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))
```

TesLSTM about:sredoc

```
In [10]:
       # Adding output layer
       regressor.add(Dense(units=1))
In [11]:
       # Compiling
       regressor.compile(optimizer='adam', loss='mean squared error')
In [16]:
       regressorting the RNN to training set
       regressor.fit(X train, Y train, epochs=250, batch size=32)
      Epoch 1/250
       38/38 [============== ] - 6s 145ms/step - loss: 0.0030
      Epoch 2/250
       38/38 [================= ] - 5s 131ms/step - loss: 0.0033
      Epoch 3/250
       38/38 [============== ] - 6s 152ms/step - loss: 0.0029
      Epoch 4/250
      38/38 [============= ] - 6s 149ms/step - loss: 0.0034
      Epoch 5/250
       38/38 [=============== ] - 5s 143ms/step - loss: 0.0028
      Epoch 6/250
      38/38 [============ ] - 6s 155ms/step - loss: 0.0032
      Epoch 7/250
       38/38 [============= ] - 6s 153ms/step - loss: 0.0028
      Epoch 8/250
       38/38 [============= ] - 5s 135ms/step - loss: 0.0029
      Epoch 9/250
      38/38 [============ ] - 6s 153ms/step - loss: 0.0028
      Epoch 10/250
      38/38 [============ ] - 6s 150ms/step - loss: 0.0030
      Epoch 11/250
       38/38 [============== ] - 5s 144ms/step - loss: 0.0028
      Epoch 12/250
      38/38 [============ ] - 6s 153ms/step - loss: 0.0026
      Epoch 13/250
       Epoch 14/250
       38/38 [============== ] - 5s 140ms/step - loss: 0.0025
      Epoch 15/250
       38/38 [============== ] - 6s 152ms/step - loss: 0.0030
      Epoch 16/250
       38/38 [=============== ] - 5s 136ms/step - loss: 0.0026
      Epoch 17/250
       38/38 [============= ] - 6s 148ms/step - loss: 0.0026
      Epoch 18/250
       38/38 [============== ] - 6s 150ms/step - loss: 0.0023
      Epoch 19/250
       38/38 [=============== ] - 5s 131ms/step - loss: 0.0025
      Epoch 20/250
       Epoch 21/250
      38/38 [============ ] - 6s 150ms/step - loss: 0.0025
      Epoch 22/250
       38/38 [=============== ] - 5s 130ms/step - loss: 0.0024
      Epoch 23/250
       38/38 [============== ] - 6s 148ms/step - loss: 0.0029
      Epoch 24/250
       38/38 [============== ] - 6s 149ms/step - loss: 0.0027
      Epoch 25/250
       38/38 [============= ] - 5s 132ms/step - loss: 0.0025
```

```
Epoch 26/250
38/38 [============= ] - 6s 152ms/step - loss: 0.0024
Epoch 27/250
38/38 [============== ] - 6s 152ms/step - loss: 0.0023
Epoch 28/250
38/38 [=============== ] - 5s 136ms/step - loss: 0.0023
Epoch 29/250
38/38 [============= ] - 6s 150ms/step - loss: 0.0022
Epoch 30/250
38/38 [=============== ] - 6s 148ms/step - loss: 0.0021
Epoch 31/250
38/38 [============== ] - 5s 142ms/step - loss: 0.0021
Epoch 32/250
38/38 [=============== ] - 6s 152ms/step - loss: 0.0024
Epoch 33/250
38/38 [============== ] - 6s 147ms/step - loss: 0.0022
Epoch 34/250
Epoch 35/250
38/38 [============= ] - 6s 150ms/step - loss: 0.0022
