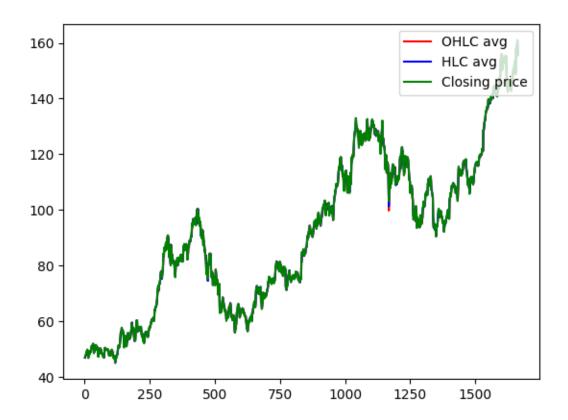
## 01. stock prediction. rev. 01

## August 14, 2024

```
[]: import pandas as pd
     import matplotlib.pyplot as plt
     import numpy as np
     import math
     from sklearn.preprocessing import MinMaxScaler
     from sklearn.metrics import mean squared error
     from keras.models import Sequential
     from keras.layers import Dense, Activation
     from keras.layers import LSTM
     import preprocessing
[]: np.random.seed(7)
[]: dataset = pd.read_csv('apple_share_price.csv', usecols=[1,2,3,4])
     dataset = dataset.reindex(index = dataset.index[::-1])
[]: obs = np.arange(1, len(dataset) + 1, 1)
     OHLC_avg = dataset.mean(axis = 1)
     HLC_avg = dataset[['High', 'Low', 'Close']].mean(axis = 1)
     close_val = dataset[['Close']]
[]: plt.plot(obs, OHLC_avg, 'r', label = 'OHLC avg')
    plt.plot(obs, HLC_avg, 'b', label = 'HLC avg')
     plt.plot(obs, close_val, 'g', label = 'Closing price')
     plt.legend(loc = 'upper right')
     plt.show()
```



```
[]: OHLC_avg = np.reshape(OHLC_avg.values, (len(OHLC_avg),1)) # 1664
     scaler = MinMaxScaler(feature_range=(0, 1))
     OHLC_avg = scaler.fit_transform(OHLC_avg)
[]: train_OHLC = int(len(OHLC_avg) * 0.75)
     test_OHLC = len(OHLC_avg) - train_OHLC
     train_OHLC, test_OHLC = OHLC_avg[0:train_OHLC,:], OHLC_avg[train_OHLC:
      →len(OHLC_avg),:]
[]: trainX, trainY = preprocessing.new_dataset(train_OHLC, 1)
     testX, testY = preprocessing.new_dataset(test_OHLC, 1)
[]: trainX = np.reshape(trainX, (trainX.shape[0], 1, trainX.shape[1]))
     testX = np.reshape(testX, (testX.shape[0], 1, testX.shape[1]))
     step\_size = 1
[]: model = Sequential()
     model.add(LSTM(32, input_shape=(1, step_size), return_sequences = True))
     model.add(LSTM(16))
     model.add(Dense(1))
     model.add(Activation('linear'))
```

```
[]: model.compile(loss='mean_squared_error', optimizer='adagrad') # Try SGD, adam, __
      ⇒adagrad and compare!!!
     model.fit(trainX, trainY, epochs=20, batch_size=1, verbose=1)
    Epoch 1/20
    1246/1246
                           5s 2ms/step -
    loss: 0.1274
    Epoch 2/20
    1246/1246
                           2s 2ms/step -
    loss: 0.0821
    Epoch 3/20
    1246/1246
                           2s 2ms/step -
    loss: 0.0677
    Epoch 4/20
    1246/1246
                           2s 2ms/step -
    loss: 0.0593
    Epoch 5/20
                           2s 2ms/step -
    1246/1246
    loss: 0.0520
    Epoch 6/20
    1246/1246
                           2s 2ms/step -
    loss: 0.0505
    Epoch 7/20
    1246/1246
                           2s 2ms/step -
    loss: 0.0422
    Epoch 8/20
    1246/1246
                           2s 1ms/step -
    loss: 0.0450
    Epoch 9/20
    1246/1246
                           2s 2ms/step -
    loss: 0.0403
    Epoch 10/20
                           3s 2ms/step -
    1246/1246
    loss: 0.0413
    Epoch 11/20
    1246/1246
                           3s 2ms/step -
    loss: 0.0381
    Epoch 12/20
    1246/1246
                           2s 2ms/step -
    loss: 0.0369
    Epoch 13/20
    1246/1246
                           3s 2ms/step -
    loss: 0.0402
    Epoch 14/20
    1246/1246
                           2s 2ms/step -
    loss: 0.0389
    Epoch 15/20
```

