

Large Scale Log Analytics through ELK

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July 5, 2018

Introduction

Problem statement

- Events
- Centralized location
- ELK-stack
 - Elastic Search
 - Logstash
 - Kibana
- Lambda Architecture
 - Batch layer (raw data sets and pre-compute Batch views)
 - Speed layer (Real-time views)
 - Serving Layer (ad-hoc queries)

According to the introduction, the following research question is defined as:

- *Which way of applying Artificial Intelligence in the form of Machine Learning/Deep learning can be used to find relevant actionable information from event data suitable for use within the batch layer in a Lambda Architecture?*

This main research question is divided into the following sub questions:

- Which Deep Learning (DL) model fits for this type of data?
- How is the accuracy of the DL model calculated?
- How can the model be tuned during operations?

Introduction

Approach

- Keras Deep Learning API
 - High-level implementation of Artificial Neural Networks (ANN) written python
 - Run on top of Tensorflow, Theano or CNTK
- Define layers
 - Sequence layer
 - Function layer
- Compile model
 - How to learn - Optimizing Stochastic Gradient Decent algorithms?
 - SGD
 - RMSprop
 - Adagrad
 - Adadelta
 - Adam

- Scherer, R. Lstm recurrent neural networks for short text and sentiment classification. 2017
- Ruder, S. An overview of gradient descent optimization algorithms. 2018

Technical background

Recurrent Artificial Neural Network and Artificial Neural Network

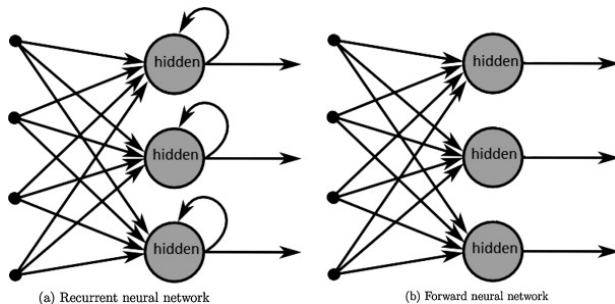


Figure: Recurrent vs Forward neural network; *Source:* [1]

Technical background

Long Short-Term Memory

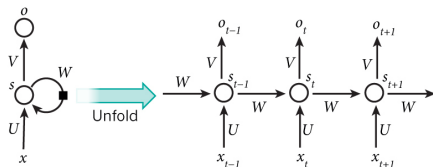


Figure: RNN sequence; *Source:* [2]

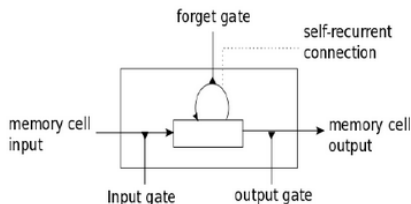
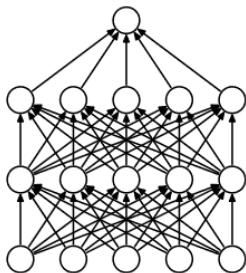


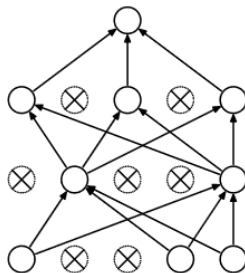
Figure: LSTM memory cell; *Source:* [3]

Technical background

Dense layer vs. Dropout layer



(a) Standard Neural Net



(b) After applying dropout.

Figure: Dense vs Dropout layer; *Source:* [4]

Technical background

Cross-entropy, Epochs, Batch size and Softmax

- Categorical Cross-entropy
 - Loss Function
 - Indicates distance between what the model the output distribution should be and what the original distribution really is
- Epoch
 - Entire data set went through the algorithm in number of times
- Batch size
 - The amount of test/training examples in one forward or backward pass
 - RAM memory
- Softmax function
 - Calculate the probability distribution of each target class over n different/possible target classes for the given inputs