Epoch 36/250
38/38 [=============== ] - 6s 152ms/step - loss: 0.0021
Epoch 37/250
38/38 [=============== ] - 5s 140ms/step - loss: 0.0023
Epoch 38/250
38/38 [============= ] - 6s 151ms/step - loss: 0.0020
Epoch 39/250
38/38 [============= ] - 5s 142ms/step - loss: 0.0020
Epoch 40/250
38/38 [============== ] - 5s 139ms/step - loss: 0.0020
Epoch 41/250
38/38 [============== ] - 6s 152ms/step - loss: 0.0020
Epoch 42/250
Epoch 43/250
38/38 [============== ] - 5s 144ms/step - loss: 0.0017
Epoch 44/250
38/38 [=============== ] - 6s 145ms/step - loss: 0.0020
Epoch 45/250
38/38 [=============== ] - 5s 144ms/step - loss: 0.0019
Epoch 46/250
38/38 [============ ] - 6s 147ms/step - loss: 0.0025
Epoch 47/250
38/38 [============= ] - 6s 156ms/step - loss: 0.0019
Epoch 48/250
38/38 [=============== ] - 5s 132ms/step - loss: 0.0018
Epoch 49/250
38/38 [============== ] - 6s 152ms/step - loss: 0.0018
Epoch 50/250
38/38 [=============== ] - 6s 152ms/step - loss: 0.0017
Epoch 51/250
38/38 [============= ] - 5s 131ms/step - loss: 0.0018
Epoch 52/250
38/38 [============ ] - 6s 152ms/step - loss: 0.0017
Epoch 53/250
38/38 [=============== ] - 6s 148ms/step - loss: 0.0018
Epoch 54/250
38/38 [============== ] - 5s 132ms/step - loss: 0.0016
Epoch 55/250
38/38 [============= ] - 5s 145ms/step - loss: 0.0016
Epoch 56/250
38/38 [============== ] - 6s 151ms/step - loss: 0.0017
Epoch 57/250
```

```
38/38 [============= ] - 5s 134ms/step - loss: 0.0017
Epoch 58/250
38/38 [=============== ] - 6s 152ms/step - loss: 0.0017
Epoch 59/250
38/38 [============= ] - 6s 152ms/step - loss: 0.0016
Epoch 60/250
38/38 [=============== ] - 5s 134ms/step - loss: 0.0016
Epoch 61/250
38/38 [============= ] - 6s 152ms/step - loss: 0.0016
Epoch 62/250
38/38 [=============== ] - 6s 149ms/step - loss: 0.0015
Epoch 63/250
38/38 [============== ] - 5s 130ms/step - loss: 0.0015
Epoch 64/250
38/38 [=============== ] - 6s 154ms/step - loss: 0.0016
Epoch 65/250
38/38 [============ ] - 6s 155ms/step - loss: 0.0017
Epoch 66/250
38/38 [=============== ] - 5s 130ms/step - loss: 0.0018
Epoch 67/250
38/38 [=============== ] - 6s 149ms/step - loss: 0.0015
Epoch 68/250
38/38 [=============== ] - 6s 149ms/step - loss: 0.0014
Epoch 69/250
38/38 [============= ] - 5s 138ms/step - loss: 0.0015
Epoch 70/250
Epoch 71/250
38/38 [=============== ] - 5s 141ms/step - loss: 0.0015
Epoch 72/250
38/38 [============== ] - 5s 143ms/step - loss: 0.0015
Epoch 73/250
38/38 [=============== ] - 6s 151ms/step - loss: 0.0014
Epoch 74/250
38/38 [============ ] - 5s 139ms/step - loss: 0.0013
Epoch 75/250
38/38 [============ ] - 5s 143ms/step - loss: 0.0014
Epoch 76/250
38/38 [=============== ] - 6s 151ms/step - loss: 0.0016
Epoch 77/250
38/38 [=============== ] - 5s 136ms/step - loss: 0.0016
Epoch 78/250
38/38 [============ ] - 6s 150ms/step - loss: 0.0015
Epoch 79/250
