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# A Holistic Analysis of Cloud Based Big Data Mining

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ABSTRACT This paper presents a holistic analysis of the main challenges of exploring big data that has recently emerged especially due to the application of new digital technology such as social networking systems, electronic business applications among others. These applications are generated huge amounts of data that require the adaptation of the existing data mining approaches and systems. The paper describes the background of big data and cloud computing based on the latest approaches in these areas. The opportunities provided by cloud computing paradigm to enable the development of big data mining applications, business intelligence and analytics are dealt with. The main contribution of this paper is a theoretical definition of the possibilities of using cloud computing for the deployment of business intelligence applications using big data mining and analytics.

*Keywords:* big data mining, cloud computing, software-as-a-service, data/information-as -a service, analytics-as-a service

### Introduction

We live in an era of big data that has embedded a huge potential and increased information complexity, risks and insecurity as well as information overload and irrelevance. Also business intelligence and analytics are important in dealing with data driven problems and solutions in the contemporary society and economy. Analysts, computer scientists, economists, mathematicians, political scientists, sociologists, and other scholars are clamouring for access to the massive quantities of data in order to extract meaningful information and knowledge. Very large data sets are generated by and about organisations, people, and their collaboration and interactions in the digital business ecosystems. For example, the connected devices such as smartphones, RFID readers, webcams, and sensor networks add a huge number of autonomous data sources. Scholars argue about the potential benefits, limitations, and risks of accessing and analysing huge amounts of data such as financial data, genetic sequences, social media interactions, medical records, phone/email logs, government records, and other digital traces generated by people and organisations.

With the development of internet communication and collaboration, data is playing a central and crucial role. Currently data intensive applications are developed and used. Also applications such as the Google+, Twitter, LinkedIn and Facebook are generating massive of data. Generally, data intensive applications including eBay, Amazon store and process data in a cloud environment.

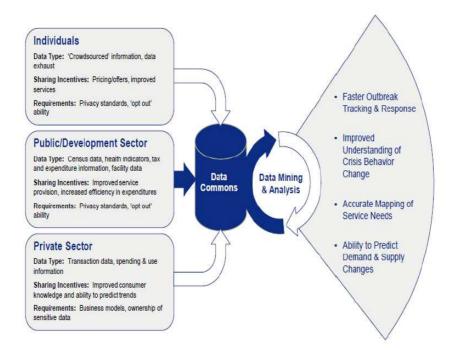


Figure 1: Complex Data Infrastructure/Ecosystem (source: World Economic Forum, 2012)

Big data could be beneficial to resolve critical issues providing the potential of new insights for the advancements of medical especially cancer research, global security, logistics and transportation solutions, identification and predicting terrorism activities, and dealing with socio-economic and environmental issues. The logistics sector is ideally placed to benefit from the technological and methodological advancements of big data. Logistics providers manage a massive flow of goods and that create massive data sets. Millions of shipments every day, of different origins and destinations, size, weight, content, and locations are tracked across global delivery networks (e.g. DHL, UPS) However this present and past data tracking is not fully exploited in order to deliver business value. Most likely there is huge untapped potential for improving operational efficiency and customer experience, and creating useful new business models based on the exploration of big data.

Big data is defined as a complex data infrastructure and new powerful data technologies and management approaches are needed. These solutions are directed to improve the decision making processes and forecasting through application of advanced data exploratory techniques, data mining, predictive analytics and knowledge discovery as presented in figure 1.

The main key characteristics that define big data are volume, velocity, variety and value. Veracity could be also considered an additional characteristic. The related big data models are presented in figure 2.



Figure 2: Big Data Characteristics based on the source: McKinsey Global Institute, 2011

On the other hand because of the characteristics of the cloud, this is an enabler of big data acquisition, and associated software processing tools/strategies. Based on Gartner's estimation, 50% of data will be stored on the cloud by 2016 (Schouten, 2012). However in the reality, cloud has not been widely used for data analytics especially in practical applications.

