

# LSTM Without Google Trend and Financial Indicators

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```
[1]: import pandas as pd
import numpy as np
import datetime
import pytz #function of time region
import statsmodels.api as sm # Unit root test
import matplotlib.pyplot as plt
import seaborn as sns
import matplotlib.dates as mdate

from keras.layers import Dropout
from statsmodels.tsa.arima.model import ARIMA #ARIMA model
from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
from datetime import datetime
from pmdarima.arima import auto_arima
from math import sqrt
from sklearn.metrics import mean_squared_error

[3]: bitcoin = pd.read_csv(r"E:\PhD study\ELEG5491 Introduction to Deep_
↳Learning\bitcoin\datasets\bitcoin1dimtrain.csv")
training_set1=bitcoin.values #converting to 2d array
training_set1

[3]: array([[ 382.845],
[ 386.475],
[ 383.158],
...,
[7152.302],
[6932.48 ],
[6640.515]])

[4]: from sklearn.preprocessing import MinMaxScaler
sc = MinMaxScaler() #scaling using normalisation
training_set1 = sc.fit_transform(training_set1)
xtrain=training_set1[0:1884] #input values of rows_
↳[0-2694]
ytrain=training_set1[1:1885]
```

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[5]: today=pd.DataFrame(xtrain)           #assigning the values of xtrain to
      ↪ today
      tomorrow=pd.DataFrame(ytrain)       #assigning the values of xtrain to
      ↪ tomorrow
      ex= pd.concat([today,tomorrow],axis=1)   #concat two columns
      ex.columns=(['today','tomorrow'])
      xtrain = np.reshape(xtrain, (1884, 1, 1))
```

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[6]: from keras.models import Sequential
      from keras.layers import Dense
      from keras.layers import LSTM
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[7]: regressor=Sequential()
      ↪ #initialize the RNN
      regressor.add(LSTM(units=4,activation='sigmoid',input_shape=(None,1)))
      regressor.add(Dropout(0.01))           #adding input layerand the LSTM layer
      regressor.add(Dense(units=1))
      ↪ #adding output layers
      regressor.compile(optimizer='adam',loss='mean_squared_error')
      ↪ #compiling the RNN
      regressor.fit(xtrain,ytrain,batch_size=25,epochs=30)
```

```
Epoch 1/30
76/76 [=====] - 0s 682us/step - loss: 0.0843
Epoch 2/30
76/76 [=====] - 0s 656us/step - loss: 0.0427
Epoch 3/30
76/76 [=====] - 0s 682us/step - loss: 0.0358
Epoch 4/30
76/76 [=====] - 0s 696us/step - loss: 0.0327
Epoch 5/30
76/76 [=====] - 0s 682us/step - loss: 0.0313
Epoch 6/30
76/76 [=====] - 0s 656us/step - loss: 0.0298
Epoch 7/30
76/76 [=====] - 0s 630us/step - loss: 0.0285
Epoch 8/30
76/76 [=====] - 0s 654us/step - loss: 0.0269
Epoch 9/30
76/76 [=====] - 0s 656us/step - loss: 0.0254
Epoch 10/30
76/76 [=====] - 0s 669us/step - loss: 0.0244
Epoch 11/30
76/76 [=====] - 0s 706us/step - loss: 0.0218
Epoch 12/30
76/76 [=====] - 0s 669us/step - loss: 0.0212
Epoch 13/30
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76/76 [=====] - 0s 720us/step - loss: 0.0195
Epoch 14/30
76/76 [=====] - 0s 709us/step - loss: 0.0188
Epoch 15/30
76/76 [=====] - 0s 643us/step - loss: 0.0171
Epoch 16/30
76/76 [=====] - 0s 643us/step - loss: 0.0153
Epoch 17/30
76/76 [=====] - 0s 653us/step - loss: 0.0145
Epoch 18/30
76/76 [=====] - 0s 722us/step - loss: 0.0132
Epoch 19/30
76/76 [=====] - 0s 782us/step - loss: 0.0115
Epoch 20/30
76/76 [=====] - 0s 718us/step - loss: 0.0107
Epoch 21/30
76/76 [=====] - 0s 774us/step - loss: 0.0098
Epoch 22/30
76/76 [=====] - 0s 722us/step - loss: 0.0088
Epoch 23/30
76/76 [=====] - 0s 709us/step - loss: 0.0078
Epoch 24/30
76/76 [=====] - 0s 643us/step - loss: 0.0068
Epoch 25/30
76/76 [=====] - 0s 682us/step - loss: 0.0063
Epoch 26/30
76/76 [=====] - 0s 577us/step - loss: 0.0075
Epoch 27/30
76/76 [=====] - 0s 630us/step - loss: 0.0068
Epoch 28/30
76/76 [=====] - 0s 617us/step - loss: 0.0041
Epoch 29/30
76/76 [=====] - 0s 643us/step - loss: 0.0048
Epoch 30/30
76/76 [=====] - 0s 643us/step - loss: 0.0059

```

[7]: <tensorflow.python.keras.callbacks.History at 0x1e42fd10fd0>

```

[9]: #Get the test set
test_set = pd.read_csv(r"E:\PhD study\ELEG5491 Introduction to Deep_
↳ Learning\bitcoin\datasets\bitcoindimtest.csv")
test_set.head()

```

```

[9]:      Close
0  7276.803
1  7202.844
2  7218.816

```

```
3 7191.159
4 7511.589
```

```
[13]: #Get the prediction result
inputs = test_set.values          #converting to 2D array
inputs = sc.fit_transform(inputs)
inputs = np.reshape(inputs, (471, 1, 1))
predicted_price = regressor.predict(inputs)
predicted_price = sc.inverse_transform(predicted_price)
testScore = sqrt(mean_squared_error(predicted_price, test_set))
print('Test Score: %.2f RMSE' % (testScore))
```

Test Score: 3094.80 RMSE

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[ ]:
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