
MAE: 48784.73131211767
```

In [80]:

```
x1=nse_tcs.drop(['Close','Symbol','Series'],axis=1)
y1=nse_tcs['Close']
x1_train,x1_test,y1_train,y1_test=model_selection.train_test_split(x1,y1,test_size=0.20,ran
dom_state=200)
```

In [81]:

```
reg1=linear_model.Lasso(max_iter=1000, normalize=True)
reg1=reg1.fit(x1_train, y1_train)
```

```
greg1=model selection.GridSearchCV(reg1,param grid={'alpha':np.arange(0.1,100,1).tolist()})
greg1=greg1.fit(x1 train,y1 train)
greg1.best_params_
C:\Users\TINU ROHITH\Anaconda3\lib\site-packages\sklearn\linear model\coordinate descent.p
y:491: ConvergenceWarning: Objective did not converge. You might want to increase the numbe
r of iterations. Fitting data with very small alpha may cause precision problems.
  ConvergenceWarning)
C:\Users\TINU ROHITH\Anaconda3\lib\site-packages\sklearn\linear model\coordinate descent.p
y:491: ConvergenceWarning: Objective did not converge. You might want to increase the numbe
r of iterations. Fitting data with very small alpha may cause precision problems.
  ConvergenceWarning)
C:\Users\TINU ROHITH\Anaconda3\lib\site-packages\sklearn\linear model\coordinate descent.p
y:491: ConvergenceWarning: Objective did not converge. You might want to increase the numbe
r of iterations. Fitting data with very small alpha may cause precision problems.
  ConvergenceWarning)
C:\Users\TINU ROHITH\Anaconda3\lib\site-packages\sklearn\linear model\coordinate descent.p
y:491: ConvergenceWarning: Objective did not converge. You might want to increase the numbe
r of iterations. Fitting data with very small alpha may cause precision problems.
  ConvergenceWarning)
C:\Users\TINU ROHITH\Anaconda3\lib\site-packages\sklearn\linear model\coordinate descent.p
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r of iterations. Fitting data with very small alpha may cause precision problems.
  ConvergenceWarning)
C:\Users\TINU ROHITH\Anaconda3\lib\site-packages\sklearn\linear model\coordinate descent.p
y:491: ConvergenceWarning: Objective did not converge. You might want to increase the numbe
r of iterations. Fitting data with very small alpha may cause precision problems.
  ConvergenceWarning)
C:\Users\TINU ROHITH\Anaconda3\lib\site-packages\sklearn\linear model\coordinate descent.p
y:491: ConvergenceWarning: Objective did not converge. You might want to increase the numbe
r of iterations. Fitting data with very small alpha may cause precision problems.
  ConvergenceWarning)
C:\Users\TINU ROHITH\Anaconda3\lib\site-packages\sklearn\linear model\coordinate descent.p
y:491: ConvergenceWarning: Objective did not converge. You might want to increase the numbe
r of iterations. Fitting data with very small alpha may cause precision problems.
  ConvergenceWarning)
C:\Users\TINU ROHITH\Anaconda3\lib\site-packages\sklearn\linear model\coordinate descent.p
y:491: ConvergenceWarning: Objective did not converge. You might want to increase the numbe
r of iterations. Fitting data with very small alpha may cause precision problems.
  ConvergenceWarning)
                                                                                   Out[82]:
{'alpha': 14.1}
                                                                                   In [83]:
```

reg1=linear_model.Lasso(max_iter=1000,normalize=True,alpha=14.1)

reg1=reg1.fit(x1_train,y1_train)

regl.coef

