

# Initial Results and Code

March 12, 2021

```
[1]: pip install pandas
```

```
Requirement already satisfied: pandas in /opt/conda/lib/python3.7/site-packages (1.0.3)
Requirement already satisfied: pytz>=2017.2 in /opt/conda/lib/python3.7/site-packages (from pandas) (2020.1)
Requirement already satisfied: numpy>=1.13.3 in /opt/conda/lib/python3.7/site-packages (from pandas) (1.19.5)
Requirement already satisfied: python-dateutil>=2.6.1 in /opt/conda/lib/python3.7/site-packages (from pandas) (2.8.1)
Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.7/site-packages (from python-dateutil>=2.6.1->pandas) (1.15.0)
Note: you may need to restart the kernel to use updated packages.
```

```
[2]: pip install sklearn
```

```
Requirement already satisfied: sklearn in /opt/conda/lib/python3.7/site-packages (0.0)
Requirement already satisfied: scikit-learn in /opt/conda/lib/python3.7/site-packages (from sklearn) (0.22.2.post1)
Requirement already satisfied: joblib>=0.11 in /opt/conda/lib/python3.7/site-packages (from scikit-learn->sklearn) (0.15.1)
Requirement already satisfied: numpy>=1.11.0 in /opt/conda/lib/python3.7/site-packages (from scikit-learn->sklearn) (1.19.5)
Requirement already satisfied: scipy>=0.17.0 in /opt/conda/lib/python3.7/site-packages (from scikit-learn->sklearn) (1.4.1)
Note: you may need to restart the kernel to use updated packages.
```

```
[3]: pip install numpy
```

```
Requirement already satisfied: numpy in /opt/conda/lib/python3.7/site-packages (1.19.5)
Note: you may need to restart the kernel to use updated packages.
```

```
[4]: pip install keras
```

```
Collecting keras
  Using cached Keras-2.4.3-py2.py3-none-any.whl (36 kB)
```

```

Requirement already satisfied: scipy>=0.14 in /opt/conda/lib/python3.7/site-
packages (from keras) (1.4.1)
Requirement already satisfied: numpy>=1.9.1 in /opt/conda/lib/python3.7/site-
packages (from keras) (1.19.5)
Requirement already satisfied: pyyaml in /opt/conda/lib/python3.7/site-packages
(from keras) (5.3.1)
Requirement already satisfied: h5py in /opt/conda/lib/python3.7/site-packages
(from keras) (2.10.0)
Requirement already satisfied: six in /opt/conda/lib/python3.7/site-packages
(from h5py->keras) (1.15.0)
Installing collected packages: keras
Successfully installed keras-2.4.3
Note: you may need to restart the kernel to use updated packages.

```

```
[5]: pip install tensorflow
```

```

Requirement already satisfied: tensorflow in /opt/conda/lib/python3.7/site-
packages (2.4.1)
Requirement already satisfied: flatbuffers~=1.12.0 in
/opt/conda/lib/python3.7/site-packages (from tensorflow) (1.12)
Requirement already satisfied: opt-einsum~=3.3.0 in
/opt/conda/lib/python3.7/site-packages (from tensorflow) (3.3.0)
Requirement already satisfied: wheel~=0.35 in /opt/conda/lib/python3.7/site-
packages (from tensorflow) (0.36.2)
Requirement already satisfied: wrapt~=1.12.1 in /opt/conda/lib/python3.7/site-
packages (from tensorflow) (1.12.1)
Requirement already satisfied: tensorboard~=2.4 in
/opt/conda/lib/python3.7/site-packages (from tensorflow) (2.4.1)
Requirement already satisfied: grpcio~=1.32.0 in /opt/conda/lib/python3.7/site-
packages (from tensorflow) (1.32.0)
Requirement already satisfied: keras-preprocessing~=1.1.2 in
/opt/conda/lib/python3.7/site-packages (from tensorflow) (1.1.2)
Requirement already satisfied: h5py~=2.10.0 in /opt/conda/lib/python3.7/site-
packages (from tensorflow) (2.10.0)
Requirement already satisfied: google-pasta~=0.2 in
/opt/conda/lib/python3.7/site-packages (from tensorflow) (0.2.0)
Requirement already satisfied: protobuf>=3.9.2 in /opt/conda/lib/python3.7/site-
packages (from tensorflow) (3.11.4)
Requirement already satisfied: typing-extensions~=3.7.4 in
/opt/conda/lib/python3.7/site-packages (from tensorflow) (3.7.4.2)
Requirement already satisfied: astunparse~=1.6.3 in
/opt/conda/lib/python3.7/site-packages (from tensorflow) (1.6.3)
Requirement already satisfied: absl-py~=0.10 in /opt/conda/lib/python3.7/site-
packages (from tensorflow) (0.12.0)
Requirement already satisfied: tensorflow-estimator<2.5.0,>=2.4.0 in
/opt/conda/lib/python3.7/site-packages (from tensorflow) (2.4.0)
Requirement already satisfied: termcolor~=1.1.0 in
/opt/conda/lib/python3.7/site-packages (from tensorflow) (1.1.0)

