

BRONX DATA - LSTM - Multi-Step Forecast - Vector Output Model

Here I have done the following:

1. Followed steps from this website: <https://machinelearningmastery.com/how-to-develop-lstm-models-for-time-series-forecasting/> (<https://machinelearningmastery.com/how-to-develop-lstm-models-for-time-series-forecasting/>)
2. Import necessary modules
3. Fixed the parameters of the code to take in input of previous 60 days and output the next 30 days
 - n_steps_in = 60
 - n_steps_out = 30
4. Define the model and predict 30 days of data
5. Note any observations

```
In [2]: # Imports
import numpy as np
from numpy import array
from keras.models import Sequential
from keras.layers import LSTM
from keras.layers import Dense
import matplotlib.pyplot as plt
import pandas as pd
#Supress default INFO Logging
%matplotlib inline
import warnings
warnings.filterwarnings('ignore')
import logging
logger = logging.getLogger()
logger.setLevel(logging.CRITICAL)
import logging, sys
warnings.simplefilter(action='ignore', category=FutureWarning)

In [3]: df = pd.read_csv('datasets/rollingsales_bronx.xls_prepped_bare.csv', usecols=['S',

In [4]: df = df.dropna()
df = df.reset_index(drop=True)

In [5]: df = df.rename(columns={'SALE DATE': 'ts', 'SALE PRICE': 'y'})
df.columns = df.columns.astype(str)
df = df.set_index(['ts'], drop=True)
df.index= pd.to_datetime(df.index)

In [6]: # df

In [7]: df = df.resample('D').mean()
df = df.reset_index()
```

In [8]:

df.dropna(inplace=True)
df

Out[8]:

	ts	y
0	2020-04-01	1.234333e+06
1	2020-04-02	5.502250e+05
2	2020-04-03	6.185000e+05
5	2020-04-06	5.919458e+05
6	2020-04-07	3.577680e+05
...
358	2021-03-25	5.115159e+05
359	2021-03-26	1.018866e+06
362	2021-03-29	4.119000e+05
363	2021-03-30	5.161438e+05
364	2021-03-31	5.053775e+05

265 rows × 2 columns

In [9]:

raw_input_test = list(df['y'])
raw_input_test
np.shape(df.index)

Out[9]:

(265,)

Below steps are taken from:

<https://machinelearningmastery.com/how-to-develop-lstm-models-for-time-series-forecasting/>
(<https://machinelearningmastery.com/how-to-develop-lstm-models-for-time-series-forecasting/>)

```
In [10]: # split a univariate sequence into samples
def split_sequence(sequence, n_steps_in, n_steps_out):
    X, y = list(), list()
    for i in range(len(sequence)):
        # find the end of this pattern
        end_ix = i + n_steps_in
        out_end_ix = end_ix + n_steps_out
        # check if we are beyond the sequence
        if out_end_ix > len(sequence):
            break
        # gather input and output parts of the pattern
        seq_x, seq_y = sequence[i:end_ix], sequence[end_ix:out_end_ix]
        X.append(seq_x)
        y.append(seq_y)
    return array(X), array(y)

# define input sequence
raw_seq = raw_input_test

# choose a number of time steps
n_steps_in, n_steps_out = 60, 30

# split into samples
X, y = split_sequence(raw_seq, n_steps_in, n_steps_out)

# reshape from [samples, timesteps] into [samples, timesteps, features]
n_features = 1
X = X.reshape((X.shape[0], X.shape[1], n_features))

# define model
model = Sequential()
model.add(LSTM(100, activation='relu', return_sequences=True, input_shape=(n_steps_in, n_features)))
model.add(LSTM(100, activation='relu'))
model.add(Dense(n_steps_out))
model.compile(optimizer='adam', loss='mse')
# fit model
model.fit(X, y, epochs=100, verbose=0)
```

Out[10]: <tensorflow.python.keras.callbacks.History at 0x19e24d7ec40>

```
In [12]: # demonstrate prediction

x_input = array(raw_input_test[205:265])
x_input = x_input.reshape((1, n_steps_in, n_features))
yhat = model.predict(x_input, verbose=0)
print(yhat)
```

```
[[1055523.1    669216.4    658051.8    960284.1    1299690.5
  601752.44   406897.84   815741.9    578513.4     3591.9185
  223313.23   386154.72   881171.75   329858.44   455366.72
 -420783.06  -40041.72   -234504.58   197148.53   148469.48
  207373.     292068.4    -91707.44   259172.25   477038.94
  419743.     662983.1    119253.05   1154972.8    -2869.8867]]
```

```
In [13]: np.shape(list(yhat))
```

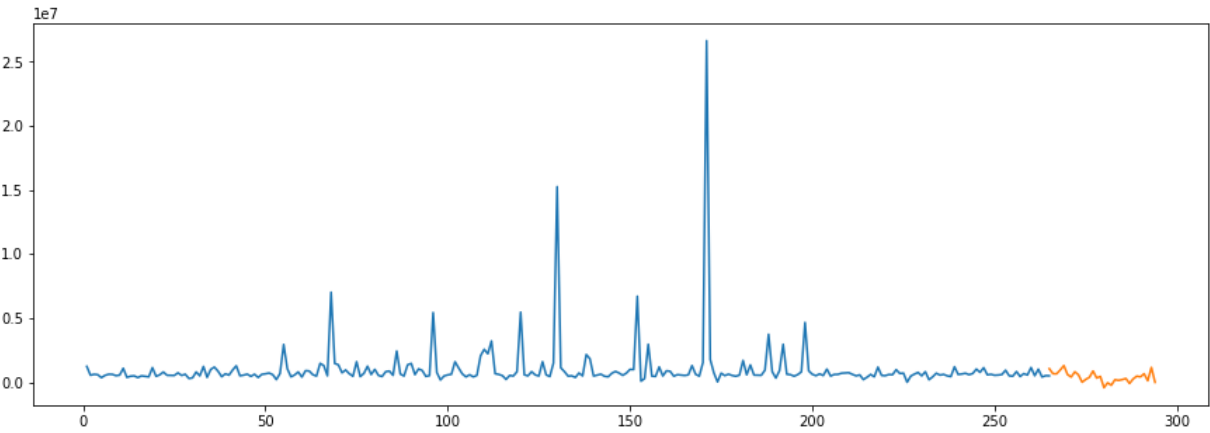
Out[13]: (1, 30)

```
In [14]: y_hat1 = np.reshape(yhat, (30,1))
np.shape(y_hat1)
```

Out[14]: (30, 1)

```
In [15]: # I increased the epochs and the predictions went higher.
x_list = list(range(1,295))
```

```
In [17]: plt.figure(figsize=(15,5))
fig =plt.plot(x_list[0:265], df['y'][0:265])
ax = plt.plot(x_list[264:295], y_hat1)
```



Observation

Bronx prices per model show downward prices with some dips