```
Out[83]:
array([ 0. , 0.
                            , 0.
                                       , 0.0610708 , 0.4857956 ,
       0.00667997, -0.
                            , -0.
                                        , -0. , -0.
                            , 0.
                                        , 0.
                                                      , 0.
      -0. , 0.
       0.
                , -0.
                            , -0.
                                        , -0.
                                                      , 0.
                                                                  ])
                                                                               In [84]:
regl.intercept
                                                                               Out[84]:
1123.7605381386804
                                                                               In [85]:
x1_test=preprocessing.normalize(x1_test)
                                                                               In [86]:
print('MAE:',metrics.mean squared error(y1 test,greg1.predict(x1 test)))
MAE: 1749904.2418575333
                                                                               In [87]:
import sklearn.tree as tree
from sklearn.ensemble import GradientBoostingClassifier
x2=nse infy.drop(['VolumeShocks','Symbol','Series'],axis=1)
x2=pd.get dummies(x2)
y2=nse_infy['VolumeShocks']
x2_train,x2_test,y2_train,y2_test=model_selection.train_test_split(x2,y2,test_size=0.20,ran
dom state=200)
                                                                               In [88]:
clf=GradientBoostingClassifier(n estimators=80,random state=400)
clf.fit(x2 train,y2 train)
                                                                               Out[88]:
GradientBoostingClassifier(criterion='friedman mse', init=None,
             learning rate=0.1, loss='deviance', max depth=3,
             max_features=None, max_leaf_nodes=None,
             min impurity decrease=0.0, min impurity split=None,
             min samples leaf=1, min samples split=2,
             min weight fraction leaf=0.0, n estimators=80,
             presort='auto', random_state=400, subsample=1.0, verbose=0,
             warm start=False)
```

In [89]:

```
clf.score(x2_test,y2_test)
```

1.0

In [90]:

Out[89]:

```
clf.feature_importances_
```

Out[90]:

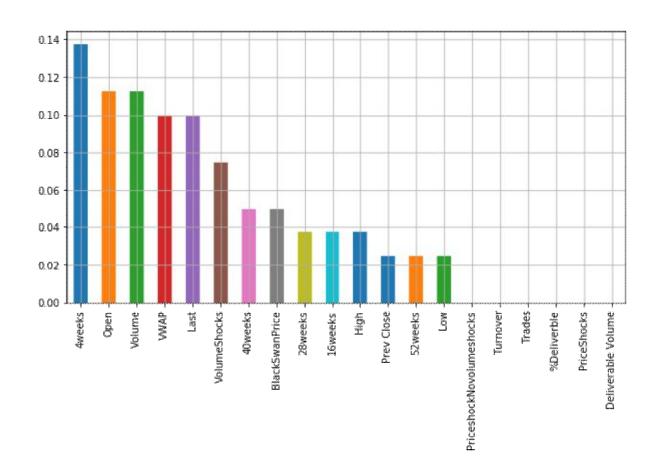
```
array([0.025 , 0.1125, 0.0375, 0.025 , 0.1 , 0.1 , 0.1125, 0. , 0. , 0. , 0. , 0.1375, 0.0375, 0.0375, 0.05 , 0.025 , 0.075 , 0. , 0.05 , 0. ])
```

In [91]:

```
feature=pd.Series(clf.feature_importances_,index=x.columns)
feature.sort_values(ascending=False)
feature.sort_values(ascending=False).plot(kind='bar',figsize=(10,5),grid=True)
```

Out[91]:

<matplotlib.axes._subplots.AxesSubplot at 0x204f5762a20>



```
In [92]:
```

```
x3=nse tcs.drop(['VolumeShocks','Symbol','Series'],axis=1)
x3=pd.get dummies(x3)
y3=nse_tcs['VolumeShocks']
x3_train,x3_test,y3_train,y3_test=model_selection.train_test_split(x3,y3,test_size=0.20,ran
dom state=200)
                                                                                In [93]:
clf1=GradientBoostingClassifier(n estimators=80,random state=400)
clf1.fit(x3_train,y3_train)
                                                                               Out[93]:
GradientBoostingClassifier(criterion='friedman mse', init=None,
             learning rate=0.1, loss='deviance', max depth=3,
             max features=None, max leaf nodes=None,
             min impurity decrease=0.0, min impurity split=None,
             min_samples_leaf=1, min_samples_split=2,
             min weight fraction leaf=0.0, n estimators=80,
             presort='auto', random state=400, subsample=1.0, verbose=0,
             warm start=False)
                                                                               In [94]:
clf1.score(x3_test,y3_test)
                                                                                Out[94]:
0.0
                                                                                In [95]:
clf1.feature_importances_
                                                                               Out[95]:
array([0. , 0. , 0.125 , 0. , 0. , 0. , 0.2125, 0.
      0. , 0. , 0. , 0. , 0.075 , 0.125 , 0.1125, 0.05 ,
      0.225 , 0. , 0. , 0.
                                  ])
                                                                                In [98]:
                                                                               Out[98]:
```

<matplotlib.axes. subplots.AxesSubplot at 0x204f59cc7f0>