The availability of cloud based solutions has dramatically lowered the cost of storage, amplified by the use of commodity hardware even on a "pay as-you-go" basis that is directed to effectively and timely processing large data sets. The big data could be analysed "as -a -service". Google BigQuery<sup>1</sup> is an example of providing real-time insights about massive data sets in a cloud based platform.

In cloud computing, data and software applications are defined, developed and implemented as services. These services have defined a multi-layered infrastructure and are described as follows (Grace, 2010; Mell and Grance, 2009):

1. https://cloud.google.com/products/big-query

- 1. *Software as a Service* (SaaS): applications are hosted and delivered online via a web browser offering traditional desktop functionality
- 2. *Platform as a Service* (PaaS): the cloud provides the software platform for systems (as opposed to just software)
- Infrastructure as a Service (IaaS): a set of virtual computing resources, such as storage and computing capacity, are hosted in the cloud; customers deploy and run only their own applications for obtaining the needed services.

On the other hand it is recognised the tension between big data strategies, and solutions versus information security and data privacy requirements. The big data might enable the violation of the privacy and information security breaches and by consequence decreasing the trust in data defined as a service in the cloud. Big data stored and processed in the cloud could lack a centralized control and ownership.

According to McKinsey Global Institute (2011), big data is seen as "the next frontier for innovation, competition and productivity" and as such the related applications will contribute to economic growth. The positive impacts of big data provide a huge potential for organisations. In order to achieve these aspirations several issues should be analysed and discussed in the context of complex systems and using systems approaches such as holistic thinking and system dynamics.

Therefore major issues are emerging and this work-in-progress attempts to discuss a few key aspects directed to the development and adopting data mining techniques and strategies for cloud based big data applications.

#### **Background and Research Approach**

Demirkan and Delen (2013) have defined some research directions including dealing with affordable analytics for big data in the cloud. This means using open-source, freeof-charge data/text mining algorithms and associated commercial tools (e.g. R, Rapid-Miner, Weka, Gate, etc.) New approaches need to provide solutions for moving these tools to the cloud and produce efficient and affordable applications for discovering knowledge and patterns from very large/big data sets directed to support business intelligence and decision support systems applications.

The principles of data/information- as- a- service, data/information-security-as-aservice, and analytics- as- a- service are explained in the context of using service oriented architecture.

However the cloud platforms are not completely following service oriented thinking and even more there is a debate that cloud computing is different of service oriented architectures, and grid computing.

The main motivation of adopting cloud computing for analytics applied for large (big) data sets are based on the accessibility of cloud solutions outside the a web based organisation communication secured with firewalls. Cloud based business analytics are also cost effective, easy to set up and test. The results are easy to be shared outside the organisations. Greg Sheldon, CIO of Elite Brands said "The biggest benefit, is to be able to access huge amounts of information from anywhere you have web access, specifically on an iPad. This is beneficial to our field sales team when information is needed on the fly." (Fields, 2013:2)

The main research questions are related but not limited to the following aspects:

- 1. In the context of cloud based big data how analytics (e.g. data mining), information and knowledge management disciplines and strategies will evolve?
- 2. What should be the techniques, strategies and practices to increase the benefits and minimise the information risks ?
- 3. How to deal with the growing number of security breaches and cyber security risks and increase organisational awareness, business agility and resilience?
- 4. How to adapt the existing legislation such as data protection law, regulations and standards? Moreover, the ethics issues will be considered.