38/38 [============== ] - 6s 153ms/step - loss: 0.0013
Epoch 80/250
38/38 [============== ] - 5s 133ms/step - loss: 0.0018
Epoch 81/250
38/38 [=============== ] - 6s 150ms/step - loss: 0.0014
Epoch 82/250
38/38 [=============== ] - 6s 151ms/step - loss: 0.0015
Epoch 83/250
38/38 [=============== ] - 5s 134ms/step - loss: 0.0014
Epoch 84/250
38/38 [=============== ] - 6s 151ms/step - loss: 0.0014
Epoch 85/250
38/38 [=============== ] - 6s 152ms/step - loss: 0.0014
Epoch 86/250
38/38 [=============== ] - 5s 138ms/step - loss: 0.0014
Epoch 87/250
38/38 [============= ] - 6s 150ms/step - loss: 0.0014
Epoch 88/250
38/38 [============ ] - 6s 147ms/step - loss: 0.0014
```

Fnoch	89/250						
	[=======]	_	5s	136ms/step	_	loss:	0.0013
	90/250		<i>C</i> =	150		1	0 0013
	[======] 91/250	_	68	15UMS/Step	_	loss:	0.0013
38/38	[=====]	-	6s	153ms/step	-	loss:	0.0013
-	92/250 [=======]	_	5 s	131ms/sten	_	1088.	0 0014
	93/250		55	1011113/ 500p		1055.	0.0014
	[======================================	-	6s	153ms/step	-	loss:	0.0013
-	94/250 [======]	_	6s	151ms/step	_	loss:	0.0012
Epoch	95/250						
	[======] 96/250	_	58	13/ms/step	_	loss:	0.0015
38/38	[=====]	-	6s	150ms/step	-	loss:	0.0014
-	97/250	_	68	148ms/sten	_	1099.	0 0012
Epoch	98/250			_			
	[======================================	-	5s	137ms/step	-	loss:	0.0012
-	99/250 [=======]	_	6s	151ms/step	_	loss:	0.0012
-	100/250		_	100 /		7	0 0010
	[======] 101/250	-	5s	129ms/step	-	loss:	0.0013
38/38	[=====]	-	5s	124ms/step	-	loss:	0.0012
	102/250 [========]	_	68	152ms/sten	_	1088.	0 0014
Epoch	103/250						
	[======] 104/250	-	6s	149ms/step	-	loss:	0.0013
	[=======]	_	5s	138ms/step	_	loss:	0.0013
	105/250		<i>C</i> -	150/		1	0 0012
	[======] 106/250	_	68	152MS/Step	_	loss:	0.0013
	[======]	-	6s	145ms/step	-	loss:	0.0012
_	107/250	_	5s	144ms/step	_	loss:	0.0014
Epoch	108/250			_			
	[=======] 109/250	-	6s	151ms/step	-	loss:	0.0013
38/38	[=====]	-	5s	137ms/step	-	loss:	0.0012
	110/250 [========]	_	68	147ms/sten	_	1088.	0 0014
Epoch	111/250			_			
	[======] 112/250	-	6s	150ms/step	-	loss:	0.0012
-	[=======]	_	5s	138ms/step	_	loss:	0.0013
-	113/250		<i>C</i> =	151		1	0 0010
	[======] 114/250	_	bS	151ms/step	_	loss:	0.0012
	[======]	-	6s	154ms/step	-	loss:	0.0013
-	115/250 [=========]	_	5s	130ms/step	_	loss:	0.0011
Epoch	116/250						
	[======] 117/250	-	6s	153ms/step	-	loss:	0.0011
38/38	[=====]	_	6s	154ms/step	_	loss:	0.0012
	118/250 [========]	_	50	131mg/g+an	_	1000	0 0011
Epoch	119/250						
	[=======]	-	6s	150ms/step	-	loss:	0.0011
тЬоси	120/250						

```
38/38 [=============== ] - 6s 147ms/step - loss: 0.0011
Epoch 121/250
38/38 [=============== ] - 5s 136ms/step - loss: 0.0012
Epoch 122/250
38/38 [============= ] - 6s 153ms/step - loss: 0.0013
Epoch 123/250
38/38 [=============== ] - 6s 152ms/step - loss: 0.0011
Epoch 124/250
38/38 [============= ] - 5s 135ms/step - loss: 0.0012
Epoch 125/250