2s 2ms/step -

1246/1246

```
loss: 0.0382
    Epoch 16/20
    1246/1246
                          3s 2ms/step -
    loss: 0.0370
    Epoch 17/20
    1246/1246
                          3s 2ms/step -
    loss: 0.0356
    Epoch 18/20
    1246/1246
                          2s 2ms/step -
    loss: 0.0356
    Epoch 19/20
    1246/1246
                          2s 2ms/step -
    loss: 0.0352
    Epoch 20/20
                          2s 2ms/step -
    1246/1246
    loss: 0.0354
[]: <keras.src.callbacks.history.History at 0x1fb8774e2a0>
[]: trainPredict = model.predict(trainX)
     testPredict = model.predict(testX)
     trainPredict = scaler.inverse_transform(trainPredict)
     trainY = scaler.inverse_transform([trainY])
     testPredict = scaler.inverse_transform(testPredict)
     testY = scaler.inverse_transform([testY])
    39/39
                      1s 8ms/step
    13/13
                      Os 5ms/step
Г1:
     trainScore = math.sqrt(mean_squared_error(trainY[0], trainPredict[:,0]))
     print('Train RMSE: %.2f' % (trainScore))
     testScore = math.sqrt(mean_squared_error(testY[0], testPredict[:,0]))
     print('Test RMSE: %.2f' % (testScore))
     trainPredictPlot = np.empty_like(OHLC_avg)
     trainPredictPlot[:, :] = np.nan
     trainPredictPlot[step_size:len(trainPredict)+step_size, :] = trainPredict
     testPredictPlot = np.empty_like(OHLC_avg)
     testPredictPlot[:, :] = np.nan
     testPredictPlot[len(trainPredict)+(step_size*2)+1:len(OHLC_avg)-1, :] = __
      →testPredict
     OHLC_avg = scaler.inverse_transform(OHLC_avg)
```

Train RMSE: 21.69

Test RMSE: 35.35

```
[]: import numpy

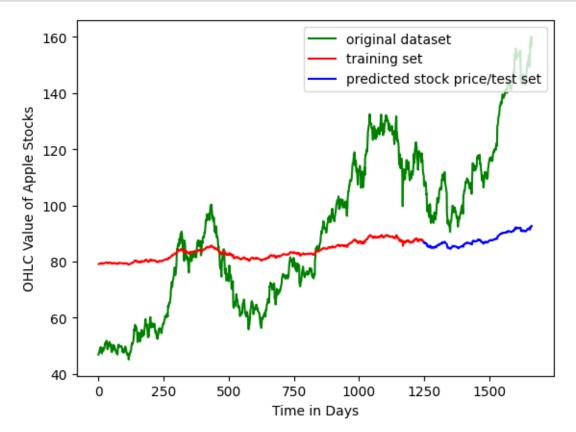
def patch_asscalar(a):
    return a.item()

setattr(numpy, "asscalar", patch_asscalar)

[]: plt.plot(OHLC avg, 'g', label = 'original dataset')
```

```
[]: plt.plot(OHLC_avg, 'g', label = 'original dataset')
   plt.plot(trainPredictPlot, 'r', label = 'training set')
   plt.plot(testPredictPlot, 'b', label = 'predicted stock price/test set')
   plt.legend(loc = 'upper right')
   plt.xlabel('Time in Days')
   plt.ylabel('OHLC Value of Apple Stocks')
   plt.show()

last_val = testPredict[-1]
   last_val_scaled = last_val/last_val
   next_val = model.predict(np.reshape(last_val_scaled, (1,1,1)))
   print("Last Day Value:", np.asscalar(last_val))
   print("Next Day Value:", np.asscalar(last_val*next_val))
```



1/1 0s 29ms/step

Last Day Value: 92.63001251220703 Next Day Value: 38.290130615234375

1/1 0s 29ms/step

Last Day Value: 92.63001251220703 Next Day Value: 38.290130615234375