```

Requirement already satisfied: gast==0.3.3 in /opt/conda/lib/python3.7/site-packages (from tensorflow) (0.3.3)

Requirement already satisfied: numpy~=1.19.2 in /opt/conda/lib/python3.7/site-packages (from tensorflow) (1.19.5)

Requirement already satisfied: six~=1.15.0 in /opt/conda/lib/python3.7/site-packages (from tensorflow) (1.15.0)

Requirement already satisfied: google-auth<2,>=1.6.3 in /opt/conda/lib/python3.7/site-packages (from tensorboard~=2.4->tensorflow) (1.16.1)

Requirement already satisfied: setuptools>=41.0.0 in /opt/conda/lib/python3.7/site-packages (from tensorboard~=2.4->tensorflow) (46.1.3.post20200325)

Requirement already satisfied: markdown>=2.6.8 in /opt/conda/lib/python3.7/site-packages (from tensorboard~=2.4->tensorflow) (3.3.4)

Requirement already satisfied: tensorboard-plugin-wit>=1.6.0 in /opt/conda/lib/python3.7/site-packages (from tensorboard~=2.4->tensorflow) (1.8.0)

Requirement already satisfied: google-auth-oauthlib<0.5,>=0.4.1 in /opt/conda/lib/python3.7/site-packages (from tensorboard~=2.4->tensorflow) (0.4.3)

Requirement already satisfied: requests<3,>=2.21.0 in /opt/conda/lib/python3.7/site-packages (from tensorboard~=2.4->tensorflow) (2.23.0)

Requirement already satisfied: werkzeug>=0.11.15 in /opt/conda/lib/python3.7/site-packages (from tensorboard~=2.4->tensorflow) (1.0.1)

Requirement already satisfied: cachetools<5.0,>=2.0.0 in /opt/conda/lib/python3.7/site-packages (from google-auth<2,>=1.6.3->tensorboard~=2.4->tensorflow) (4.1.0)

Requirement already satisfied: pyasn1-modules>=0.2.1 in /opt/conda/lib/python3.7/site-packages (from google-auth<2,>=1.6.3->tensorboard~=2.4->tensorflow) (0.2.8)

Requirement already satisfied: rsa<4.1,>=3.1.4 in /opt/conda/lib/python3.7/site-packages (from google-auth<2,>=1.6.3->tensorboard~=2.4->tensorflow) (4.0)

Requirement already satisfied: importlib-metadata; python\_version < "3.8" in /opt/conda/lib/python3.7/site-packages (from markdown>=2.6.8->tensorboard~=2.4->tensorflow) (1.6.0)

Requirement already satisfied: requests-oauthlib>=0.7.0 in /opt/conda/lib/python3.7/site-packages (from google-auth-oauthlib<0.5,>=0.4.1->tensorboard~=2.4->tensorflow) (1.3.0)

Requirement already satisfied: urllib3!=1.25.0,!1.25.1,<1.26,>=1.21.1 in /opt/conda/lib/python3.7/site-packages (from requests<3,>=2.21.0->tensorboard~=2.4->tensorflow) (1.25.9)

Requirement already satisfied: chardet<4,>=3.0.2 in /opt/conda/lib/python3.7/site-packages (from requests<3,>=2.21.0->tensorboard~=2.4->tensorflow) (3.0.4)