38/38 [=============== ] - 6s 149ms/step - loss: 0.0010
Epoch 126/250
38/38 [============== ] - 6s 149ms/step - loss: 0.0011
Epoch 127/250
38/38 [=============== ] - 5s 136ms/step - loss: 0.0011
Epoch 128/250
38/38 [============= ] - 6s 151ms/step - loss: 0.0011
Epoch 129/250
38/38 [=============== ] - 6s 149ms/step - loss: 0.0012
Epoch 130/250
38/38 [=============== ] - 5s 143ms/step - loss: 0.0011
Epoch 131/250
38/38 [=============== ] - 5s 144ms/step - loss: 0.0012
Epoch 132/250
38/38 [============= ] - 6s 145ms/step - loss: 0.0011
Epoch 133/250
Epoch 134/250
38/38 [============= ] - 6s 152ms/step - loss: 0.0013
Epoch 135/250
38/38 [============== ] - 5s 139ms/step - loss: 0.0011
Epoch 136/250
38/38 [============== ] - 5s 144ms/step - loss: 0.0010
Epoch 137/250
38/38 [============ ] - 6s 150ms/step - loss: 0.0011
Epoch 138/250
38/38 [============= ] - 5s 143ms/step - loss: 0.0010
Epoch 139/250
Epoch 140/250
38/38 [=============== ] - 6s 151ms/step - loss: 0.0010
Epoch 141/250
Epoch 142/250
38/38 [============== ] - 6s 148ms/step - loss: 0.0010
Epoch 143/250
38/38 [============== ] - 6s 149ms/step - loss: 0.0013
Epoch 144/250
38/38 [=============== ] - 5s 143ms/step - loss: 0.0012
Epoch 145/250
38/38 [=============== ] - 6s 152ms/step - loss: 0.0010
Epoch 146/250
38/38 [============== ] - 6s 157ms/step - loss: 0.0011
Epoch 147/250
38/38 [================ ] - 5s 136ms/step - loss: 9.9366e-04
Epoch 148/250
38/38 [=============== ] - 6s 149ms/step - loss: 0.0011
Epoch 149/250
38/38 [=============== ] - 6s 154ms/step - loss: 0.0010
Epoch 150/250
38/38 [=============== ] - 5s 139ms/step - loss: 0.0010
Epoch 151/250
38/38 [============ ] - 6s 152ms/step - loss: 0.0010
```

```
Epoch 152/250
38/38 [============= ] - 6s 151ms/step - loss: 0.0010
Epoch 153/250
Epoch 154/250
38/38 [=============== ] - 6s 151ms/step - loss: 0.0010
Epoch 155/250
38/38 [============= ] - 6s 147ms/step - loss: 0.0011
Epoch 156/250
Epoch 157/250
38/38 [============== ] - 6s 152ms/step - loss: 0.0011
Epoch 158/250
38/38 [=============== ] - 5s 138ms/step - loss: 0.0012
Epoch 159/250
Epoch 160/250
Epoch 161/250
38/38 [============= ] - 5s 136ms/step - loss: 0.0010
Epoch 162/250
38/38 [=============== ] - 5s 139ms/step - loss: 0.0011
Epoch 163/250
38/38 [================ ] - 5s 144ms/step - loss: 9.9739e-04
Epoch 164/250
38/38 [============= ] - 5s 135ms/step - loss: 0.0012
Epoch 165/250
38/38 [============== ] - 6s 152ms/step - loss: 9.9564e-04
Epoch 166/250
Epoch 167/250
38/38 [============== ] - 5s 137ms/step - loss: 0.0010
Epoch 168/250
Epoch 169/250
Epoch 170/250
38/38 [=============== ] - 5s 132ms/step - loss: 9.1730e-04
Epoch 171/250
38/38 [============= ] - 6s 147ms/step - loss: 0.0011
Epoch 172/250
38/38 [============= ] - 5s 131ms/step - loss: 9.2256e-04
Epoch 173/250
38/38 [============= ] - 5s 130ms/step - loss: 0.0010
Epoch 174/250
