TesLSTM about:srcdoc

```
In [1]:
        # Importing library
        import numpy as np
        import matplotlib.pyplot as plt
        import pandas as pd
In [2]:
        # Importing training set
        dataset train = pd.read csv('Google Stock Price Train.csv')
        training set = dataset train.iloc[:, 1:2].values
In [3]:
        # feature scaling
        from sklearn.preprocessing import MinMaxScaler
        sc = MinMaxScaler(feature range = (0,1))
        training set scaled = sc.fit transform(training set)
In [4]:
        # Creating a data scructure with 60 timesteps and 1 output
        X train = []
        Y train = []
        for i in range (60, 1258):
            X train.append(training set scaled[i-60:i, 0])
            Y_train.append(training_set_scaled[i, 0])
        X_train, Y_train = np.array(X_train), np.array(Y_train)
        # Reshaping
        X train = np.reshape(X train, (X train.shape[0], X train.shape[1], 1))
In [5]:
        # Importing KERAS
        from keras.models import Sequential
        from keras.layers import Dense
        from keras.layers import LSTM
        from keras.layers import Dropout
In [6]:
        # Initializing the RNN
        regressor = Sequential()
        # Adding the first LSTM layer and some dropout regularization
        regressor.add(LSTM(units=50, return sequences=True, input shape=(X train.sl
        regressor.add(Dropout(0.2))
In [7]:
        # Adding second LSTM layer and some another dropout regularization
        regressor.add(LSTM(units=50, return sequences=True))
        regressor.add(Dropout(0.2))
In [8]:
        # Adding third LSTM layer and some another dropout regularization
        regressor.add(LSTM(units=50, return sequences=True))
        regressor.add(Dropout(0.2))
In [9]:
        # Adding fourth LSTM layer and some another dropout regularization
        regressor.add(LSTM(units=50))
        regressor.add(Dropout(0.2))
```

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```
In [10]:
       # Adding output layer
       regressor.add(Dense(units=1))
In [11]:
       # Compiling
       regressor.compile(optimizer='adam', loss='mean squared error')
In [12]:
       # fitting the RNN to training set
       regressor.fit(X train, Y train, epochs=150, batch size=32)
      Epoch 1/150
      Epoch 2/150
      38/38 [=============== ] - 5s 130ms/step - loss: 0.0056
      Epoch 3/150
      38/38 [============== ] - 6s 148ms/step - loss: 0.0059
      Epoch 4/150
      38/38 [============ ] - 6s 148ms/step - loss: 0.0050
      Epoch 5/150
      38/38 [============== ] - 5s 130ms/step - loss: 0.0050
      Epoch 6/150
      38/38 [============ ] - 6s 147ms/step - loss: 0.0047
      Epoch 7/150
      38/38 [============ ] - 6s 145ms/step - loss: 0.0050
      Epoch 8/150
      38/38 [============= ] - 5s 130ms/step - loss: 0.0045
      Epoch 9/150
      38/38 [============== ] - 6s 146ms/step - loss: 0.0045
      Epoch 10/150
      38/38 [============ ] - 6s 146ms/step - loss: 0.0043
      Epoch 11/150
      38/38 [============ ] - 5s 126ms/step - loss: 0.0042
      Epoch 12/150
      Epoch 13/150
      Epoch 14/150
      38/38 [============== ] - 5s 126ms/step - loss: 0.0036
      Epoch 15/150
      38/38 [============ ] - 5s 142ms/step - loss: 0.0040
      Epoch 16/150
      38/38 [=============== ] - 6s 145ms/step - loss: 0.0039
      Epoch 17/150
      38/38 [============== ] - 5s 136ms/step - loss: 0.0035
      Epoch 18/150
      38/38 [============== ] - 5s 134ms/step - loss: 0.0034
      Epoch 19/150
      38/38 [============== ] - 5s 144ms/step - loss: 0.0042
      Epoch 20/150
      38/38 [=============== ] - 5s 135ms/step - loss: 0.0036
      Epoch 21/150
      38/38 [============ ] - 5s 136ms/step - loss: 0.0035
      Epoch 22/150
      38/38 [============== ] - 6s 146ms/step - loss: 0.0035
      Epoch 23/150
      38/38 [=============== ] - 5s 138ms/step - loss: 0.0035
      Epoch 24/150
      38/38 [=============== ] - 5s 130ms/step - loss: 0.0031
      Epoch 25/150
      38/38 [============= ] - 5s 144ms/step - loss: 0.0035
```

