

Use case

Task of the work – make neural network to predict stock prices as far as it's possible.

Data set

As a dataset I am using daily prices from several years grabbed from Yahoo Finance web site. For this task I am using ready-made python library, but it is only BeautifulSoup inside.

As a result, I am getting data set with bunch of prices for selected ticker.

Data exploration, visualization and Quality assessment.

We have a data set with Date, Open, High, Low and Close prices, Adjusted close price and daily volume of trade

| | Date | Open | High | Low | Close | Adj Close | Volume |
|------|------------|-------------|-------------|-------------|-------------|-------------|-----------|
| 0 | 2010-01-04 | 951.719971 | 964.729980 | 951.719971 | 963.559998 | 963.559998 | 82515400 |
| 1 | 2010-01-05 | 964.030029 | 968.679993 | 961.460022 | 967.270020 | 967.270020 | 62738400 |
| 2 | 2010-01-07 | 967.390015 | 970.260010 | 962.270020 | 965.820007 | 965.820007 | 132590900 |
| 3 | 2010-01-08 | 965.700012 | 974.900024 | 965.700012 | 973.440002 | 973.440002 | 108703800 |
| 4 | 2010-01-11 | 973.150024 | 985.690002 | 973.150024 | 978.179993 | 978.179993 | 108829800 |
| ... | ... | ... | ... | ... | ... | ... | ... |
| 2612 | 2020-05-29 | 1642.130005 | 1645.849976 | 1625.699951 | 1629.760010 | 1629.760010 | 210944400 |
| 2613 | 2020-06-01 | 1650.989990 | 1651.280029 | 1633.060059 | 1649.380005 | 1649.380005 | 81839400 |
| 2614 | 2020-06-02 | 1657.689941 | 1678.729980 | 1654.209961 | 1666.979980 | 1666.979980 | 112254900 |
| 2615 | 2020-06-03 | 1684.280029 | 1709.959961 | 1682.520020 | 1709.959961 | 1709.959961 | 123035700 |
| 2616 | 2020-06-04 | 1703.000000 | 1708.630005 | 1690.069946 | 1700.479980 | 1700.479980 | 105834400 |

We can plot a graph.

We need only date and close price. Rest of prices can be dropped.

There are two possible issues in that dataframe.

1) Sometimes we have "null" value (spelled exactly like this). I am just dropping such rows, because it is only few cases it will not affect training or testing neural network.

2) Sometimes Yahoo adds dividend data like here

| Currency in USD | | | | | | Download |
|-----------------|-----------------|-------|-------|--------|-------------|--------------------------|
| Date | Open | High | Low | Close* | Adj Close** | Volume |
| Sep 01, 2016 | 26.53 | 29.05 | 25.63 | 28.26 | 26.55 | 3,872,062,400 |
| Aug 04, 2016 | 0.1425 Dividend | | | | | |
| Aug 01, 2016 | 26.10 | 27.56 | 26.00 | 26.52 | 24.78 | 2,520,514,000 |
| Jul 01, 2016 | 23.87 | 26.14 | 23.59 | 26.05 | 24.34 | 2,743,118,400 |
| Jun 01, 2016 | 24.75 | 25.47 | 22.88 | 23.90 | 22.33 | 3,117,990,800 |
| May 05, 2016 | 0.1425 Dividend | | | | | |
| May 01, 2016 | 23.48 | 25.18 | 23.37 | 24.07 | 23.18 | 3,603,686,000 |

On the screenshot we have monthly data, but issue with daily is exactly the same. Such rows shall be deleted.

All columns are strings, so data needs to be converted to datetime and float64 formats

Data is normalized and divided between training and test set (approx.. 90% and 10%)

Model and metrics selection.

Stock price is a process developed in time, but do not really depending on time. There is no direct correlation between time and price, so most of simple Machine Learning techniques won't work.

To solve the task, I will use an LSTM neural network.

After few experiments I came to an architecture:

Sequence model, 3 LSTM layers, 3 Dropout layers between them and Dense layer at the end.

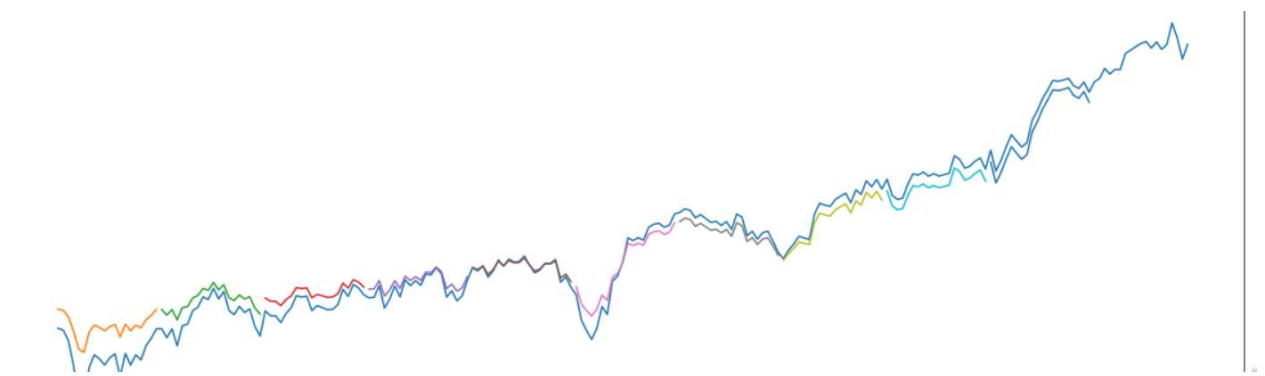
MSE loss function will be use.

To measure prediction performance, we can use any "distance" metrics, but it is better to just build a plot.

Testing

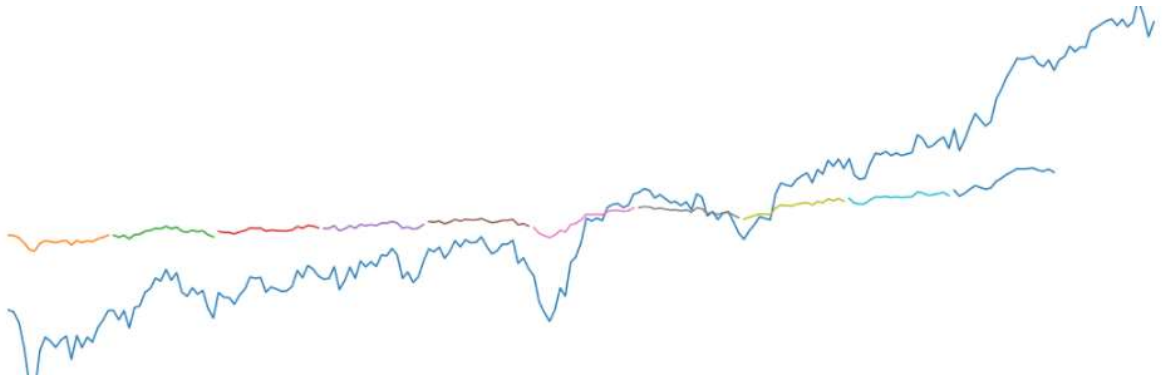
To test model I am using completely new set of data (1 year) which directly follows after training set

Predicted prices are presented as short color lines on a picture. Experiments showed, that we have a good prediction up to 25-30 steps forward. After that, we have much worse result



This graph is a result of long feature engineering. For example SGD optimizer gives us very bad performance after model training.

If you will reduce training set two times result will be like on the picture below.



Reducing dropout rate to 0.05 improves performance, but model is still underperforming

