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# **Processing the Query in Memory Management System Using Scheduling Techniques in Big Data**

<sup>[1]</sup> Vijayakumar G, <sup>[2]</sup> Sabarinathan P

<sup>[1]</sup>PG Scholor, Dept of CSE, Pavendhar Bharathidasan College of Engineering and Technology, Trichy, India

<sup>[2]</sup>Assistant Professor, Dept of CSE, Pavendhar Bharathidasan College of Engineering and Technology, Trichy, India

**ABSTRACT :** In the concept of big facts running, Big data is a broad term for figures set as a result great or compound that usual data processing applications be insufficient. challenge take in study, incarcerate, hunt, division, cargo space, transport, apparition, and in arrange privacy. The term often refers simply to the use of predictive analytics or other certain advanced methods to extract value from records, and not often to a fussy amount of information set. The main idea of this projected occupation be spread storage space arrangement and Cache remembrance. During the data meting out, its check the size of the file whether if its inside the extent wealth its stock up on the folder if its exceeds means its stored on the collection, During user handing out the data its retrieves the data since the together recollection administration user can download the documents by means of near mechanism (revise DA).

KEYWORDS: Primary memory, DRAM, relational databases, distributed databases, query processing

### I. INTRODUCTION

In big data, RDBMS are used to maintain the memory management processing efficiently. It contains two processes such as storing information and retrieving information from memory. The retrieved information is retrieved from two location such that database (distributed), or cache memory. Same like that, if the information is less than the limited size, it will store in database. Otherwise it will store in cache memory. Regarding User query processing the information is retrieved from the available servers. Cloud computing is a model for enabling ubiquitous, convenient, and on-demand network access to a shared pool of configurable computing resources (e.g., networks, servers, storage, applications, and services) that can be rapidly provisioned and released with minimal management effort or service provider interaction. There are two main categories of cloud infrastructure: public cloud and private cloud. To take advantage of public clouds, data owners must upload their data to commercial cloud service providers which are usually considered to be semi trusted, that is, honest but curious. That means the cloud service providers will try to find out as much secret information in the users' outsourced data as possible, but they will honestly follow the protocol in general. Traditional access control techniques are based on the assumption that the server is in the trusted domain of the data owner, and therefore an omniscient reference monitor can be used to enforce access policies against authenticated users. However, in the cloud computing paradigm this assumption usually does not hold, and therefore these solutions are not applicable. There is a need for a decentralized, scalable, and flexible way to control access to cloud data without fully relying on the cloud service providers. Data encryption is the most effective in regard to preventing sensitive data from unauthorized access. In traditional public key encryption or identity-based encryption systems, encrypted data is targeted for decryption by a single known user. Unfortunately, this functionality lacks the expressiveness needed for more advanced data sharingBig data is a buzzword, or catch-phrase, used to describe a massive volume of both structured and unstructured data that is so large it is difficult to process using traditional database and software techniques. The volume of data is too big or it moves too fast or it exceeds current processing capacity. This data comes from everywhere: sensors used to gather climate information, posts to social media sites, digital pictures and videos, purchase transaction records, and cell phone GPS signals to name a few. This data is big data. The unacceptable performance was initially encountered by Internet companies such as Amazon, Google, Face book and Twitter, but is now also becoming an obstacle for other companies/organizations which desire to provide a meaningful real-time service (e.g., real-time bidding, advertising, social gaming). Generally, data mining (sometimes called data or knowledge discovery) is the process of analyzing data from different perspectives and summarizing it into useful



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information - information that can be used to increase revenue, cuts costs, or both. Data mining software is one of a number of analytical tools for analyzing data. It allows users to analyze data from many different dimensions or angles, categorize it, and summarize the relationships identified. Technically, data mining is the process of finding correlations or patterns among dozens of fields in large relational databases.