```
Epoch 26/150
38/38 [============= ] - 5s 135ms/step - loss: 0.0030
Epoch 27/150
38/38 [============== ] - 5s 123ms/step - loss: 0.0029
Epoch 28/150
38/38 [=============== ] - 6s 152ms/step - loss: 0.0029
Epoch 29/150
38/38 [============ ] - 5s 123ms/step - loss: 0.0030
Epoch 30/150
Epoch 31/150
38/38 [============== ] - 5s 141ms/step - loss: 0.0030
Epoch 32/150
38/38 [=============== ] - 6s 146ms/step - loss: 0.0029
Epoch 33/150
38/38 [============== ] - 5s 127ms/step - loss: 0.0029
Epoch 34/150
Epoch 35/150
38/38 [============= ] - 6s 146ms/step - loss: 0.0028
Epoch 36/150
38/38 [============== ] - 5s 122ms/step - loss: 0.0029
Epoch 37/150
38/38 [=============== ] - 5s 143ms/step - loss: 0.0028
Epoch 38/150
38/38 [============= ] - 6s 147ms/step - loss: 0.0028
Epoch 39/150
38/38 [============= ] - 5s 127ms/step - loss: 0.0026
Epoch 40/150
38/38 [============== ] - 6s 147ms/step - loss: 0.0029
Epoch 41/150
38/38 [=============== ] - 5s 144ms/step - loss: 0.0023
Epoch 42/150
Epoch 43/150
38/38 [============== ] - 6s 146ms/step - loss: 0.0025
Epoch 44/150
38/38 [=============== ] - 6s 146ms/step - loss: 0.0025
Epoch 45/150
38/38 [=============== ] - 5s 131ms/step - loss: 0.0025
Epoch 46/150
38/38 [============ ] - 6s 146ms/step - loss: 0.0023
Epoch 47/150
38/38 [============= ] - 5s 138ms/step - loss: 0.0023
Epoch 48/150
38/38 [=============== ] - 5s 127ms/step - loss: 0.0024
Epoch 49/150
38/38 [============== ] - 5s 143ms/step - loss: 0.0026
Epoch 50/150
38/38 [=============== ] - 5s 144ms/step - loss: 0.0024
Epoch 51/150
38/38 [============= ] - 5s 128ms/step - loss: 0.0021
Epoch 52/150
38/38 [============= ] - 6s 145ms/step - loss: 0.0021
Epoch 53/150
38/38 [=============== ] - 6s 147ms/step - loss: 0.0026
Epoch 54/150
38/38 [=============== ] - 5s 129ms/step - loss: 0.0024
Epoch 55/150
38/38 [============ ] - 6s 147ms/step - loss: 0.0020
Epoch 56/150
38/38 [=============== ] - 6s 149ms/step - loss: 0.0021
Epoch 57/150
```

