

A Study on Data Mining with Big Data

B. Zakia Bano¹, Dr. K. Pavan Kumar²

¹M.Tech student, Dr K V Subba reddy college of Engineering for women,

²Professor, Dr K V Subba reddy college of Engineering for women,
Kurnool, Andhra Pradesh, India.

Abstract- Big data is the term for a collection of data sets which are large and complex, it contain structured and unstructured both type of data. Data comes from everywhere, sensors used to gather climate information, posts to social media sites, digital pictures and videos etc. This data is known as big data. Useful data can be extracted from this big data with the help of data mining.

I. INTRODUCTION

The term 'Big Data' appeared for first time in 1998 in a Silicon Graphics (SGI) slide deck by John Mashey with the title of "Big Data and the Next Wave of Infrastructure" [4]. Big Data mining was very relevant from the beginning, as the first book mentioning 'Big Data' is a data mining book that appeared also in 1998 by Weiss and Indrukya [10]. However, the first academic paper with the words 'Big Data' in the title appeared a bit later in 2000 in a paper by Diebold [3]. The origin of the term 'Big Data' is due to the fact that we are creating a huge amount of data every day. Usama Fayyad [6] in his invited talk at the KDD BigMine'12 Workshop presented amazing data numbers about internet usage, among them the following: each day Google has more than 1 billion queries per day, Twitter has more than 250 million tweets per day, Facebook has more than 800 million updates per day, and YouTube has more than 4 billion views per day.

The data produced nowadays is estimated in the order of zettabytes, and it is growing around 40% every year. A new large source of data is going to be generated from mobile devices, and big companies as Google, Apple, Facebook, Yahoo, Twitter are starting to look carefully to this data to find useful patterns to improve user experience.

In Big data the information comes from multiple, heterogeneous, autonomous sources with complex relationship and continuously growing. upto

2.5 quintillion bytes of data are created daily and 90 percent data in the world today were produced within past two years [1]. For example Flickr, a public picture sharing site, where in an average 1.8 million photos per day are received from February to March 2012 [5]. This shows that it is very difficult for big data applications to manage, process and retrieve data from large volume of data using existing software tools. It's become a challenge to extract knowledgeable information for future use [7]. There are different challenges of Data mining with Big Data. We overlook it in next section. Currently Big Data processing depends upon parallel programming models like MapReduce, as well as providing computing platform of Big Data services. Data mining algorithms need to scan through the training data for obtaining the statistics for solving or optimizing model parameter. Due to the large size of data it is becoming expensive to analyze data cube. The Map-Reduce based approach is used for data cube materialization and mining over massive datasets using holistic (non algebraic) measures like TOP-k for the top-k most frequent queries. MRCube approach is used for efficient cube computation. We need new algorithms, and new tools to deal with all of this data. Doug Laney was the first one in talking about 3 V's in Big Data management:

- Volume: there is more data than ever before, its size continues increasing, but not the percent of data that our tools can process.
 - Variety: there are many different types of data, as text, sensor data, audio, video, graph, and more.
 - Velocity: data is arriving continuously as streams of data, and we are interested in obtaining useful information from it in real time.
- Nowadays, there are two more V's:
- Variability: there are changes in the structure of the data and how users want to interpret that data.

- Value: business value that gives organization a compelling advantage, due to the ability of making decisions based in answering questions that were previously considered beyond reach.

These different characteristics of Big Data make it challenge for discovering useful information or knowledge from it. After analyzing and research challenge form a three tier structure framework to mention different challenges at different tier, as shown in fig.1.

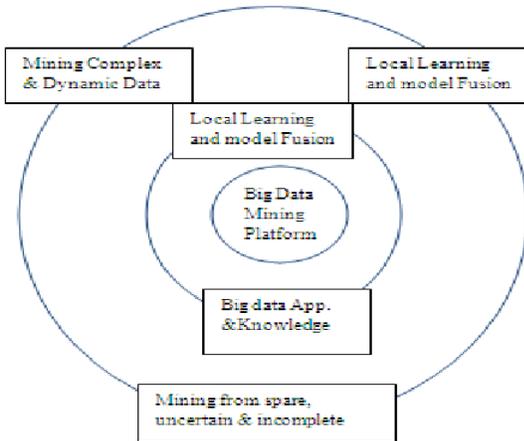


Fig.1. A Big Data processing Framework

The challenges at tier I focus on low-level data accessing and arithmetic computing procedures, Challenges on information sharing and privacy. Big Data often stored on different location and it is continuously growing that's why an effective computing platform to take distributed large scale data storage into consideration for computing.

