Mining Big Data and Streams

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INTRODUCTION

Mining big data is getting a lot of attention currently because the businesses need more complex information in order to increase their revenue and gain competitive advantage. The growing of the telecommunication data traffic according to Cisco annual forecasting will reach 8.6 zettabytes by the end of 2018 up from 3.1 zettabytes per year in 2013 (Cisco analysis, 2014). Therefore, mining the huge amount of data as well as mining real-time data needs to be done by new data mining techniques/approaches. Big Data is a new term used to identify the datasets that are of large size and have greater complexity (Bifet, 2013). Data mining (DM) is the process of searching large volumes of data automatically for patterns such as association rules (Gupta et al., 2014). Big data mining is defined as the capability of extracting valuable information from large datasets or streams of data that due to its characteristics it is not possible before to do it (Fan & Bifet, 2013). This chapter will discuss the challenges of big data, new data mining techniques compared with traditional techniques and the main DM tools for handling very large datasets. Moreover, the chapter will focus on two industrial areas telecommunications and healthcare and lessons learned from them.

BACKGROUND

The data produced nowadays is estimated in the order of zettabytes, and it is growing around 40% every year. We need new algorithms, and new tools to deal with all of these data. Therefore, the use of big data is becoming a crucial way for leading companies. For example, in healthcare, data pioneers are analyzing the health outcomes of pharmaceuticals when they were widely prescribed, and discovering benefits and risks that were not evident during necessarily more limited clinical trials (McGuire, 2012). We selected some significant articles that discussed challenges, techniques and tools for mining big data. Yadav et al. (2013) presented a review of several algorithms from 1994-2013 necessary for handling big data set. It gives an overview of architecture and algorithms used in large data sets, various tools that were developed for analyzing them as well as various security issues and trends. Bifet (2013) discussed data stream mining and how it offers many challenges and also many opportunities. Che et al. (2013) presented an overview of mining big data and its challenges include heterogeneity, scalability, speed, accuracy, trust, provenance and privacy. This paper also provides an overview of the platforms for processing and managing big data as well as platforms and libraries for mining big data. Jovic et al. (2014) discussed several data mining tools including RapidMiner, R, Weka, KNIME, Orange, and scikit-learn. Fan and Bifet (2013) presented big data challenges, applications of mining big data, Apache Hadoop and other open sources for big data mining and big graphic mining. Singh (2014) discussed machine learning techniques to capturing the value hidden in big data. He presented supervised learning using neural networks, Support Vector Machines (SVMs) and Naive Bayes classifiers, and also unsupervised learning using k-Means, hierarchical clustering and self-organizing maps.
CHALLENGES OF BIG DATA SYSTEMS

Big data has five key elements: Volume, Velocity, Variety, Veracity and value. These 5 V’s are considered challenges of Big Data systems (Yin & Kaynak, 2015; Ishwarappa & Anuradha, 2015; Marr, 2015).

Volume refers to the huge amount of data. Many companies have large archived data in the form of logs but do not have the capacity to manipulate and analyze that data using traditional database technology. Now big data technology can help store and use these datasets in order to gain benefits from them.

Velocity represents the speed at which data generated and the speed at which data moves around. The speed at which credit card transactions is checked for fraudulent activities and the social media messages going viral in seconds. Thus, big data technology can be used to analyze the data while it is being generated without putting it into databases.

Variety means different data types or format. Traditional database can store and process structured data that fit into tables such as financial data. Now 90% of data generated is in unstructured form and it cannot easily be put into relational databases such as photos, video sequences or social media updates. Big data technology can now harness various types of data like messages, photos, sensor data, and social media conversations and bring them together with more structured data.

Veracity refers to the trustworthiness of the data. The quality and accuracy of big data are less controllable because there will be dirty data. For instance, twitter posts with hash-tags, abbreviations, typos and colloquial speech. Big data analytics now allows us to work with these types of data. The volume, variety and velocity of data often make up for the lack of quality or accuracy.

Value refers to the ability to turn big data into value. Value is the most important aspects of big data because implement IT infrastructure systems are very costly to store big data and businesses are going to require a return on investment. Big data can deliver value in almost any area of business or society such as improving healthcare and better understanding and serving customers.

There are other important challenges in big data management and analytics such as analytics architecture and hidden big data (Singh, 2014). Some key issues like accuracy and privacy are also very critical in mining big data (Che et al., 2013).

DATA MINING TECHNIQUES FOR LARGE SCALE DATA

The challenges in handling big data include capturing, storage, analysis, sharing, visualizing and more. In addition to connection and correlation of data which describes more about relationship among the data. Therefore mining big data needs new architecture, algorithms, techniques for its implementation. This section is focusing on the data mining techniques/methods that can be used for handling big data. These techniques/methods are classified classification, clustering, association rules, time series and data streams as follows:

Classification

One of the important classification techniques is decision tree. Decision tree learning is fast and accurate. Building Decision trees from large datasets requires long time for processing all the training instances and the available memory may be not sufficient for storing the whole training set. Implementation of traditional algorithms for building Decision trees becomes very time consuming. Therefore, many incremental algorithms are available such as BOAT (optimistic decision tree construction), ICE (implication counterexamples) and VFDT (very fast decision tree) for handling large datasets (Franco-Arcega et al., 2013). The hybrid approach combining both decision tree and genetic algorithm are also used to create optimized decision tree in order to improve classification performance (Yadav et al., 2013). Another clas-
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