#### II. RELATED WORK

In [1] Michael Vrable, Stefan Savage, Geoffrey M. Voelker et al Presents Cumulus is a system for efficiently implementing file system backups over the Internet, taking advantage of the growing availability of cheap storage options available online. Cloud service offerings such as Amazon's Simple Storage Service (S3), a part of Amazon Web Services, offer cheap storage at a fixed cost per gigabyte (no minimums or maximums) and are appealing for backup, since they provide an easy way to safely store data off-site. There are pre-packaged online services specifically built for backup, such as Mozy and Carbonite. Cumulus explores the other end of the design space: building on top of a very generic cloud storage layer, an example of what we refer to as building on the "thin cloud." Using a generic, minimalist interface means that Cumulus is portable to virtually any online storage service the client implements all application logic.In [2] D. Lomet, and S. Sengupta et al presents As the world moves to digital storage for archival purposes, there is an increasing demand for systems that can provide secure data storage in a cost-effective manner. By identifying common chunks of data both within and between files and storing them only once, deduplication can yield cost savings by increasing the utility of a given amount of storage. Unfortunately, deduplication exploits identical content, while encryption attempts to make all content appear random; the same content encrypted with two different keys results in very different cipher text.In [3] John R. Douceur, Atul Adya, William J. Bolosky, Dan Simon, Marvin Theimer et al Presents The Farsite distributed file system provides availability by replicating each file onto multiple desktop computers. Since the replication consumes significant storage space, it is important to reclaim used space where possible. Measurement of over 500 desktop file systems shows that nearly half of all consumed space is occupied by duplicate files. present a mechanism to reclaim space from this incidental duplication to make it available for controlled file replication.In [4] Austin T. Clements\_ Irfan Ahmad Murali Vilayannur Jinyuan Li et al presents File systems hosting virtual machines typically contain many duplicated blocks of data resulting in wasted storage space and increased storage array cache footprint. Deduplication addresses these problems by storing a single instance of each unique data block and sharing it between all original sources of that data. While deduplication is well understood for file systems with a centralized component, we investigate it in a decentralized cluster file system, specifically in the context of VM storage. Each host periodically and independently processes the summaries of its locked files, merges them with a shared index of blocks, and reclaims any duplicate blocks. In [5] William J. Bolosky, John R. Douceur, David Ely, And Marvin Theimer et al presents consider an architecture for a serverless distributed file system that does not assume mutual trust among the client computers. It provides security, availability, and reliability by distributing multiple encrypted replicas of each file among the client machines. To assess the feasibility of deploying this system on an existing desktop infrastructure, measure and analyze a large set of client machines in a commercial environment. In particularly, measure and report results on disk usage and content; file activity; and machine uptimes and lifetimes, and loads.conclude that the measured desktop infrastructure would passably support, providing availability on the order of one unfilled file request per user per thousand days.

### III. PROPOSED ALGORITHM

Revised DA algorithmiteratively find a better weakly stable matching with respect to jobs. The blocking job is removed from the previous machine, so that it can make new offers to machines that have rejected it before. This ensures that the algorithm does not produce new type with blocking pairs. At each stage, Revised DA is proposed with the selected set of proposing jobs and the entire set of machines with updated capacity.

#### **IV.PSEUDO CODE**

Step1: A Multistage graph G=(V,E) is a directed graph in which vertices are partitioned into k>=2 disjoint set(set Vi) where  $[1 \le i \le k]$ . In addition, if (u, v) is an edge E then  $u \in vi, v \in vi+1$ .



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Step2: Let c(I,j) be the cost of edge(I,j). The cost of a path from(S to T) is the sum of costs of the edges on the path. Step3: Ihe multistage graph problem is to find the minimum cost path from "S" to "T". Step4: The value on the edges are called cost of edges.  $Cost(i,j)=min\{c(j,l)+cost(i+1,l)\}$  $l \notin vi+1$  (i.e)cost(i,j)=cost(levels,no of nodes at the level)C is weight assigned on each edge. K is the no of stages.

#### IV. EXPERIMENTAL RESULTS

First up all to register the user for giving all the details(name,username,password,d.o.b,phone number,e-mail-id).if the registration is completed to send the security key for corresponding user email id.after next to perform the login function enter the username and password then finally enter the security key code from user e-mail id then complete the login function.next to perform specific function for user account.if you don't know the security code did not to be processed

Memory Management	Home	Service Provider	Cloud User	New User	LogOut
User Login					
Usemame					
्रांह्यप्र Password					
•••••					
Login Key					
submit					

Fig 1: login the user

#### search:

next one is search a file the admin upload the files, the user download the files. To search the files if the file available to display, otherwise to send the replay for not available for the data.



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nemoi	ymanagemeni	Search File	View Files	Edit Files	LogOut
		Search a File			
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	Search the File	Search			
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#### Download:

if the file is available to download using select resource provision technique(virtual machine)for user choice, the resource provision having the capability to download the file It can be processed otherwise to send the message for to select another resource provision.

Fig 2:search the files

Memory Management		Search File		LogOut	
	Select Re	source	e Provision		
elcome vijoy e Size 3.9233630307006836 MB					
	RP Name Instance T	ype Available S	pace Download Time		
	RP1 c1.mediur	n 3	5 min		
	RP3 cc2.8klorg	e 31	1 min		
Subrect RPSolect • Subr	mit				

Fig 3:select resource provision to download the files

### VI .CONCLUSION AND FUTURE WORKS

A memory becomes the new disk, processing the query in mermory management system using scheduling techniquies in big data becomes increasingly interesting for both academy and industry. Shifting the data storage layer from disks to main memory can lead to improvement in terms of response time and throughput. When data access becomes faster, every source of overhead that does not matter in traditional disk based systems, may degrade the overall performance significantly. The shifting prompts a re-thinking of the design of traditional systems, especially for databases, query processing, fault-tolerance, etc. The future direction is based on file processing using, the virtual machine. The user can download the files by using the virtual machine. The Virtual machine implement the Online Scheduling algorithm and



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Multistage DA Algorithm. In online scheduling the decisions regarding how to schedule tasks are done during the runtime of the system. The scheduling decisions are based on the tasks priorities which are either assigned dynamically or statically.Static priority driven algorithms assign fixed priorities to the tasks. Dynamic priority driven algorithms assign the priorities to tasks during runtime. Multistage DA algorithm iteratively finds a better weakly stable matching with respect to job.

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