38/38 [================ ] - 6s 154ms/step - loss: 8.7857e-04
Epoch 175/250
Epoch 176/250
38/38 [=============== ] - 5s 132ms/step - loss: 0.0010
Epoch 177/250
38/38 [============== ] - 6s 146ms/step - loss: 9.6927e-04
Epoch 178/250
38/38 [============= ] - 6s 150ms/step - loss: 9.5806e-04
Epoch 179/250
38/38 [=============== ] - 5s 133ms/step - loss: 0.0010
Epoch 180/250
38/38 [============= ] - 6s 149ms/step - loss: 8.8994e-04
Epoch 181/250
38/38 [============ ] - 6s 152ms/step - loss: 0.0010
Epoch 182/250
38/38 [=============== ] - 5s 133ms/step - loss: 0.0011
Epoch 183/250
```

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Epoch 184/250
38/38 [=============== ] - 6s 147ms/step - loss: 0.0011
Epoch 185/250
38/38 [============= ] - 5s 129ms/step - loss: 0.0011
Epoch 186/250
38/38 [=============== ] - 6s 146ms/step - loss: 0.0010
Epoch 187/250
38/38 [============== ] - 6s 150ms/step - loss: 9.5920e-04
Epoch 188/250
38/38 [=============== ] - 5s 133ms/step - loss: 9.6178e-04
Epoch 189/250
38/38 [============== ] - 6s 151ms/step - loss: 0.0010
Epoch 190/250
38/38 [=============== ] - 6s 151ms/step - loss: 0.0011
Epoch 191/250
38/38 [============== ] - 5s 135ms/step - loss: 9.1960e-04
Epoch 192/250
38/38 [============= ] - 6s 152ms/step - loss: 0.0011
Epoch 193/250
38/38 [================ ] - 5s 144ms/step - loss: 9.3426e-04
Epoch 194/250
Epoch 195/250
38/38 [============== ] - 6s 150ms/step - loss: 9.8969e-04
Epoch 196/250
Epoch 197/250
38/38 [============== ] - 5s 140ms/step - loss: 9.2980e-04
Epoch 198/250
Epoch 199/250
38/38 [================ ] - 5s 136ms/step - loss: 9.4370e-04
Epoch 200/250
38/38 [============= ] - 5s 141ms/step - loss: 9.9796e-04
Epoch 201/250
38/38 [============ ] - 6s 154ms/step - loss: 0.0010
Epoch 202/250
38/38 [============== ] - 5s 139ms/step - loss: 0.0012
Epoch 203/250
38/38 [============== ] - 6s 154ms/step - loss: 0.0010
Epoch 204/250
Epoch 205/250
38/38 [================ ] - 5s 138ms/step - loss: 0.0011
Epoch 206/250
Epoch 207/250
38/38 [=============== ] - 6s 155ms/step - loss: 0.0011
Epoch 208/250
38/38 [============== ] - 5s 142ms/step - loss: 0.0010
Epoch 209/250
38/38 [=============== ] - 6s 150ms/step - loss: 0.0010
Epoch 210/250
38/38 [============= ] - 5s 139ms/step - loss: 9.6151e-04
Epoch 211/250
38/38 [================ ] - 5s 137ms/step - loss: 9.0391e-04
Epoch 212/250
38/38 [============ ] - 6s 152ms/step - loss: 0.0010
Epoch 213/250
38/38 [=============== ] - 6s 149ms/step - loss: 9.4700e-04
Epoch 214/250
```

```
Epoch 215/250
38/38 [============== ] - 6s 151ms/step - loss: 0.0012
Epoch 216/250
Epoch 217/250
38/38 [================ ] - 5s 133ms/step - loss: 9.7451e-04
Epoch 218/250
38/38 [============== ] - 6s 145ms/step - loss: 9.7341e-04
Epoch 219/250
38/38 [=============== ] - 6s 149ms/step - loss: 9.6572e-04
Epoch 220/250
38/38 [================ ] - 5s 131ms/step - loss: 9.8095e-04