Requirement already satisfied: idna<3,>=2.5 in /opt/conda/lib/python3.7/site-packages (from requests<3,>=2.21.0->tensorboard~=2.4->tensorflow) (2.9)

Requirement already satisfied: certifi>=2017.4.17 in /opt/conda/lib/python3.7/site-packages (from requests<3,>=2.21.0->tensorboard~=2.4->tensorflow) (2020.4.5.2)  
 Requirement already satisfied: pyasn1<0.5.0,>=0.4.6 in /opt/conda/lib/python3.7/site-packages (from pyasn1-modules>=0.2.1->google-auth<2,>=1.6.3->tensorboard~=2.4->tensorflow) (0.4.8)  
 Requirement already satisfied: zipp>=0.5 in /opt/conda/lib/python3.7/site-packages (from importlib-metadata; python\_version < "3.8"->markdown>=2.6.8->tensorboard~=2.4->tensorflow) (3.1.0)  
 Requirement already satisfied: oauthlib>=3.0.0 in /opt/conda/lib/python3.7/site-packages (from requests-oauthlib>=0.7.0->google-auth-oauthlib<0.5,>=0.4.1->tensorboard~=2.4->tensorflow) (3.0.1)  
 Note: you may need to restart the kernel to use updated packages.

[6]: `pip install matplotlib`

Requirement already satisfied: matplotlib in /opt/conda/lib/python3.7/site-packages (3.2.1)  
 Requirement already satisfied: kiwisolver>=1.0.1 in /opt/conda/lib/python3.7/site-packages (from matplotlib) (1.2.0)  
 Requirement already satisfied: numpy>=1.11 in /opt/conda/lib/python3.7/site-packages (from matplotlib) (1.19.5)  
 Requirement already satisfied: cyclor>=0.10 in /opt/conda/lib/python3.7/site-packages (from matplotlib) (0.10.0)  
 Requirement already satisfied: python-dateutil>=2.1 in /opt/conda/lib/python3.7/site-packages (from matplotlib) (2.8.1)  
 Requirement already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 in /opt/conda/lib/python3.7/site-packages (from matplotlib) (2.4.7)  
 Requirement already satisfied: six in /opt/conda/lib/python3.7/site-packages (from cyclor>=0.10->matplotlib) (1.15.0)  
 Note: you may need to restart the kernel to use updated packages.

[18]: `import pandas as pd  
 from sklearn import preprocessing  
 import numpy as np  
 np.random.seed(1)  
 import keras  
 from keras.models import Model  
 from keras.layers import Activation, Dense, Dropout, Input, LSTM, concatenate  
 from keras import optimizers  
 import tensorflow as tf  
 tf.random.set_seed(1)`

[19]: `#Function to calculate Exponential Moving Average using simple moving average  
 def ema(values, i):  
#Simple moving average  
 sma = np.mean(values[:, 3])`

```

ema_values = [sma]
k = 2 / (1 + i)
for i in range(len(his) - i, len(his)):
    close = his[i][3]
    ema_values.append(close * k + ema_values[-1] * (1 - k))
return ema_values[-1]

```

```

[20]: #history_points represents the set number of days that affect the next
history_points = 50

```

```

[21]: #Import .csv data and remove the Date attribute
data = pd.read_csv('SPY.csv')
data = data.drop('Date', axis=1)
data = data.values

```

```

[22]: #Normalize the data to a MinMaxScaler
data_normaliser = preprocessing.MinMaxScaler()
data_normalised = data_normaliser.fit_transform(data)

```

```

[23]: #Using history_points, open, close, high, low, volume, data, points create
      →OpenHighLowChart and next day open values
ohlc_histories_normalised = np.array([data_normalised[i:i + history_points].
      →copy() for i in range(len(data_normalised) - history_points)])
next_day_open_values_normalised = np.array([data_normalised[:, 0][i +
      →history_points].copy() for i in range(len(data_normalised) -
      →history_points)])
next_day_open_values_normalised = np.
      →expand_dims(next_day_open_values_normalised, -1)