Tier II concentrate on high-level semantics, application domain knowledge for different applications of big data and the user privacy issues. This information provides benefits to Big data access but also add a technical barriers to Big Data access (Tier I) and mining algorithms (Tier II). The Outmost tier is tier III which challenges the actual mining algorithms.

At this tier III the mining challenges concentrate on algorithm designs in tacking the difficulties which is raised by the big data volumes, distributed data distribution, complex and dynamic characteristics. Tier III contains three stages. In first stage sparse, heterogeneous, uncertain, incomplete and multisource data is preprocessed by data fusion technique. In second stage after preprocessing stage complex and dynamic data are mined. Third stage is

for local learning and model fusion, where the global knowledge is obtained by local learning and model fusion is tested and the relevant information is feedback to preprocessing stage.

II. BIG DATA CHARACTERISTICS

A. HACE Theorem.

Big Data starts with large-volume, heterogeneous, autonomous sources with distributed and decentralized control, and seeks to explore complex and evolving relationships among data. These characteristics make it an extreme challenge for discovering useful knowledge from the Big Data. In a naïve sense, we can imagine that a number of blind men are trying to size up a giant Camel, which will be the Big Data in this context. The goal of each blind man is to draw a picture (or conclusion) of the Camel according to the part of information he collects during the process. Because each person's view is limited to his local region, it is not surprising that the blind men will each conclude independently that the camel "feels" like a rope, a hose, or a wall, depending on the region each of them is limited to.

To make the problem even more complicated, let us assume that the camel is growing rapidly and its pose changes constantly, and each blind man may have his own (possible unreliable and inaccurate) information sources that tell him about biased knowledge about the camel (e.g., one blind man may exchange his feeling about the camel with another blind man, where the exchanged knowledge is inherently biased). Exploring the Big Data in this scenario is equivalent to aggregating heterogeneous information from different sources (blind men) to help draw a best possible picture to reveal the genuine gesture of the camel in a real-time fashion. Indeed, this task is not as simple as asking each blind man to describe his feelings about the camel and then getting an expert to draw one single picture with a combined view, concerning that each individual may speak a different language (heterogeneous and diverse information sources) and they may even have privacy concerns about the messages they deliberate in the information exchange process. The term Big Data literally concerns about data volumes, HACE theorem suggests that the key characteristics of the Big Data are

Huge with heterogeneous and diverse data sources: -One of the fundamental characteristics of

the Big Data is the huge volume of data represented by heterogeneous and diverse dimensionalities. This huge volume of data comes from various sites like Twitter, Myspace, Orkut and LinkedIn etc.

Decentralized control:- Autonomous data sources with distributed and decentralized controls are a main characteristic of Big Data applications. Being autonomous, each data source is able to generate and collect information without involving (or relying on) any centralized control. This is similar to the World Wide Web (WWW) setting where each web server provides a certain amount of information and each server is able to fully function without necessarily relying on other servers

Complex data and knowledge associations:- Multi-structure, multi-source data is complex data, Examples of complex data types are bills of materials, word processing documents, maps, time-series, images and video. Such combined characteristics suggest that Big Data require a “big mind” to consolidate data for maximum values.