Epoch 221/250
38/38 [=============== ] - 6s 153ms/step - loss: 8.7400e-04
Epoch 222/250
Epoch 223/250
Epoch 224/250
38/38 [============= ] - 5s 136ms/step - loss: 9.1128e-04
Epoch 225/250
Epoch 226/250
38/38 [=============== ] - 5s 137ms/step - loss: 0.0011
Epoch 227/250
38/38 [============= ] - 6s 151ms/step - loss: 0.0011
Epoch 228/250
Epoch 229/250
Epoch 230/250
38/38 [=============== ] - 6s 151ms/step - loss: 0.0012
Epoch 231/250
38/38 [================ ] - 6s 145ms/step - loss: 8.8872e-04
Epoch 232/250
Epoch 233/250
38/38 [=============== ] - 5s 133ms/step - loss: 9.4380e-04
Epoch 234/250
38/38 [================= ] - 6s 146ms/step - loss: 9.7277e-04
Epoch 235/250
38/38 [============= ] - 5s 138ms/step - loss: 9.9370e-04
Epoch 236/250
38/38 [============= ] - 6s 152ms/step - loss: 0.0011
Epoch 237/250
38/38 [================ ] - 5s 141ms/step - loss: 9.2497e-04
Epoch 238/250
Epoch 239/250
38/38 [================ ] - 6s 152ms/step - loss: 9.8241e-04
Epoch 240/250
38/38 [============== ] - 5s 141ms/step - loss: 8.9683e-04
Epoch 241/250
38/38 [=========== ] - 5s 143ms/step - loss: 0.0011
Epoch 242/250
Epoch 243/250
38/38 [================== ] - 5s 138ms/step - loss: 9.4539e-04
Epoch 244/250
38/38 [=============== ] - 5s 145ms/step - loss: 9.1624e-04
Epoch 245/250
38/38 [=================== ] - 6s 152ms/step - loss: 9.4371e-04
Epoch 246/250
```

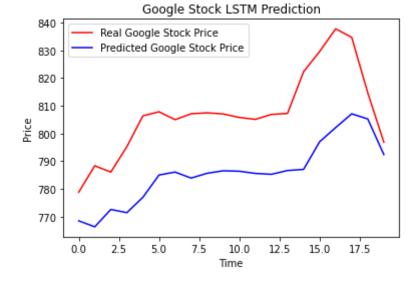
TesLSTM about:srcdoc

```
Epoch 247/250
        Epoch 248/250
        Epoch 249/250
        38/38 [=============== ] - 5s 129ms/step - loss: 0.0010
        <keras.callbacks.History at 0x20128ac3e80>
In [17]:
        regressor.save('model')
        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,
        1stm cell 1 layer call and return conditional losses, 1stm 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: model\assets
        INFO:tensorflow:Assets written to: model\assets
        WARNING:absl:<keras.layers.recurrent.LSTMCell object at 0x0000020119F545B0>
        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 0x000002011B1F2640>
        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 0x0000020119F1DA90>
        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 0x000002011B343880>
        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 [18]:
        # Importing testing set
        dataset test = pd.read csv('Google Stock Price Test.csv')
        real stock set = dataset test.iloc[:, 1:2].values
In [19]:
        # 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 [20]: # visualizing
    plt.plot(real_stock_set, color='red', label='Real Google Stock Price')
    plt.plot(predicted_stock_set, color='blue', label='Predicted Google Stock E
    plt.title('Google Stock LSTM Prediction')
    plt.xlabel('Time')
    plt.ylabel('Price')

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



```
In [ ]:
```