```

```

[24]: next_day_open_values = np.array([data[:, 0][i + history_points].copy() for i in
      →range(len(data) - history_points)])
next_day_open_values = np.expand_dims(next_day_open_values, -1)

```

```

[25]: #Normalize the data to a MinMaxScaler
y_normaliser = preprocessing.MinMaxScaler()
y_normaliser.fit(next_day_open_values)

```

```

[25]: MinMaxScaler(copy=True, feature_range=(0, 1))

```

```

[26]: #Create technical indicators using simple and exponential moving average
technical_indicators = []
for his in ohlc_histories_normalised:
    #Simple moving average of the closing price
    sma = np.mean(his[:, 3])
    macd = ema(his, 12) - ema(his, 26)
    #Add the simple and exponential moving average to the technical indicator
    technical_indicators.append(np.array([sma, macd,]))

```

```

[27]: technical_indicators = np.array(technical_indicators)

tech_ind_scaler = preprocessing.MinMaxScaler()
technical_indicators_normalised = tech_ind_scaler.
    ↪fit_transform(technical_indicators)

[29]: assert ohlcv_histories_normalised.shape[0] == next_day_open_values_normalised.
    ↪shape[0] == technical_indicators_normalised.shape[0], "data shapes are_
    ↪inconsistent"

[30]: #Data Preparation
#Split the data into Training and Testing 9-1 ratio
test_split = 0.9
#Split number (6322)
n = int(ohlcv_histories_normalised.shape[0] * test_split)

[31]: ohlcv_train = ohlcv_histories_normalised[:n]
tech_ind_train = technical_indicators_normalised[:n]
y_train = next_day_open_values_normalised[:n]

[32]: ohlcv_test = ohlcv_histories_normalised[n:]
tech_ind_test = technical_indicators_normalised[n:]
y_test = next_day_open_values_normalised[n:]

unscaled_y_test = next_day_open_values[n:]

[33]: print(ohlcv_train.shape)
print(ohlcv_test.shape)

(6322, 50, 6)
(703, 50, 6)

[34]: #Create LSTM model
#Define two sets of inputs
lstm_input = Input(shape=(history_points, 6), name='lstm_input')
dense_input = Input(shape=(technical_indicators_normalised.shape[1],),
    ↪name='tech_input')

[35]: #First branch operates on the first input
x = LSTM(history_points, name='lstm_0')(lstm_input)
x = Dropout(0.2, name='lstm_dropout_0')(x)
lstm_branch = Model(inputs=lstm_input, outputs=x)

[36]: #Second branch operates on the second input
y = Dense(20, name='tech_dense_0')(dense_input)
y = Activation("relu", name='tech_relu_0')(y)
y = Dropout(0.2, name='tech_dropout_0')(y)

```

```
technical_indicators_branch = Model(inputs=dense_input, outputs=y)
```

```
[37]: #Combine the output of the two branches  
combined = concatenate([lstm_branch.output, technical_indicators_branch.  
    ↳output], name='concatenate')
```

```
[38]: z = Dense(64, activation="sigmoid", name='dense_pooling')(combined)  
z = Dense(1, activation="linear", name='dense_out')(z)
```

```
[39]: #Model will accept the inputs of the two branches and then output a single value  
model = Model(inputs=[lstm_branch.input, technical_indicators_branch.input],  
    ↳outputs=z)  
adam = optimizers.Adam(lr=0.0005)  
model.compile(optimizer=adam, loss='mse')
```

```
[40]: #Train Model model.fit(x_train, y_train, validation_data = (x_test, y_test),  
    ↳epochs = 100, batch_size = 64, verbose = 1)  
model.fit(x=[ohlc_v_train, tech_ind_train], y=y_train, batch_size=32,  
    ↳epochs=history_points, shuffle=True, validation_split=0.1)
```

```
Epoch 1/50  
178/178 [=====] - 8s 35ms/step - loss: 0.1805 -  
val_loss: 0.0060  
Epoch 2/50  
178/178 [=====] - 5s 31ms/step - loss: 0.0014 -  
val_loss: 0.0032  
Epoch 3/50  
178/178 [=====] - 6s 31ms/step - loss: 9.0594e-04 -  
val_loss: 5.2866e-04  
Epoch 4/50  
178/178 [=====] - 5s 31ms/step - loss: 7.6005e-04 -  
val_loss: 2.5086e-04  
Epoch 5/50  
178/178 [=====] - 5s 30ms/step - loss: 6.3473e-04 -  
val_loss: 1.7366e-04  
Epoch 6/50  
178/178 [=====] - 5s 30ms/step - loss: 5.6505e-04 -  
val_loss: 2.6322e-04  
Epoch 7/50  
178/178 [=====] - 5s 30ms/step - loss: 4.4788e-04 -  
val_loss: 1.4584e-04  
Epoch 8/50  
178/178 [=====] - 5s 30ms/step - loss: 4.3198e-04 -  
val_loss: 2.6031e-04  
Epoch 9/50  
178/178 [=====] - 5s 31ms/step - loss: 3.7250e-04 -  
val_loss: 1.8484e-04
```

