```
# for importing data set i am using yfinance library
In [1]:
          import yfinance as yf
          bitcoin=yf.Ticker("BTC-USD") #Ticker is symbol for stocks
In [2]:
          bitcoin=bitcoin.history(period="max")
In [3]:
          bitcoin
In [4]:
Out[4]:
                                                                                                          Sto
                                                                                      Volume Dividends
                                 Open
                                               High
                                                              Low
                                                                          Close
                                                                                                          Spl
                   Date
             2014-09-17
                            465.864014
                                          468.174011
                                                       452.421997
                                                                     457.334015
                                                                                    21056800
                                                                                                     0.0
          00:00:00+00:00
             2014-09-18
                            456.859985
                                          456.859985
                                                                     424.440002
                                                                                    34483200
                                                                                                     0.0
                                                       413.104004
          00:00:00+00:00
             2014-09-19
                            424.102997
                                          427.834991
                                                        384.532013
                                                                      394.795990
                                                                                    37919700
                                                                                                     0.0
          00:00:00+00:00
             2014-09-20
                            394.673004
                                                       389.882996
                                                                     408.903992
                                                                                                     0.0
                                          423.295990
                                                                                    36863600
          00:00:00+00:00
             2014-09-21
                            408.084991
                                          412.425995
                                                       393.181000
                                                                      398.821014
                                                                                    26580100
                                                                                                     0.0
          00:00:00+00:00
             2023-07-12
                          30622.246094
                                       30959.964844
                                                     30228.835938
                                                                   30391.646484
                                                                                 14805659717
                                                                                                     0.0
                                                                                                            (
         00:00:00+00:00
             2023-07-13
                          30387.488281
                                                     30268.351562
                                       31814.515625
                                                                   31476.048828
                                                                                 23686079548
                                                                                                     0.0
          00:00:00+00:00
             2023-07-14
                          31474.720703 31582.253906
                                                     29966.386719
                                                                   30334.068359
                                                                                                     0.0
                                                                                 20917902660
                                                                                                            (
         00:00:00+00:00
             2023-07-15
                          30331.783203 30407.781250 30263.462891
                                                                   30295.806641
                                                                                  8011667756
                                                                                                     0.0
          00:00:00+00:00
             2023-07-16
                          30304.167969 30336.707031 30122.935547 30208.246094
                                                                                                     0.0
                                                                                                            (
                                                                                  8101786624
         00:00:00+00:00
         3225 rows × 7 columns
```

In [5]: bitcoin.index

```
Bitcoin_prediction_final
         DatetimeIndex(['2014-09-17 00:00:00+00:00', '2014-09-18 00:00:00+00:00',
Out[5]:
                          '2014-09-19 00:00:00+00:00', '2014-09-20 00:00:00+00:00',
                          '2014-09-21 00:00:00+00:00', '2014-09-22 00:00:00+00:00',
                          '2014-09-23 00:00:00+00:00', '2014-09-24 00:00:00+00:00',
                          '2014-09-25 00:00:00+00:00', '2014-09-26 00:00:00+00:00',
                          '2023-07-07 00:00:00+00:00', '2023-07-08 00:00:00+00:00',
                          '2023-07-09 00:00:00+00:00', '2023-07-10 00:00:00+00:00'
                         '2023-07-11 00:00:00+00:00', '2023-07-12 00:00:00+00:00', '2023-07-13 00:00:00+00:00', '2023-07-14 00:00:00+00:00',
                          '2023-07-15 00:00:00+00:00', '2023-07-16 00:00:00+00:00'],
                        dtype='datetime64[ns, UTC]', name='Date', length=3225, freq=None)
         bitcoin.plot.line(y="Close", use_index=True, color={"Close": "Green"})
In [6]:
         <AxesSubplot:xlabel='Date'>
Out[6]:
          70000
                          Close
          60000
          50000
          40000
          30000
```

```
Date
        #deleting unwanted columns
In [7]:
        del bitcoin["Dividends"]
        del bitcoin["Stock Splits"]
        # adding nextdays closing value to make it easy for solving
In [8]:
        bitcoin["Tomorrow"]=bitcoin["Close"].shift(-1) #Here we shifted one value up
        bitcoin
In [9]:
```

2019

2020

2021

2018

2022

2023

2017

20000

10000

0

2015

2016

Out[9]:

