Pluto: A Deep Learning based Watchdog for Anti Money Laundering

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Abstract
Banks are faced with Anti-Money Laundering (AML) obligations so they have to identify customers by conducting negative news, aka “adverse media”, screening which consists of searching information in the public domain for news items, publications, and government advisories and bulletins for information related an individual or entity’s involvement in financial crime matters. Although it is an essential way to determine who in the group poses a higher risk for potential financial crime concerns by catching sophisticated activities from negative news across globe, it also requires heavy human capital and processing time on screening daily produced negative news. Therefore, poor efficiency becomes the most unacceptable obstacle based on this approach. To mitigate this issue, Pluto¹ offers a distributed and scalable batch system embedded deep learning-based Natural Language Processing (NLP) techniques for AML practitioners to improve daily task efficiency. It performs text preprocessing, paragraph embeddings, and clustering algorithm on a set of negative news and provide clustering result with keywords and similarities for AML practitioners. The overall feedback from AML practitioners are very positive on such an impressive enhancement in which Pluto reduces 67% efforts of negative news screening.

1 Introduction
Compliance with Anti-Money Laundering (AML) and Know Your Customer (KYC) duties require identity checks against potential money launderers, terrorists, or people considered high risk in the financial field. Although the concept is reasonable and feasible, there are tremendous challenges on insufficient approaches capable of offering an effective and efficient solution. It triggers lots of techniques been proposed in real world industry [Han et al., 2018].

We leverages GlobalVision’s Patriot Officer² to discover whether customers are high-risk individuals or not via check-

¹A cartoon dog created in 1930 at Walt Disney Productions.
²http://www.gv-systems.com/products-solutions/patriot-officer/

1 Design and Implementation
In this section, we introduce the details on Pluto architecture and method based on NLP and fundamental algorithms. It only supports the articles published in Traditional Chinese (zh-TW) at this stage.

2.1 Architecture
The client-server model batch processing system follows a micro-service oriented distributed architecture [Fehling et al., 2014] presented in Figure 1.
The API service provides REST endpoints for the interaction between the UI and the Patriot Officer. It triggers processing pipelines by sending clustering job into AMQP when receiving requests from Patriot Officer and provides the clustering result information for UI from DB. The Workers service fetches jobs from AMQP to proceed pipeline for document clustering based on proposed solutions. The UI allows

²AMQP is an open standard application layer protocol used for queuing and routing messages between the services in a secure and reliable way
practitioners to retrieve clustering information related to the specific watchlist.

The batch processing system is scalable. The quantity of Workers can be scale up and down according to quantity of jobs residing in AMQP, which makes the system achieve throughput guarantee.

### 2.2 Method

The pipeline comprises three sequential components: text preprocessing, paragraph embeddings construction, and clustering. Input is a set of documents associated to a single watchlist, and the output is a set of clusters containing all documents and the similarities within a single cluster.

Initially, we pick up tokens that match keywords offered by AML practitioners or be tagged as human names and proprietary nouns such as political groups, administrative unit, facilities, locations, organizations, and etc. Sinica CKIP tokenizer and POS tagger [Ma and Chen, 2003] are the major NLP techniques on this topic. The extracted tokens are regarded as representative for each document.

Next, we build the vector representation by Distributed Bag of Words version of Paragraph Vector (PV-DBOW) model [Le and Mikolov, 2014] which not only being conceptually simple but also requires to store less data. Moreover, it ignores the context words, which aligns the empirical insights from AML practitioner. The adopted deep learning-based techniques revealed from [Mikolov et al., 2013a] and [Mikolov et al., 2013b] is implemented by Gensim4 for this stage.

Eventually, we infer vectors for documents based on paragraph embeddings and apply Balanced Iterative Reducing and Clustering Using Hierarchies (BIRCH) [Zhang et al., 1996] to proceed clustering. The reason why we adopt it is that BIRCH, one of the developments in hierarchical clustering, does not require us to pre-specify the number of clusters and is a memory-efficient and time-efficient approach. [Xu and Tian, 2015]

### 3 Evaluation

To show the effect of Pluto, we cooperate with professionals of AML to quantitate the performance via user testing. We divide users into 2 groups: (1) reading news without clustered news and (2) reading news with Pluto.

For Pluto users, we implement a web interface to show the performance difference. As shown in Figure 2, the original system provides only an URL list, while Pluto visualizes negative news into clusters with keywords and similarities. With the interface, users can comprehend the news efficiently, and determine the suspect customer fast.

To evaluate the performance, both groups read the same news set contains 3000 news. As a result, Pluto reduces 67% time in average to judge if the customer suspected of monen laundering.

### 4 Conclusion and Future Work

In this work, we propose a distributed architecture batch system based on NLP techniques to organize daily negative news and offer an UI for information visualization. At present, the entire system is at piloting stage in our private cloud and facilitates efficient work flow among AML investigations pipeline.

In the future, we will (1) adopt our system to multilingual use cases, especially including Simplified Chinese (zh-CN) and English (en). (2) utilize NLP techniques in further investigation in which we may embrace named entity recognition (NER) and relation extraction (RE) to build the relation network identifying target suspicious entities, events, and time, and location.
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