### Efforts and Challenges of Big Data Mining and Discovery

Considering big data a collection of complex and large data sets that are difficult to process and mine for patterns and knowledge using traditional database management tools or data processing and mining systems a briefing of the existing efforts and challenges is provided in this paragraph. While presently the term big data literally concerns about data volumes, Wu et al. (2013) have introduced HACE theorem that described the key characteristics of the big data as (1) huge based on heterogeneous and diverse data sources, (2) autonomous with distributed and decentralized control, and (3) complex and evolving in data and knowledge associations. Generally, business intelligence applications are using analytics that are grounded mostly in data mining and statistical methods and techniques. These strategies are usually based on the mature commercial software systems of RDBMS, data warehousing, OLAP, and BPM. Since the late 1980s, various data mining algorithms have been developed mainly within the artificial intelligence, and database communities. In the IEEE 2006 International Conference on Data Mining (ICDM), the 10 most influential data mining algorithms were identified based on expert nominations, citation counts, and a community survey (Chen et al, 2012). In ranked order, these techniques are as follows C4.5, k-means, SVM (support vector machine), Apriori, EM (expectation maximization), PageRank, AdaBoost, kNN (k-nearest neighbors), Naïve Bayes, and CART (Wu et al, 2007). These algorithms are for classification, clustering, regression, association rules, and network analysis. Most of these well known data mining algorithms have been implemented and deployed in commercial and open source data mining systems (Witten et al. 2011). Chen at al. (2012) have compared data base management systems and analytics as well as ETL with using MapReduce and Hadoop. Hadoop was originally a (distributed) file system approach applying the MapReduce framework that is a software approach introduced by Google in 2004 to support distributed computing on large/big data sets. Recently, Hadoop has been developed and used as a complex ecosystem that includes a wider range of software systems, such as HBase (a distributed table store), Zookeeper (a reliable coordination service), and the Pig and Hive highlevel languages that compile down MapReduce components (Rabkin and Katz, 2013). Therefore in the recent conceptual approaches Hadoop is primarily considered an eco-

system or an infrastructure or a framework and not just the file system alongside MapReduce components.

The big data and cloud computing frameworks include the Google MapReduce, Hadoop Reduce, Twister, Hadoop++, Haloop, and Spark etc. which are used to process big data and run computational tasks. The cloud databases are used to store massive structured and semi-structured data generated from different types of applications. The most important cloud databases include the BigTable, Hbase, and HadoopDB. In order to implement an efficient big data mining and analysis framework, the data warehouse processing is also important. The most important data warehouse processing technologies include the Pig, and Hive.

Strambei (2012) has suggested a different conceptual interpretation of the OLAP technology considering the emergence of web services, cloud computing and big data. One of the most important consequences could be widely open access to web analytical technologies. The related approach has evaluated the OLAP Web Services viability in the context of the cloud based architectures.

There are also a few reported practical applications of big data mining in the cloud. Patel et al. (2012) have explored a practical solution to big data problem using the Hadoop data cluster, Hadoop Distributed File System alongside Map Reduce framework using big data prototype application and scenarios. The outcomes obtained from various experiments indicate promising results to address big data implementation problems.

The challenges for moving beyond existing data mining and knowledge discovery techniques (NESSI, 2012, Witten et al, 2011) are as follows:

- 1. a solid scientific foundation to support the selection of a suitable analytical method and a software design solution
- 2. new efficiency and scalable algorithms and machine learning techniques
- the motivation of using cloud architecture for big data solutions and how to achieve the best performance of implementing data analytics using cloud platform (e.g. big data as- a-service)
- dealing with data protection and privacy in the context of exploratory or predictive analysis of big data
- 5. software platforms and architectures alongside adequate knowledge and development skills to be able to implement them
- 6. ability to understand not only the data structures (and the usability for a given processing method), but also the information and business value that is extracted from big data.

## **Concluding Remarks**

The emergence of big data movement has energized the data mining, knowledge discovery in data bases and associated software development communities, and it has introduced complex, interesting questions for researchers and practitioners. As organisations continue to increase the amount and values of collected data formalizing the process of big data analysis and analytics becomes overwhelming. In this paper, we discuss some existing approaches and have analysed the main issues of big data mining, knowledge, and patterns discovery in the data driven cloud computing environment. This research will be progressed providing theoretical and practical approaches that will be tested through the development of case studies for the application of big data particularly in collaborative logistics.

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