

Computer Science Tripos - Part II Project

Deep Learning Techniques for Credit Card Fraud Detection

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Proforma

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Detection

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Project Originators: H. Graham & B. Dimanov Supervisors: Dr M. Jamnik & B. Dimanov

Original Aims of the Project

The primary aim of the project was to implement and compare some deep learning techniques, alongside some baseline models, for credit card fraud detection (CCFD). More specifically, I aimed to experiment with two popular types of architecture, namely Convolutional Neural Networks (CNNs) [1] and Generative Adversarial Networks (GANs). These have been successful in the image classification space and the aim of this project was to shed light on their use in the credit card fraud space. This kind of experimentation of predominantly image-based models, on single dimensional, time-series data is a relatively novel approach for CCFD.

Work Completed

All of the core project aims set out in the proposal have been met, meaning results have been collated and evaluated across the three main components of the project: Baseline Models, CNN methods and GAN methods. I have also gone on to do some extension work relating to further investigation on the models I have experimented with. This is in the form of parameter tuning and further analysis not originally set out in the project proposal.

Special Difficulties

None.

Declaration

I, Harry Graham of Christ's College, being a cand	didate for Part II of the Computer
Science Tripos, hereby declare that this dissertation a	and the work described in it are my
own work, unaided except as may be specified below	, and that the dissertation does no
contain material that has already been used to any s	substantial extent for a comparable
purpose.	
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Introduction

Credit card fraud is a globally significant and increasing problem. According to the Nilson Report [2], annual global fraud losses reached \$22.80 billion in 2016, up 4.4% over 2015. Machine learning has contributed a lot to this problem over the years, helping to automatically learn to classify fraudulent transactions. However, this is still somewhat tedious and clearly, the money lost due to fraud is not decreasing. Not to mention, we still have this difficult business decision of when to draw the cutoff points between classifying fraud but perhaps allowing more benign transactions to be blocked.

A lot of machine learning concepts have been around for decades but ongoing research into deep learning architectures and their applications, makes for an interesting experimentation space. In this project I explore the performance of some particular models, focusing on deep learning, applied to the particular problem of credit card fraud detection (CCFD).

In particular, the aim is to shed light on the use of architectures that have had success in the image classification/generation space, in the context of non-image data i.e transactional vectors and time series data. This is something that has recently seen some success [3] and is novel to credit card fraud data. I first explore a set of baseline classifiers, which are primarily a handful of out-of-the-box supervised learning classifiers such as Random Forest. The point of these is to set the scene for experimenting with the data and to see what can be achieved with what is easily available, in other words without any 'deep' learning components. Here, I also establish techniques and methods for processing and evaluating the data i.e cross-validation, datapoint scaling, and data visualisation.

Then the project shifts to experimenting with Convolutional Neural Networks (CNNs) [1] and Generative Adversarial Networks (GANs) [4].

Preparation

Implementation

Evaluation

Conclusions

Bibliography

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