LSTM Time Series Example - Stock price prediction

```
In [1]:
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
In [2]:
tr_data = pd.read_csv('AAPL_data.csv')
In [3]:
tr data proc = tr data.iloc[:, 1:2].values
In [4]:
data dates = tr data.iloc[:, 0].values
Preprocess data
In [5]:
from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler(feature range = (0, 1))
tr data scaled = scaler.fit transform(tr data proc)
tr data scaled.shape
Out[5]:
(1280, 1)
```

```
labels = []
for i in range(60, 1270):
    features_set.append(tr_data_scaled[i-60:i, 0])
    labels.append(tr_data_scaled[i, 0])

In [7]:
features_set, labels = np.array(features_set), np.array(labels)
```

In [6]:

features set = []

```
pe[1], 1))
In [9]:
features set.shape
Out[9]:
(1210, 60, 1)
LSTM Training
In [10]:
from keras.models import Sequential
from keras.layers import Dense
from keras.layers import LSTM
from keras.layers import Dropout
Using TensorFlow backend.
In [11]:
model = Sequential()
model.add(LSTM(units=50, return sequences=True, input shape=(features set.shape[
1], 1)))
model.add(Dropout(0.2))
model.add(LSTM(units=50, return sequences=True))
model.add(Dropout(0.2))
model.add(LSTM(units=50, return sequences=True))
model.add(Dropout(0.2))
model.add(LSTM(units=50))
model.add(Dropout(0.2))
model.add(Dense(units = 1))
model.compile(optimizer = 'adam', loss = 'mean_squared_error')
model.fit(features set, labels, epochs = 100, batch size = 32)
Epoch 1/100
460
Epoch 2/100
047
Epoch 3/100
047
Epoch 4/100
040
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features set = np.reshape(features set, (features set.shape[0], features set.sha

In [8]:

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Epoch 5/100
037
Epoch 6/100
036
Epoch 7/100
036
Epoch 8/100
036
Epoch 9/100
034
Epoch 10/100
033
Epoch 11/100
032
Epoch 12/100
031
Epoch 13/100
030
Epoch 14/100
026
Epoch 15/100
Epoch 16/100
026
Epoch 17/100
026
Epoch 18/100
026
Epoch 19/100
026
Epoch 20/100
028
Epoch 21/100
025
Epoch 22/100
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022
Epoch 23/100
025
Epoch 24/100
025
Epoch 25/100
Epoch 26/100
022
Epoch 27/100
023
Epoch 28/100
023
Epoch 29/100
028
Epoch 30/100
024
Epoch 31/100
021
Epoch 32/100
025
Epoch 33/100
022
Epoch 34/100
020
Epoch 35/100
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Epoch 36/100
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Epoch 37/100
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Epoch 38/100
018
Epoch 39/100
019
Epoch 40/100
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018
Epoch 41/100
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Epoch 42/100
018
Epoch 43/100
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Epoch 44/100
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Epoch 45/100
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Epoch 46/100
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Epoch 47/100
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Epoch 48/100
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Epoch 49/100
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Epoch 50/100
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Epoch 51/100
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Epoch 52/100
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Epoch 53/100
017
Epoch 54/100
015
Epoch 55/100
016
Epoch 56/100
015
Epoch 57/100
014
```

```
Epoch 58/100
017
Epoch 59/100
015
Epoch 60/100
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Epoch 61/100
014
Epoch 62/100
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Epoch 63/100
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Epoch 64/100
015
Epoch 65/100
015
Epoch 66/100
015
Epoch 67/100
016
Epoch 68/100
Epoch 69/100
016
Epoch 70/100
015
Epoch 71/100
013
Epoch 72/100
014
Epoch 73/100
013
Epoch 74/100
013
Epoch 75/100
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013
Epoch 76/100
014
Epoch 77/100
014
Epoch 78/100
Epoch 79/100
1210/1210 [=============== ] - 7s 6ms/step - loss: 0.0
011
Epoch 80/100
013
Epoch 81/100
012
Epoch 82/100
012
Epoch 83/100
012
Epoch 84/100
013
Epoch 85/100
013
Epoch 86/100
011
Epoch 87/100
012
Epoch 88/100
012
Epoch 89/100
011
Epoch 90/100
013
Epoch 91/100
012
Epoch 92/100
012
Epoch 93/100
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```
011
Epoch 94/100
012
Epoch 95/100
011
Epoch 96/100
012
Epoch 97/100
013
Epoch 98/100
010
Epoch 99/100
011
Epoch 100/100
010
Out[11]:
<keras.callbacks.History at 0x127c58f28>
```

LSTM Testing

```
In [12]:

test_data = pd.read_csv('AAPL_testdata.csv')

test_data_proc = test_data.iloc[:, 1:2].values

test_data_dates = test_data.iloc[:, 0].values
```

Preprocess test data

test_inputs = test_inputs.reshape(-1,1)

test inputs = scaler.transform(test inputs)

```
In [13]:
data_all = pd.concat((tr_data['Open'], test_data['Open']), axis=0)
In [14]:
test_inputs = data_all[len(data_all) - len(test_data) - 60:].values
In [15]:
```

```
In [16]:
test features = []
for i in range(60, 80):
    test features.append(test inputs[i-60:i, 0])
In [17]:
test_features = np.array(test_features)
test features = np.reshape(test features, (test features.shape[0], test features
.shape[1], 1))
In [18]:
predictions = model.predict(test features)
In [19]:
predictions = scaler.inverse transform(predictions)
In [20]:
plt.figure(figsize=(10,5))
plt.plot(test data dates[0:20], test data proc[0:20], color='blue', label='Actual
Stock Price')
plt.plot(predictions , color='red', label='Predicted Stock Price')
plt.xticks(rotation=90, size=10)
plt.title('AAPL Stock Price Prediction using LSTM')
plt.xlabel('Date')
plt.ylabel('Stock Price')
plt.legend()
plt.show()
```

AAPL Stock Price Prediction using LSTM