III. SYSTEM AND METHODOLOGY

Doug Laney was the first one talking about 3V's in Big Data Management

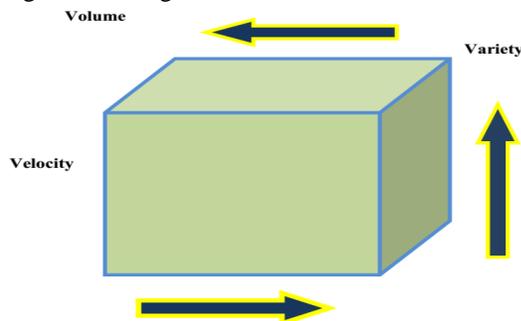


Fig 2 V's in Big Data Management

Volume: The amount of data. Perhaps the characteristic most associated with big data, volume refers to the mass quantities of data that organizations are trying to harness to improve decision-making across the enterprise. Data volumes continue to increase at an unprecedented rate.

Variety: Different types of data and data sources. Variety is about managing the complexity of multiple data types, including structured, semi-structured and unstructured data. Organizations need to integrate and analyze data from a complex array of both traditional and non-traditional information sources, from within and outside the enterprise. With the

explosion of sensors, smart devices and social collaboration technologies, data is being generated in countless forms, including: text, web data, tweets, audio, video, log files and more.

Velocity: Data in motion. The speed at which data is created, processed and analyzed continues to accelerate.

Generally, data mining (sometimes called data or knowledge discovery) is the process of analyzing data from different perspectives and summarizing it into useful information - information that can be used to increase revenue, cut costs, or both. Technically, data mining is the process of finding correlations or patterns among dozens of fields in large relational database. Data mining as a term used for the specific classes of six activities or tasks as follows:

1. Classification
2. Estimation
3. Prediction
4. Association rules
5. Clustering
6. Description

A. Classification: Classification is a process of generalizing the data according to different instances. Several major kinds of classification algorithms in data mining are Decision tree, k-nearest neighbor classifier, Naive Bayes, Apriori and AdaBoost. Classification consists of examining the features of a newly presented object and assigning to it a predefined class. The classification task is characterized by the well-defined classes, and a training set consisting of reclassified examples.

B. Estimation: Estimation deals with continuously valued outcomes. Given some input data, we use estimation to come up with a value for some unknown continuous variables such as income, height or credit card balance.

C. Prediction: It's a statement about the way things will happen in the future, often but not always based on experience or knowledge. Prediction may be a statement in which some outcome is expected.

D. Association Rules: An association rule is a rule which implies certain association relationships among a set of objects (such as “occur together” or “one implies the other”) in a database.

E. Clustering: Clustering can be considered the most important unsupervised learning problem; so, as

every other problem of this kind, it deals with finding a structure in a collection of unlabeled data.

The Big Data phenomenon is intrinsically related to the open source software revolution. Large companies as Facebook, Yahoo!, Twitter, LinkedIn benefit and contribute working on open source projects. Big Data infrastructure deals with Hadoop, and other related software as:

- **Apache Hadoop [2]:** software for data-intensive distributed applications, based in the MapReduce programming model and a distributed file system called Hadoop Distributed Filesystem (HDFS). Hadoop allows writing applications that rapidly process large amounts of data in parallel on large clusters of compute nodes. A MapReduce job divides the input dataset into independent subsets that are processed by map tasks in parallel. This step of mapping is then followed by a step of reducing tasks. These reduce tasks use the output of the maps to obtain the final result of the job.
- **Apache Hadoop related projects [11]:** Apache Pig, Apache Hive, Apache HBase, Apache ZooKeeper, Apache Cassandra, Cascading, Scribe and many others.
- **Apache S4 [8]:** platform for processing continuous data streams. S4 is designed specifically for managing data streams. S4 apps are designed combining streams and processing elements in real time.
- **Storm [9]:** software for streaming data-intensive distributed applications, similar to S4, and developed by Nathan Marz at Twitter.