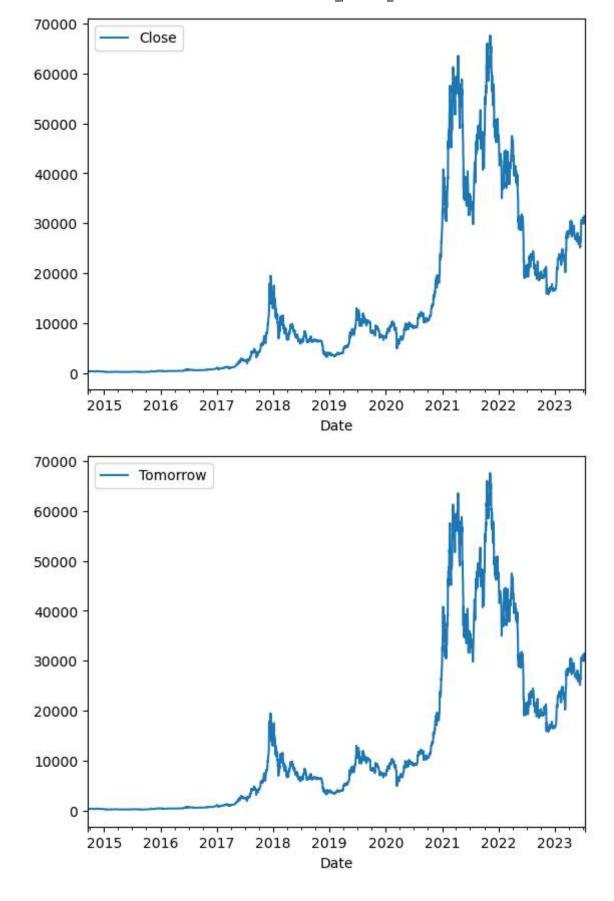
		Open	High	Low	Close	Volume	Tomorrow
	Date						
	2014-09-17 00:00:00+00:00	465.864014	468.174011	452.421997	457.334015	21056800	424.440002
	2014-09-18 00:00:00+00:00	456.859985	456.859985	413.104004	424.440002	34483200	394.795990
	2014-09-19 00:00:00+00:00	424.102997	427.834991	384.532013	394.795990	37919700	408.903992
	2014-09-20 00:00:00+00:00	394.673004	423.295990	389.882996	408.903992	36863600	398.821014
	2014-09-21 00:00:00+00:00	408.084991	412.425995	393.181000	398.821014	26580100	402.152008
	•••						
	2023-07-12 00:00:00+00:00	30622.246094	30959.964844	30228.835938	30391.646484	14805659717	31476.048828
	2023-07-13 00:00:00+00:00	30387.488281	31814.515625	30268.351562	31476.048828	23686079548	30334.068359
	2023-07-14 00:00:00+00:00	31474.720703	31582.253906	29966.386719	30334.068359	20917902660	30295.806641
	2023-07-15 00:00:00+00:00	30331.783203	30407.781250	30263.462891	30295.806641	8011667756	30208.246094
	2023-07-16 00:00:00+00:00	30304.167969	30336.707031	30122.935547	30208.246094	8101786624	NaN

3225 rows × 6 columns

```
In [10]: #creating one more column namely target here if our tomorrows value is grater
    #than closing value it will give output as 1 otherwise 0
    # for this we will use astype(int) to get valuse
    bitcoin["Target"]=(bitcoin["Tomorrow"]>bitcoin["Close"]).astype(int)

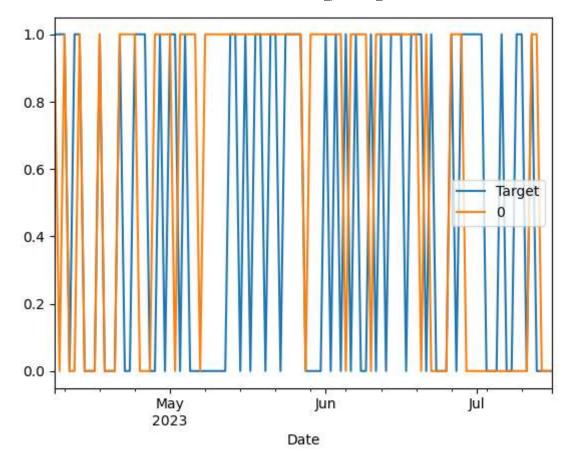
In [11]: bitcoin.plot.line(y="Close",use_index=True)
    bitcoin.plot.line(y="Tomorrow",use_index=True)

Out[11]: <AxesSubplot:xlabel='Date'>
```



using random forest classifier

```
from sklearn.ensemble import RandomForestClassifier
In [12]:
         model=RandomForestClassifier(n estimators=800,min samples split=2,random state=1)
         train=bitcoin.iloc[:-100]
         test= bitcoin.iloc[-100:]
         predictors = ["Close", "Volume", "High", "Low"]
         model.fit(train[predictors],train["Target"])
        RandomForestClassifier(n_estimators=800, random_state=1)
Out[12]:
In [13]: from sklearn.metrics import precision score
         preds = model.predict(test[predictors])
         preds
In [14]:
         array([1, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 1, 1,
Out[14]:
               1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1,
               1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0,
               0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0])
        import pandas as pd
In [15]:
         preds = pd.Series(preds,index=test.index)
         precision_score(test["Target"],preds)
In [16]:
        0.5161290322580645
Out[16]:
         combined=pd.concat([test["Target"],preds],axis=1)
In [17]:
         combined.plot()
In [19]:
         <AxesSubplot:xlabel='Date'>
Out[19]:
```



```
# backtesting
In [20]:
          def predict(train, test, predictors, model):
              model.fit(train[predictors],train["Target"])
              preds=model.predict(test[predictors])
              preds=pd.Series(preds,index=test.index,name="Predictions")
              combined=pd.concat([test["Target"],preds],axis=1)
              return combined
In [21]:
          def backtest(data,model,predictors,start=2500, step=250):
              all_prediction=[]
              for i in range(start,data.shape[0],step):
                  train=data.iloc[0:i].copy()
                  test=data.iloc[i:(i+step)].copy()
                  predictions=predict(train,test,predictors,model)
                  all_prediction.append(predictions)
              return pd.concat(all prediction)
          predictions=backtest(bitcoin,model,predictors)
In [22]:
          predictions["Predictions"].value_counts()
In [23]:
              423
Out[23]:
               302
         Name: Predictions, dtype: int64
In [24]:
          precision_score(predictions["Target"],predictions["Predictions"])
```