```
38/38 [=============== ] - 5s 126ms/step - loss: 0.0022
Epoch 58/150
38/38 [============== ] - 5s 142ms/step - loss: 0.0024
Epoch 59/150
38/38 [============= ] - 5s 146ms/step - loss: 0.0022
Epoch 60/150
38/38 [=============== ] - 5s 125ms/step - loss: 0.0020
Epoch 61/150
38/38 [============= ] - 5s 144ms/step - loss: 0.0023
Epoch 62/150
38/38 [============== ] - 6s 149ms/step - loss: 0.0020
Epoch 63/150
38/38 [============== ] - 5s 123ms/step - loss: 0.0019
Epoch 64/150
38/38 [============== ] - 6s 148ms/step - loss: 0.0020
Epoch 65/150
38/38 [============ ] - 6s 147ms/step - loss: 0.0020
Epoch 66/150
38/38 [=============== ] - 5s 129ms/step - loss: 0.0018
Epoch 67/150
38/38 [=============== ] - 6s 146ms/step - loss: 0.0019
Epoch 68/150
38/38 [============== ] - 6s 148ms/step - loss: 0.0020
Epoch 69/150
38/38 [============= ] - 5s 127ms/step - loss: 0.0019
Epoch 70/150
Epoch 71/150
38/38 [=============== ] - 6s 147ms/step - loss: 0.0018
Epoch 72/150
38/38 [============= ] - 5s 124ms/step - loss: 0.0018
Epoch 73/150
38/38 [=============== ] - 5s 144ms/step - loss: 0.0017
Epoch 74/150
38/38 [============= ] - 6s 146ms/step - loss: 0.0018
Epoch 75/150
38/38 [============ ] - 5s 127ms/step - loss: 0.0017
Epoch 76/150
38/38 [============== ] - 6s 147ms/step - loss: 0.0017
Epoch 77/150
38/38 [=============== ] - 6s 148ms/step - loss: 0.0016
Epoch 78/150
Epoch 79/150
38/38 [============== ] - 6s 149ms/step - loss: 0.0015
Epoch 80/150
38/38 [=============== ] - 6s 148ms/step - loss: 0.0018
Epoch 81/150
38/38 [=============== ] - 5s 123ms/step - loss: 0.0016
Epoch 82/150
38/38 [=============== ] - 5s 138ms/step - loss: 0.0016
Epoch 83/150
38/38 [============== ] - 6s 149ms/step - loss: 0.0017
Epoch 84/150
38/38 [=============== ] - 5s 124ms/step - loss: 0.0014
Epoch 85/150
38/38 [=============== ] - 5s 138ms/step - loss: 0.0016
Epoch 86/150
38/38 [=============== ] - 5s 122ms/step - loss: 0.0015
Epoch 87/150
38/38 [============== ] - 5s 122ms/step - loss: 0.0016
Epoch 88/150
38/38 [============ ] - 4s 110ms/step - loss: 0.0014
```

```
Epoch 89/150
38/38 [============= ] - 5s 128ms/step - loss: 0.0015
Epoch 90/150
38/38 [============== ] - 5s 122ms/step - loss: 0.0017
Epoch 91/150
38/38 [============== ] - 4s 104ms/step - loss: 0.0015
Epoch 92/150
38/38 [============ ] - 4s 118ms/step - loss: 0.0014
Epoch 93/150
Epoch 94/150
38/38 [============== ] - 4s 113ms/step - loss: 0.0016
Epoch 95/150
38/38 [=============== ] - 5s 130ms/step - loss: 0.0016
Epoch 96/150
38/38 [============== ] - 5s 132ms/step - loss: 0.0015
Epoch 97/150
Epoch 98/150
38/38 [============== ] - 5s 122ms/step - loss: 0.0017
Epoch 99/150
38/38 [============== ] - 5s 127ms/step - loss: 0.0012
Epoch 100/150
38/38 [=============== ] - 5s 129ms/step - loss: 0.0014
Epoch 101/150
38/38 [============= ] - 5s 128ms/step - loss: 0.0015
Epoch 102/150
38/38 [============= ] - 6s 152ms/step - loss: 0.0015
Epoch 103/150
38/38 [=============== ] - 5s 140ms/step - loss: 0.0014
Epoch 104/150
38/38 [=============== ] - 5s 126ms/step - loss: 0.0015
Epoch 105/150
Epoch 106/150
38/38 [============== ] - 5s 144ms/step - loss: 0.0020
Epoch 107/150
38/38 [=============== ] - 5s 137ms/step - loss: 0.0016
Epoch 108/150
38/38 [=============== ] - 5s 136ms/step - loss: 0.0012
Epoch 109/150
38/38 [============ ] - 5s 136ms/step - loss: 0.0014
Epoch 110/150
38/38 [============= ] - 5s 138ms/step - loss: 0.0012
Epoch 111/150
38/38 [=============== ] - 5s 134ms/step - loss: 0.0013
Epoch 112/150
38/38 [============== ] - 5s 140ms/step - loss: 0.0013
Epoch 113/150
38/38 [=============== ] - 5s 140ms/step - loss: 0.0016
Epoch 114/150
38/38 [============= ] - 5s 139ms/step - loss: 0.0014
Epoch 115/150
38/38 [============= ] - 5s 141ms/step - loss: 0.0013
Epoch 116/150
38/38 [=============== ] - 6s 146ms/step - loss: 0.0012
Epoch 117/150
38/38 [============== ] - 5s 138ms/step - loss: 0.0012
Epoch 118/150
38/38 [============= ] - 6s 155ms/step - loss: 0.0014
Epoch 119/150
38/38 [=============== ] - 5s 142ms/step - loss: 0.0013
Epoch 120/150
```