IV. SIMULATION RESULTS



Fig. 3

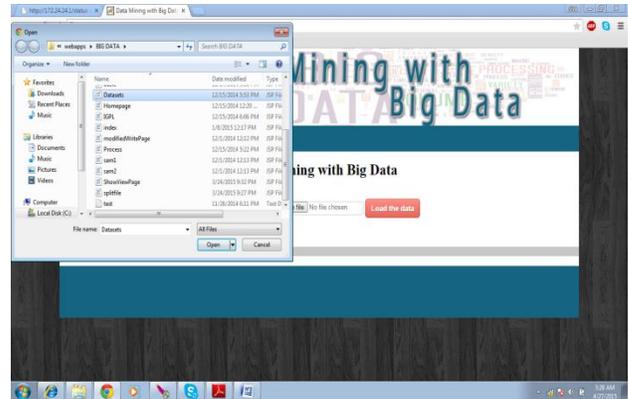


Fig. 4



Fig. 5

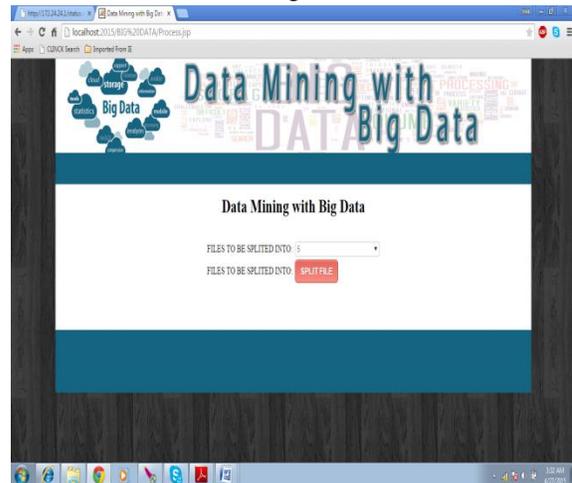


Fig. 6

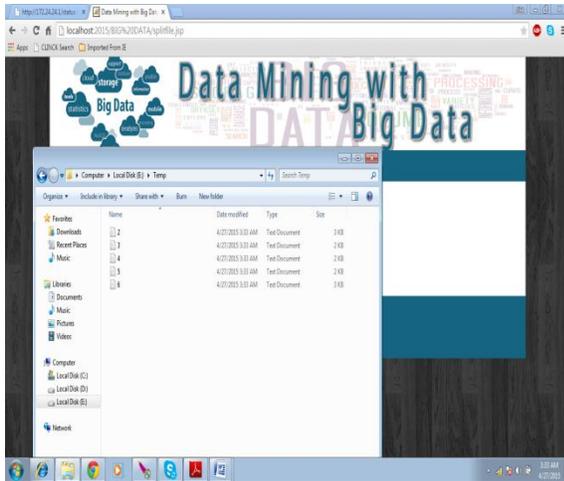


Fig. 7



Fig. 8

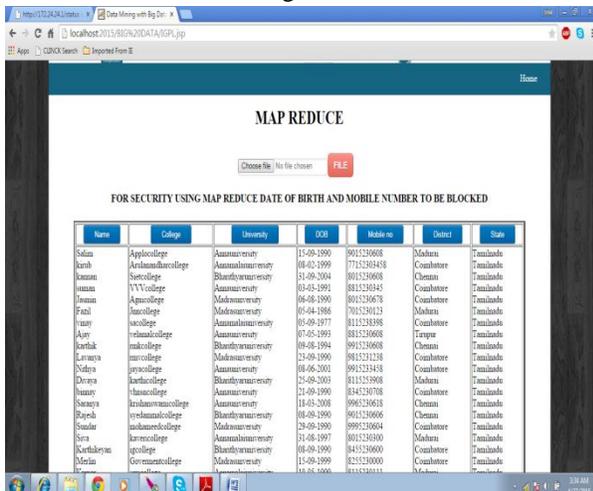


Fig. 9



Fig.10

V. CONCLUSION

In real-world applications managing and mining Big Data is Challenging task, As the data concernlarge in a volume, distributed and decentralized control and complex. There are several challenges at data,model and system level. We need computing platform to handle this Big Data. The MapReduce framework isone of the most important parts of big data processing, and batch oriented parallel computing model.

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BIODATA

Author

B.Zakia Bano presently pursuing her M.Tech (CSE) in Dr K V Subba reddy college of Engineering for women, Kurnool, Andhra Pradesh, India.

CoAuthor

Dr Pavan Kumar received Phd. Presently working as Professor in Dr K V Subba reddy college of Engineering for women, Kurnool, Andhra Pradesh, India.