

Deep Structured Energy Based Model for Anomaly Detection

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1 Introduction

We worked on the fully connected energy based model as stated in the paper of Shuangfei Zhai et al [1]. The paper presents three deep structured energy based models: fully connected Energy Based Model (used for static data), Recurrent Energy Based Model (used for sequential data) and convolutional Energy Based Model (used for spatial data). The goal is to detect anomalies in the data. Energy based models are probabilistic models that can be used to build probability density functions. We can see a fully connected Energy Based Model as a stack of Restricted Boltzmann Machines and trained it with a stochastic gradient descent.

2 Implementation

We use the set of examples KDD99¹. Our work is available at <https://github.com/NicolasDerumigny/ML-Proj>. We have worked with TensorFlow and it was really hard to get in, probably because we have never used it before. We have also had some trouble about the subject which was very specific and not clear in regard of our neural network knowledge, more of that some mistakes are present in the paper and it was very difficult to correct them (like equation 9 in the paper).

We have implemented a two layer fully connected energy based model and trained it with a stochastic gradient descent. For that we have modified the KDD99 set in order to obtain numerical features. Only 'good' examples have been used to train the model.

3 Exploitation

¹available at <https://archive.ics.uci.edu/ml/datasets/KDD+Cup+1999+Data>

References

- [1] Shuangfei Zhai, Yu Cheng, Weining Lu, and Zhongfei Zhang. Deep structured energy based models for anomaly detection. In Maria-Florina Balcan and Kilian Q. Weinberger, editors, *Proceedings of the 33rd International Conference on Machine Learning, ICML 2016, New York City, NY, USA, June 19-24, 2016*, volume 48 of *JMLR Workshop and Conference Proceedings*, pages 1100–1109. JMLR.org, 2016.