```
38/38 [=============== ] - 6s 147ms/step - loss: 0.0014
      Epoch 121/150
       38/38 [=============== ] - 6s 154ms/step - loss: 0.0015
       Epoch 122/150
       38/38 [============== ] - 5s 140ms/step - loss: 0.0012
      Epoch 123/150
       38/38 [=============== ] - 6s 153ms/step - loss: 0.0013
      Epoch 124/150
       38/38 [============= ] - 6s 153ms/step - loss: 0.0013
      Epoch 125/150
       38/38 [=============== ] - 5s 135ms/step - loss: 0.0012
      Epoch 126/150
       38/38 [============== ] - 6s 153ms/step - loss: 0.0012
      Epoch 127/150
       38/38 [=============== ] - 6s 152ms/step - loss: 0.0012
      Epoch 128/150
       38/38 [============= ] - 5s 133ms/step - loss: 0.0012
      Epoch 129/150
       38/38 [=============== ] - 6s 153ms/step - loss: 0.0012
      Epoch 130/150
       38/38 [=============== ] - 6s 153ms/step - loss: 0.0011
      Epoch 131/150
       38/38 [=============== ] - 5s 143ms/step - loss: 0.0013
      Epoch 132/150
       38/38 [============ ] - 6s 148ms/step - loss: 0.0014
      Epoch 133/150
       Epoch 134/150
       38/38 [============== ] - 6s 144ms/step - loss: 0.0011
      Epoch 135/150
       38/38 [============== ] - 6s 151ms/step - loss: 0.0012
      Epoch 136/150
       38/38 [=============== ] - 6s 149ms/step - loss: 0.0012
      Epoch 137/150
      38/38 [============ ] - 5s 141ms/step - loss: 0.0011
      Epoch 138/150
       38/38 [============= ] - 6s 149ms/step - loss: 0.0011
      Epoch 139/150
       38/38 [============== ] - 5s 141ms/step - loss: 0.0010
      Epoch 140/150
       38/38 [=============== ] - 6s 146ms/step - loss: 0.0014
      Epoch 141/150
       Epoch 142/150
       38/38 [=============== ] - 5s 142ms/step - loss: 0.0011
      Epoch 143/150
       38/38 [=============== ] - 6s 146ms/step - loss: 0.0011
      Epoch 144/150
       38/38 [=============== ] - 6s 149ms/step - loss: 0.0013
       Epoch 145/150
       38/38 [=============== ] - 5s 138ms/step - loss: 0.0011
      Epoch 146/150
       38/38 [============== ] - 6s 149ms/step - loss: 0.0011
      Epoch 147/150
       38/38 [=============== ] - 6s 149ms/step - loss: 0.0012
      Epoch 148/150
       38/38 [=============== ] - 5s 139ms/step - loss: 0.0011
      Epoch 149/150
       38/38 [=============== ] - 6s 150ms/step - loss: 0.0012
      Epoch 150/150
      <keras.callbacks.History at 0x1dd9289ff10>
Out[12]:
```

TesLSTM about:srcdoc

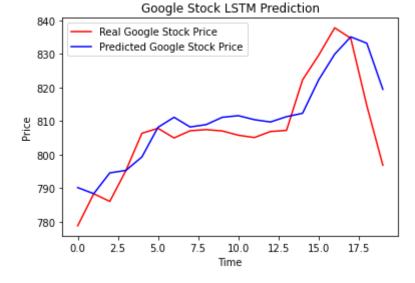
In [13]:

```
regressor.save('models/Epoch150')
        WARNING:absl:Found untraced functions such as 1stm cell layer call fn, 1stm
         cell layer call and return conditional losses, 1stm cell 1 layer call fn,
        lstm_cell_1_layer_call_and_return_conditional_losses, lstm_cell_2_layer_cal
        1 fn while saving (showing 5 of 8). These functions will not be directly ca
        llable after loading.
        INFO:tensorflow:Assets written to: models/Epoch150\assets
        INFO:tensorflow:Assets written to: models/Epoch150\assets
        WARNING:absl:<keras.layers.recurrent.LSTMCell object at 0x000001DD914080D0>
        has the same name 'LSTMCell' as a built-in Keras object. Consider renaming
        <class 'keras.layers.recurrent.LSTMCell'> to avoid naming conflicts when lo
        ading with `tf.keras.models.load_model`. If renaming is not possible, pass
        the object in the `custom objects` parameter of the load function.
        WARNING:absl:<keras.layers.recurrent.LSTMCell object at 0x000001DD926A5BE0>
        has the same name 'LSTMCell' as a built-in Keras object. Consider renaming
        <class 'keras.layers.recurrent.LSTMCell'> to avoid naming conflicts when lo
        ading with `tf.keras.models.load model`. If renaming is not possible, pass
        the object in the `custom objects` parameter of the load function.
        WARNING:absl:<keras.layers.recurrent.LSTMCell object at 0x000001DD9275C460>
        has the same name 'LSTMCell' as a built-in Keras object. Consider renaming
        <class 'keras.layers.recurrent.LSTMCell'> to avoid naming conflicts when lo
        ading with `tf.keras.models.load model`. If renaming is not possible, pass
        the object in the `custom objects` parameter of the load function.
        WARNING:absl:<keras.layers.recurrent.LSTMCell object at 0x000001DD927F9370>
        has the same name 'LSTMCell' as a built-in Keras object. Consider renaming
        <class 'keras.layers.recurrent.LSTMCell'> to avoid naming conflicts when lo
        ading with `tf.keras.models.load model`. If renaming is not possible, pass
        the object in the `custom objects` parameter of the load function.
In [14]:
         # Importing testing set
         dataset test = pd.read csv('Google Stock Price Test.csv')
         real stock set = dataset test.iloc[:, 1:2].values
In [15]:
         # Getting the predicted stock
         dataset total = pd.concat((dataset train['Open'], dataset test['Open']), ax
         inputs = dataset_total[len(dataset_total) - len(dataset_test) - 60:].values
         inputs = inputs.reshape(-1, 1)
         inputs = sc.transform(inputs)
         X \text{ test} = []
         for i in range(60, 80):
             X test.append(inputs[i-60:i, 0])
         X test = np.array(X test)
         X test = np.reshape(X test, (X test.shape[0], X test.shape[1], 1))
         predicted stock set = regressor.predict(X test)
         predicted stock set = sc.inverse transform(predicted stock set)
```

TesLSTM about:srcdoc

```
In [16]: # visualizing
    plt.plot(real_stock_set, color='red', label='Real Google Stock Price')
    plt.plot(predicted_stock_set, color='blue', label='Predicted Google Stock F
    plt.title('Google Stock LSTM Prediction')
    plt.xlabel('Time')
    plt.ylabel('Price')

    plt.legend()
    plt.show()
```



```
In [ ]:
```

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