RNN (Google Stock) Name: Atharva Ramgirkar **Registration Number:** 19BCE0114 Submission Date: 18 July, 2021 **Program:** VIT-Al Industry Certifiation **Email:** atharva.ramgirkar2019@vitstudent.ac.in Other Assignments can be found in the link: https://drive.google.com/drive/folders/1QGOLHyZykoj\_CroTJu6-YkZWf32JZ-QH?usp=sharing **Table of Content**  Importing Libraries Initailizing Objects Getting Data for Training Getting Target Variable Scaling the Data Making the Training Dataset Reshaping the Data to fit LSTM input • Building the Model Adding Hidden Layers Adding Output Layer Compiling the Model Training the Model Getting the Test Data Concatinating Test Data to match LSTM requirements Reshaping Test Data to Match model Input Format Predicting Using Trained Model Compairing Predictions with Real Values 1. Importing Libraries Back To Top import pandas as pd import numpy as np # For Plotting import matplotlib.pyplot as plt %matplotlib inline # For the Neural Network from tensorflow.keras.models import Sequential from tensorflow.keras.layers import Dense from tensorflow.keras.layers import LSTM # Scaling the Data from sklearn.preprocessing import MinMaxScaler 1.1 Initializing Objects # Initializing the MinMaxScalar Object = MinMaxScaler() # Initializing the Model model = Sequential() 2. Getting Data for Training Back To Top df\_train = pd.read\_csv("Google\_Stock\_Price\_Train.csv", parse\_dates=['Date']) In [4]: df\_train Out[4]: Date Open High Low Close Volume **0** 2012-01-03 325.25 332.83 324.97 663.59 73,80,500 **1** 2012-01-04 331.27 333.87 329.08 666.45 57,49,400 **2** 2012-01-05 329.83 330.75 326.89 657.21 65,90,300 **3** 2012-01-06 328.34 328.77 323.68 648.24 **4** 2012-01-09 322.04 322.29 309.46 620.76 1,16,88,800 **1253** 2016-12-23 790.90 792.74 787.28 789.91 6,23,400 **1254** 2016-12-27 790.68 797.86 787.66 791.55 7,89,100 **1255** 2016-12-28 793.70 794.23 783.20 785.05 11,53,800 **1256** 2016-12-29 783.33 785.93 778.92 782.79 7,44,300 **1257** 2016-12-30 782.75 782.78 770.41 771.82 17,70,000 1258 rows × 6 columns df train Open = df train[['Date','Open']] df train Open Date Open **0** 2012-01-03 325.25 **1** 2012-01-04 331.27 2 2012-01-05 329.83 **3** 2012-01-06 328.34 **4** 2012-01-09 322.04 **1253** 2016-12-23 790.90 **1254** 2016-12-27 790.68 **1255** 2016-12-28 793.70 **1256** 2016-12-29 783.33 **1257** 2016-12-30 782.75 1258 rows × 2 columns df\_vals = df\_train\_Open.reset\_index()['Open'] 2.1 Getting Target Variable df vals 325.25 0 331.27 329.83 328.34 322.04 790.90 1253 1254 790.68 1255 793.70 1256 783.33 782.75 1257 Name: Open, Length: 1258, dtype: float64 df\_vals.shape Out[9]: (1258,) 3. Scaling the Data **Back To Top** df\_vals=sc.fit\_transform(np.array(df\_vals).reshape(-1,1)) df vals Out[11]: array([[0.08581368], [0.09701243], [0.09433366], [0.95725128],[0.93796041], [0.93688146]]) df vals.shape[0]-60 Out[18]: 1198 4. Making the Training Dataset **Back To Top** X, y = [],[]for i in range(len(df\_vals)-60):  $a = df_{vals}[i:(i+60), 0]$ X.append(a)  $y.append(df_vals[i + 60, 0])$ X train = np.array(X) X train.shape Out[28]: (1198, 60) y train = np.array(y) y\_train.shape Out[29]: (1198,) Reshaping the Data to fit LSTM input **Back To Top** X\_train =X\_train.reshape(X\_train.shape[0],X\_train.shape[1] , 1) X train.shape Out[31]: (1198, 60, 1) 6. Building the Model **Back To Top** 6.1 Adding Hidden Layers model.add(LSTM(60,return sequences=True,input shape=(60,1))) model.add(LSTM(60, return sequences=True)) model.add(LSTM(30)) 6.2 Adding Output Layer model.add(Dense(1)) 7. Compiling the Model **Back To Top** model.compile(loss='mean squared error', optimizer='adam') 8. Training the Model Back To Top In [40]: model.fit(X train, y\_train, epochs=5, batch size=10) Epoch 1/5 120/120 [============== ] - 11s 91ms/step - loss: 0.0110 Epoch 2/5 120/120 [============== ] - 11s 92ms/step - loss: 0.0017 Epoch 3/5 120/120 [============= ] - 12s 98ms/step - loss: 0.0017 Epoch 4/5 120/120 [============== ] - 11s 94ms/step - loss: 0.0014 Epoch 5/5 120/120 [============= ] - 10s 84ms/step - loss: 0.0013 Out[40]: <tensorflow.python.keras.callbacks.History at 0x2139085b580> 9. Getting the Test Data Back To Top In [41]: df\_test = pd.read\_csv("Google\_Stock\_Price\_Test.csv",parse\_dates=['Date']) In [42]: df test Out[42]: Date Open High Low Close Volume **0** 2017-01-03 778.81 789.63 775.80 786.14 1,657,300 **1** 2017-01-04 788.36 791.34 783.16 786.90 **2** 2017-01-05 786.08 794.48 785.02 794.02 **3** 2017-01-06 795.26 807.90 792.20 806.15 1,640,200 **4** 2017-01-09 806.40 809.97 802.83 806.65 **5** 2017-01-10 807.86 809.13 803.51 804.79 **6** 2017-01-11 805.00 808.15 801.37 807.91 **7** 2017-01-12 807.14 807.39 799.17 806.36 1,353,100 **8** 2017-01-13 807.48 811.22 806.69 807.88 **9** 2017-01-17 807.08 807.14 800.37 804.61 **10** 2017-01-18 805.81 806.21 800.99 806.07 2017-01-19 805.12 809.48 801.80 802.17 919,300 **12** 2017-01-20 806.91 806.91 801.69 805.02 **13** 2017-01-23 807.25 820.87 803.74 819.31 **14** 2017-01-24 822.30 825.90 817.82 823.87 **15** 2017-01-25 829.62 835.77 825.06 835.67 1,494,500 **16** 2017-01-26 837.81 838.00 827.01 832.15 **17** 2017-01-27 834.71 841.95 820.44 823.31 2017-01-30 814.66 815.84 799.80 802.32 2017-01-31 796.86 801.25 790.52 796.79 2,160,600 9.1 Concatinating Test Data to match LSTM requirements

