Stock Prediction Using Recurrent Neural Network

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Outline

- Introduction
 - What is RNN
 - How does LSTM work
- Stock prediction with LSTM
 - Goal
 - Data
 - Architecture
 - Results
- Conclusion

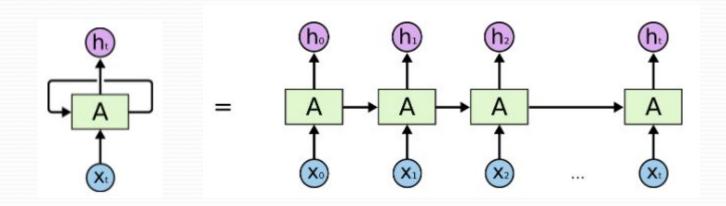


Reference

- Tesorflow documentation
- "Understanding LSTM Networks", Christopher,Olah, http://colah.github.io/posts/2015-08-Understanding-LSTM s/ (I use a lot of diagrams from this blog in this presentation)
- "Predict Stock Prices Using RNN", Lilian Weng, <u>https://lilianweng.github.io/lil-log/2017/07/08/predict-stock</u> <u>-prices-using-RNN-part-1.html</u>
- "Deep learning with long short-term memory networks for financial market", Thomas Fischer and Christopher Krauss,
 - https://www.iwf.rw.fau.de/files/2015/12/11-2017.pdf



What is RNN (Recurrent NN)



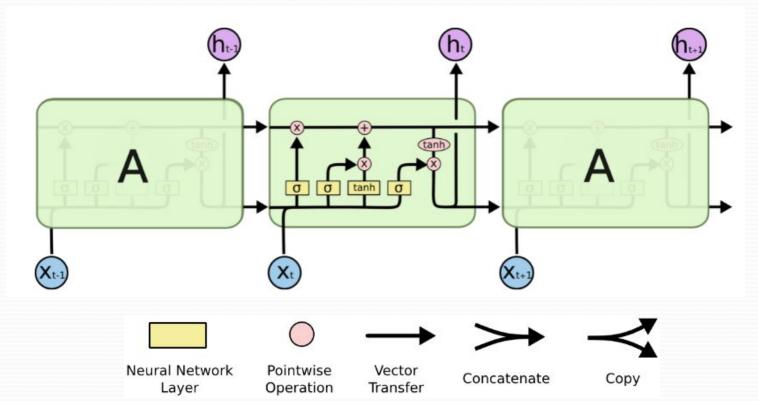
With loops

Unrolled RNN



What is LSTM

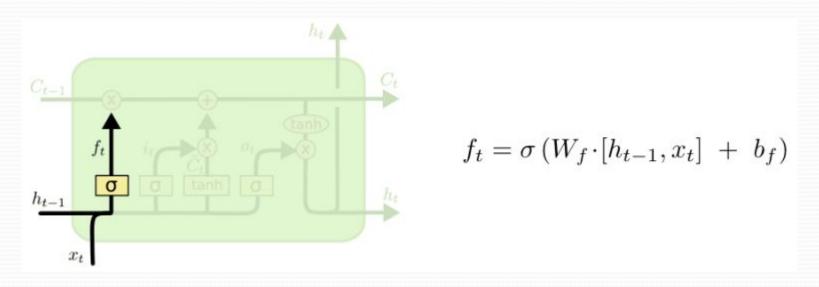
- Long Short-Term Memory is a special type of RNN
- Capable of learning long-term dependencies





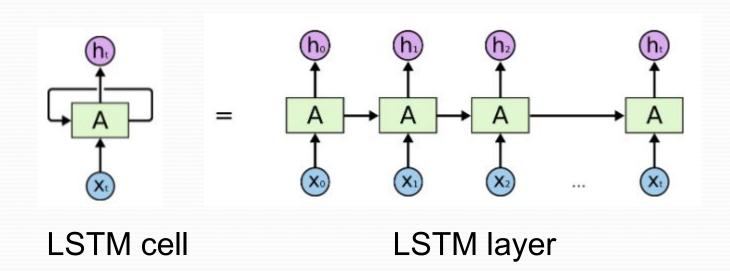
Closer look at part of LSTM

- Gate that controls how much information to keep
- Gates are defined by weights and biases that form the trainable variables



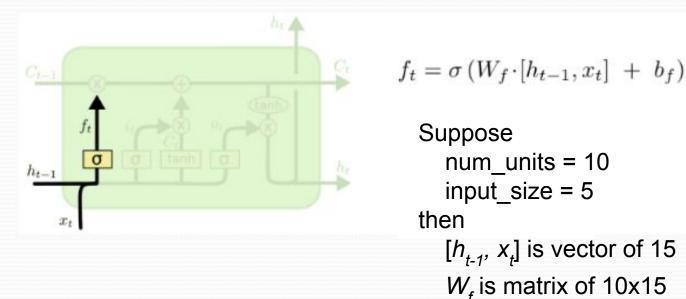
Confusing concepts and notations

- num_units in tf.contrib.rnn.LSTMCell
- num_steps: number of repeats in the unrolled version
- input_size: length of X_t (X_t is a vector)



Confusing concepts and notations

- num_units in tf.contrib.rnn.LSTMCell
 - size of lstm cell
 - output size





 b_f is vector of 10

 f_{t} is vector of 10

Hands-on

- Code
 - main.py: the main program
 - loadData.py: for data formatting
 - IstmRNN.py: for building and training LSTM RNN
- Data
 - data/SP500.csv
- Trained model
 - "model" subfolder



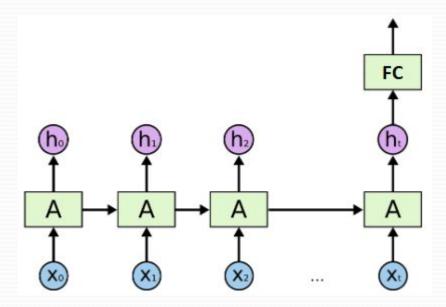
Some Tensorflow APIs

- Create an LSTM cell: tf.contrib.rnn.LSTMCell(num_units, ...)
- Wrap a cell with Dropout: tf.contrib.rnn.DropoutWrapper(cell, output_keep_prob, ...)
- Construct an RNN:
 tf.nn.dynamic_rnn(cell, inputs, ...)



Apply LSTM to stock prediction

- Goal: to demonstrate the use of LSTM, not to develop a winning model
- Data: S&P 500 index, downloaded from Yahoo
- Architecture





Data Loading/Formatting

- $[p_1, p_2, ..., p_n]$, n = 17292, closing price 1950-2018
- $[\tilde{p}_1, \tilde{p}_2, \dots, \tilde{p}_n], \tilde{p}_i = \frac{p_i p_{i-1}}{p_{i-1}}$
- Samples (num_steps = 100)

Data are split into two sets as training and evaluation



Conclusion

- "Random walk" theory
- What published works (by Krauss) tell us
- Want to do serious work?
 - Kaggle competition:
 - https://www.kaggle.com/c/two-sigma-financial-news
 - Quantopian: https://www.quantopian.com/