Epoch 10/50  
178/178 [=====] - 5s 30ms/step - loss: 3.2943e-04 -  
val\_loss: 1.7774e-04  
Epoch 11/50  
178/178 [=====] - 5s 31ms/step - loss: 2.8809e-04 -  
val\_loss: 3.8735e-04  
Epoch 12/50  
178/178 [=====] - 5s 31ms/step - loss: 2.5150e-04 -  
val\_loss: 9.7225e-04  
Epoch 13/50  
178/178 [=====] - 5s 30ms/step - loss: 2.3261e-04 -  
val\_loss: 2.9564e-04  
Epoch 14/50  
178/178 [=====] - 5s 30ms/step - loss: 2.0428e-04 -  
val\_loss: 2.7359e-04  
Epoch 15/50  
178/178 [=====] - 5s 31ms/step - loss: 1.6716e-04 -  
val\_loss: 1.6657e-04  
Epoch 16/50  
178/178 [=====] - 5s 30ms/step - loss: 1.6480e-04 -  
val\_loss: 4.0306e-04  
Epoch 17/50  
178/178 [=====] - 5s 30ms/step - loss: 1.4587e-04 -  
val\_loss: 2.3041e-04  
Epoch 18/50  
178/178 [=====] - 5s 30ms/step - loss: 1.3768e-04 -  
val\_loss: 3.3396e-04  
Epoch 19/50  
178/178 [=====] - 5s 30ms/step - loss: 1.3082e-04 -  
val\_loss: 3.3412e-04  
Epoch 20/50  
178/178 [=====] - 6s 31ms/step - loss: 1.3234e-04 -  
val\_loss: 1.3238e-04  
Epoch 21/50  
178/178 [=====] - 5s 30ms/step - loss: 1.2970e-04 -  
val\_loss: 2.1762e-04  
Epoch 22/50  
178/178 [=====] - 5s 31ms/step - loss: 1.2348e-04 -  
val\_loss: 8.5117e-05  
Epoch 23/50  
178/178 [=====] - 5s 31ms/step - loss: 1.1573e-04 -  
val\_loss: 7.1877e-05  
Epoch 24/50  
178/178 [=====] - 5s 31ms/step - loss: 1.1672e-04 -  
val\_loss: 1.0376e-04  
Epoch 25/50  
178/178 [=====] - 5s 31ms/step - loss: 1.1639e-04 -  
val\_loss: 2.7646e-04