In [43]:

In [44]:

In [45]:

In [46]:

In [47]:

In [48]:

In [49]:

Out[44]: (70, 6)

df test.shape

X, y = [],[]

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830

820

810

plt.legend()

df\_test = pd.concat([df\_train[1208:],

df\_vals = df\_test.reset\_index()['Open']

for i in range(len(df vals)-60):

X.append(a)

 $X_{test} = np.array(X)$ 

 $a = df \ vals[i:(i+60), 0]$ 

10. Predicting Using Trained Model

preds = model.predict(X test)

real = list(df test['Open'][60:])

plt.figure(figsize=(9,5))

Out[59]: <matplotlib.legend.Legend at 0x2139ab264f0>

y.append(df vals[i + 60, 0])

df test],

9.2 Reshaping Test Data to Match model Input Format

 $\label{lem:continuous} $$ df_vals=sc.fit_transform(np.array(df_vals).reshape(-1,1))$$ 

X test = X test.reshape(X test.shape[0], X test.shape[1] , 1)

preds=list(sc.inverse transform(preds).reshape(1,-1)[0])

11. Compairing Predictions with Real Values

plt.plot(df\_test.iloc[60:,0],real,label='real')

plt.plot(df test.iloc[60:,0],preds,label='predictions')

2017-01-19 2017-01-21 2017-01-23 2017-01-25 2017-01-27 2017-01-29 2017-01-31

predictions

axis=0).reset\_index(drop=True)

**Assignment 9**