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SURVEY ON BIG DATA USING DATA MINING

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Abstract: In this paper presents Big Data Problems using Data Mining. There is broad recognition of the value of data as well as products obtained through analyzing it. About big data "Size is the only thing that matters." Various Popular news media now appreciates the value of Big Data as evidenced by coverage in the Economist the New York Times and National Public Radio Industry is abuzz with the promise of Big Data Government agencies have recently announced significant programs towards addressing challenge of Big Data. But Yet, many have a very narrow interpretation of what that means, and we lose track of the fact that there are multiple steps to the data analysis pipeline, whether the data are big or small depend on size of data and there are challenges with Big Data. Big Data concern large-volume, complex, growing data sets with multiple, autonomous sources. so for managing and processing this data we must have some techniques. In this work first different existing technique will be studied and analyzed about Big Data. Here we will include the concepts related to big data like analysis and prediction etc. The existing approaches perform better but having some drawbacks. So, they cannot be applied to the various situations. To overcome some drawbacks we are going to propose Clustering Approach for Collaborative Filtering which will gives us better result. So, we can apply it to various situations, depending upon these techniques, we will try to implement concepts related to Big Data processing.

Keywords: Data Mining Challenges With Big Data, *Hadoop's Distributed File System*, Map-Reduce Framework, Clustering Approach for Collaborative

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INTRODUCTION

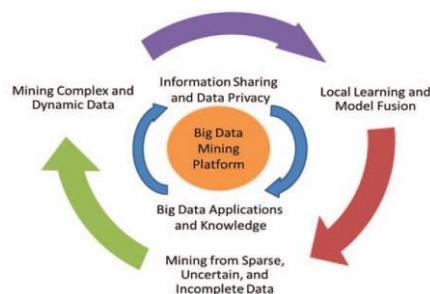
Big data is nothing the large amount of data. There is broad recognition of the value of data as well as products obtained through analyzing it [2]. About big data “Size is the only thing that matters.” Various Popular news media now appreciates the value of Big Data as evidenced by coverage in the Economist the New York Times [5], and National Public Radio Industry is abuzz with the promise of Big Data Government agencies have recently announced significant programs towards addressing challenges of Big Data. But yet, many have a very narrow interpretation of what that means, and we lose track of the fact that there are multiple steps to the data analysis, whether the data are big or small depend on size of data. There is work to be done at each step and there are challenges with Big Data.

The rise of Big Data applications where data collection has grown tremendously and is beyond the ability of commonly used software tools to capture, manage, and process within a “tolerable elapsed time.” The most fundamental challenge for Big Data applications is to explore the large volumes of data and extract useful information or knowledge for future actions [1]. In many situations, the knowledge extraction process has to be very efficient and close to real time because storing all observed data is nearly infeasible. The key challenges for such as Data accessing and computing, Data privacy and domain knowledge.

We are used to thinking of Big Data as always telling us the truth, but this is actually far from reality about big data. We have to deal with erroneous data: some news reports are inaccurate A problem with current Big Data analysis is the lack of coordination between database systems, which host the data and provide SQL querying, with analytics packages that perform various forms of non-SQL processing, such as data mining and statistical analyses.

In short, there is collaborative filtering is required to extract value from data. Heterogeneity, incompleteness, scale, timeliness, privacy and process complexity give rise to challenges at all phases.

I. DATA MINING CHALLENGES WITH BIG DATA:-



The challenges at Tier I focus on data accessing and arithmetic computing procedures. Because Big Data are often stored at different locations and data volumes may continuously grow, an effective computing platform will have to take distributed large-scale data storage into consideration for computing.

The challenges at Tier II center around semantics and domain knowledge for different Big Data applications. Such information can provide additional benefits to the mining process.

At Tier III, the data mining challenges concentrate on algorithm designs in tackling the difficulties raised by the Big Data volumes, distributed data distributions, and by complex and dynamic data characteristics.

II.HADOOP'S DISTRIBUTED FILE SYSTEM:-

Hadoop's Distributed File System is designed to reliably store very large files across machines in a large cluster.

It is inspired by the Google File System. Hadoop DFS stores each file as a sequence of blocks, all blocks in a file except the last block are the same size. Blocks belonging to a file are replicated for fault tolerance. The block size and replication factor are configurable per file. Files in HDFS are "write once" and have strictly one writer at any time.

III.MAP-REDUCE FRAMEWORK

MapReduce is a programming model and an associated implementation for processing and generating large data sets with a parallel, distributed algorithm on a cluster. Users specify a *map* function that processes a key/value pair to generate a set of intermediate key/value pairs, and a *reduce* function that merges all intermediate values associated with the same intermediate key.

MapReduce is a programming model and an associated implementation for processing and generating large data sets with a parallel, distributed algorithm on a cluster.

A MapReduce program is composed of a Map() procedure that performs filtering and sorting (such as sorting students by first name into queues, one queue for each name)

Reduce() procedure that performs a summary operation (such as counting the number of students in each queue, yielding name frequencies).

Even though the previous pseudo-code is written in terms

of string inputs and outputs, conceptually the map and reduce functions supplied by the user have associated types:

map (k1,v1) ! list(k2,v2)

reduce (k2,list(v2)) ! list(v2)

I.e., the input keys and values are drawn from a different

domain than the output keys and values. Furthermore, the intermediate keys and values are from the same domains as the output keys and values.

IV. CLUSTERING APPROACH FOR COLLABORATIVE:-

To overcome some drawbacks we are going to use Clustering Approach for Collaborative Filtering which will give us better results and also we are going to use some framework to overcome the drawback.

Collaborative filtering is required to extract value from data

Why “collaborative”? Basically, someone else (in fact many someones) have gone to the effort of viewing/filtering things, and chosen the best few. You get a recommendation of the best few, without having to spend the effort.



Fig 1. Everyday Examples of Collaborative Filtering



Fig 1.1. Everyday Examples of Collaborative Filtering

CONCLUSION:-

In this paper, some of the most important methods for Big Data are described. Thus we studied existing techniques for various problem. Depending on these techniques like Map Reduce, we are going to proposed Clustering Approach for Collaborative Filtering.

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