Proposed implementation

PoC

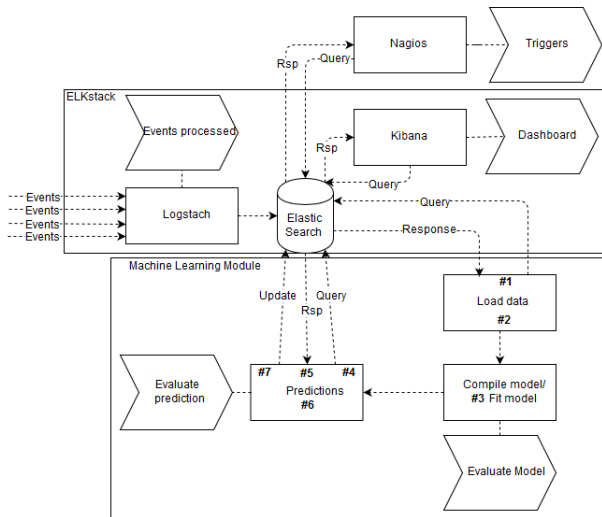


Figure: Architecture

Proposed implementation

PoC layers

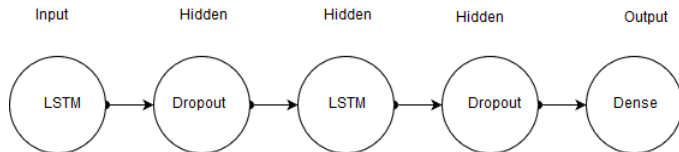
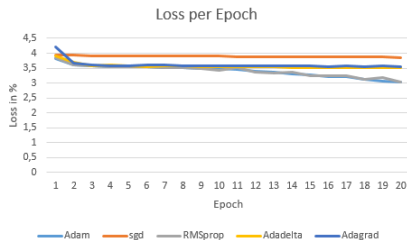
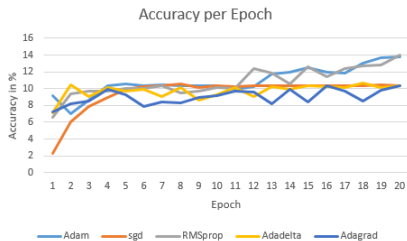


Figure: AI layers

- Advantages of this model
 - Interfaces may be changed easily
 - Change AI model easily
 - Change Algorithms easily

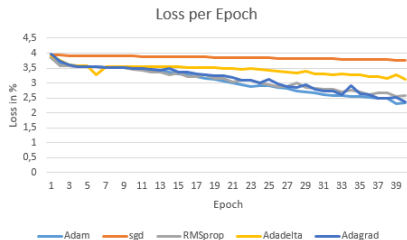
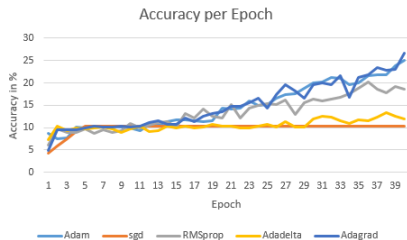
Results

20 Epochs



Results

40 Epochs



	20 Epochs		40 Epochs	
	Loss	Accuracy	Loss	Accuracy
Adam	2.96977%	16.36363%	2.21985%	29.81818%
SGD	3.85677%	10.39999%	3.75910%	10.39999%
RMSprop	3.00070%	5.127276%	2.73776%	19.41818%
Adadelta	3.44532%	10.54545%	3.11788%	13.81818%
Adagrad	3.52502%	10.39999%	2.28705%	27.27272%

Table: Total accuracy and loss per model as a whole

Discussion & Conclusion

- Discussion
 - Adam is the best choice according to the results
 - Does not guarantee the correctness.
 - Model needs a lot of data and time in order to get a better accuracy and learning rate
- Conclusion
 - Apply LSTM with Adam as optimizer
 - Softmax
 - Inject false or true events, in such a way the accuracy will change

Future work

- Apply AI on Speed layer
- Optimize AI module
- investigate better approach for accuracy and predictions.
- What should be the time-period to analyze to optimize relevant context but also trying to be as near-realtime as possible?

Questions?

References

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<https://medium.com/@camrongodbout/recurrent-neural-networks-for-beginners-7aca4e933b82#.q5otnbm2i>, year = 2016.

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