Epoch 26/50  
178/178 [=====] - 5s 31ms/step - loss: 1.0649e-04 -  
val\_loss: 7.0091e-05  
Epoch 27/50  
178/178 [=====] - 6s 31ms/step - loss: 1.1338e-04 -  
val\_loss: 3.5387e-04  
Epoch 28/50  
178/178 [=====] - 5s 30ms/step - loss: 1.1724e-04 -  
val\_loss: 1.6329e-04  
Epoch 29/50  
178/178 [=====] - 5s 30ms/step - loss: 1.1154e-04 -  
val\_loss: 1.5559e-04  
Epoch 30/50  
178/178 [=====] - 5s 30ms/step - loss: 1.1647e-04 -  
val\_loss: 8.4815e-05  
Epoch 31/50  
178/178 [=====] - 6s 31ms/step - loss: 1.1257e-04 -  
val\_loss: 7.3133e-05  
Epoch 32/50  
178/178 [=====] - 5s 30ms/step - loss: 1.0414e-04 -  
val\_loss: 5.4363e-05  
Epoch 33/50  
178/178 [=====] - 5s 30ms/step - loss: 1.0121e-04 -  
val\_loss: 1.1910e-04  
Epoch 34/50  
178/178 [=====] - 6s 32ms/step - loss: 1.1464e-04 -  
val\_loss: 1.4560e-04  
Epoch 35/50  
178/178 [=====] - 6s 32ms/step - loss: 1.0329e-04 -  
val\_loss: 1.5857e-04  
Epoch 36/50  
178/178 [=====] - 6s 32ms/step - loss: 1.0198e-04 -  
val\_loss: 1.9201e-04  
Epoch 37/50  
178/178 [=====] - 6s 31ms/step - loss: 1.0479e-04 -  
val\_loss: 8.9917e-05  
Epoch 38/50  
178/178 [=====] - 5s 30ms/step - loss: 1.1059e-04 -  
val\_loss: 8.0025e-05  
Epoch 39/50  
178/178 [=====] - 5s 30ms/step - loss: 9.7159e-05 -  
val\_loss: 4.1494e-04  
Epoch 40/50  
178/178 [=====] - 5s 30ms/step - loss: 1.1652e-04 -  
val\_loss: 1.2758e-04  
Epoch 41/50  
178/178 [=====] - 5s 30ms/step - loss: 1.1206e-04 -  
val\_loss: 1.3997e-04

```

Epoch 42/50
178/178 [=====] - 6s 31ms/step - loss: 1.0116e-04 -
val_loss: 9.2593e-05
Epoch 43/50
178/178 [=====] - 5s 30ms/step - loss: 9.2469e-05 -
val_loss: 3.8650e-05
Epoch 44/50
178/178 [=====] - 5s 30ms/step - loss: 1.1257e-04 -
val_loss: 1.5640e-04
Epoch 45/50
178/178 [=====] - 5s 30ms/step - loss: 1.0092e-04 -
val_loss: 2.8553e-04
Epoch 46/50
178/178 [=====] - 5s 30ms/step - loss: 1.0463e-04 -
val_loss: 8.8256e-05
Epoch 47/50
178/178 [=====] - 5s 30ms/step - loss: 9.4947e-05 -
val_loss: 8.2154e-05
Epoch 48/50
178/178 [=====] - 6s 31ms/step - loss: 1.0147e-04 -
val_loss: 1.5089e-04
Epoch 49/50
178/178 [=====] - 5s 30ms/step - loss: 1.0448e-04 -
val_loss: 4.3716e-05
Epoch 50/50
178/178 [=====] - 5s 30ms/step - loss: 1.0445e-04 -
val_loss: 1.6957e-04

```

```
[40]: <tensorflow.python.keras.callbacks.History at 0x7f403f646510>
```

```
[41]: #Predict and check model performance
y_test_predicted = model.predict([ohlc_test, tech_ind_test])
y_predicted = model.predict([ohlc_histories_normalised, technical_indicators])
```

```
[42]: #Denormalization or scaler inverse
y_test_predicted = y_normaliser.inverse_transform(y_test_predicted)
y_predicted = y_normaliser.inverse_transform(y_predicted)
```

```
[43]: assert unscaled_y_test.shape == y_test_predicted.shape, "data shapes are
↳inconsistent"
```

```
[44]: #Calculate Root Mean Square Error performance
rmse = np.mean(np.square(unscaled_y_test - y_test_predicted))
scaled_mse = rmse / (np.max(unscaled_y_test) - np.min(unscaled_y_test)) * 100
print(scaled_mse)
```

```
76.2622006792431
```

```
[45]: import matplotlib.pyplot as plt
#Plot
plt.gcf().set_size_inches(18, 9, forward=True)
real = plt.plot(unscaled_y_test[0:-1], label='real')
#Shift test predictions for plotting
pred = plt.plot(y_test_predicted[0:-1], label='predicted')
#Plot baseline and prediction
plt.legend(['Real', 'Predicted'])
plt.show()
```



```
[46]: #Save model for trade program
model.save(f'time_series_model.h5')
```