

Exploring Load Balancing: Issues, Methods and Strategic Outlook in Cloud Computing

Zainab Khan and Kavita Agrawal

Department of Computer Science and Engineering, Integral University, Lucknow, India

Keywords: Cloud Computing (CC), Load-Balancing (LB), Resource Allocation, Heuristics Algorithms Approach, ACO, PSO, Blockchain.

Abstract: Through the provision of scalable, on-demand resources via the Internet, cloud computing has completely transformed contemporary computing. Effective load-balancing remains a critical challenge in achieving efficient resource use, reducing response times, and preventing system overload. A comprehensive evaluation of LB strategies in distributed computing is given in this document, which divides them into conventional and contemporary categories, such as static, dynamic, heuristic, and AI-based approaches. It also examines important security, fault tolerance, and energy efficiency issues. And scalability. The study highlights new developments influencing load balancing in cloud environments going forward, such as edge computing, blockchain integration, and machine learning-driven optimization. The purpose of this survey is to give researchers and practitioners useful information about how load balancing is changing, enabling improvements in the effectiveness and performance of CC.

1 INTRODUCTION

The use of CC has become a major craze in the last few years, leading to significant advancements in distributed systems and the development of extensive computer networks. Cloud services are provided to customers worldwide by CC companies like IBM, Amazon, and Google. Under this new paradigm, end users can access apps and services whenever they want rather than having to install them on their local computers (M. Shahid, et.al, 2020)

The foundation and key component of cloud-based applications is virtualization. Inefficient handling of the migration process and the allocations of VMs can greatly impact the way on-demand as well as scalable services are rendered to customers (D. Shafiq,et.al, 2021) Among the top three difficulties with Cloud computing, according to (M. Shahid, et.al, 2020), mentioned as cloud performance. This study intends to improve the infrastructure as a service model's resource allocation, a key concept idea in cloud computing, by balancing the resources offered to the customers with the amount of work as well as requests made by users on servers (D. Shafiq,et.al, 2021). Allocating resources is one of the challenges with CC, and it also plays a role in LB.

This problem is also present in systems for wireless communication, it's essential to allocate resources in a fair and balanced manner while also considering user priorities (F. Zabini,et.al.2017)

There are two ways to classify cloud computing: by location or by services provided. A cloud has the potential to be categorized as public, private, hybrid, or community, depending on its location (J. Shah, 2017), Anybody can use public cloud services, and the infrastructure is housed on a service provider's property. Public clouds are the most economical, but they are also the most susceptible to different types of attacks. Access to a private cloud is limited to a single person or entity. Although it costs more, it offers the user the highest level of security and control. Combining both public and private clouds for various uses depending on organizational needs is known as a hybrid cloud. A community cloud is made up of a shared infrastructure that is utilized by numerous organizations with similar management and data. (P. Kumar,2019)

The following are the study's primary goals:

- To investigate different load-balancing strategies that are discussed within the literature.

- To categorize different LB methods as well as offer a summary of the difficulties and problems that LB currently faces.
- To provide an overview of potential study topics for future load-balancing technique improvements.

Challenges of Cloud Computing.

Multiple challenges exist in the usage of CC. Three important challenges in CC include data protection (A. Chaturvedi, et.al, 2019) data availability, security, management of execution and load balancing, and fault tolerance (R. Khan and M. Ahmad, 2016). Protection of data: One important factor that must be considered is data protection. The privacy issue persists when data is stored on the cloud. Likewise, the organization's privacy concerns are exacerbated by the fact that the exact location of repository sites is frequently unknown. Most current models use data centers to protect knowledge through firewalls (A. Chaturvedi, et.al, 2019).

1. Problems with data availability and retrieval: SLAs will fully adhere to the business's needs. Here, the operational staff is crucial to managing system time and supervising SLAs
2. Security: To safeguard data and the virtualized Internet, apps, services, as the associated CC infrastructure, security of cloud computing, more popularly, cloud protection, includes a wide array of laws, technologies, applications, and control mechanisms. It falls under the subdomains of information security, internet security, and more generally, computer security (R. Khan and M. Ahmad, 2016).
3. Managing execution: Each stack's cores are subject to tension adjustment. Additionally, this boosts device output. Many recent figures provide an effective use of resources and a change to the stack. Cloud stacks can be created in a variety of ways, including memory, CPU, and structure stacks. The path to center point overload and subsequent store relocation to other centers is to alter the strain.
4. Load balancing: One of the main problems with the CC at the moment is load balancing, which prevents some nodes from being underutilized while others are overloaded, so the idle ones need to be put to work. Price, reaction time, dependability, effectiveness, and use of resources are some of the QoS

metrics that load balancing may enhance (R. Khan and M. Ahmad, 2016).

5. Fault tolerance: Because resource dropping affects unit performance, job outcomes, productivity, reaction time, and high quality, FT is among the most important parameters. Consequently, in order to identify errors, fix them, and improve performance metrics, a fault tolerance strategy is needed. To guarantee the continuity of the essential services and the program's completion, fault tolerance is a crucial consideration.

2 LOAD-BALANCING

One crucial technique for distributing workloads among cloud computing users is load-balancing, such as several computer resources to maximize resource use, improve performance, and guarantee system dependability.

Research environments rely on load balancing as their essential practice to accomplish tasks that combine artificial intelligence processing with large-scale simulation, big data analytics, and high-performance computing, which is essential for handling large volumes of data.

The primary objective of LB is to efficiently divide the workloads across different cloud endpoints, preventing any node from being overloaded and sometimes underloaded (J. Shah, 2017) To optimize its utilization of resources and enhance the overall reaction time, LB can be defined as the procedure for allocating a load among network links on several gadgets or groups of systems. It prevents excessive asset replication and shortens the device's overall waiting time. To distribute and process data without waiting, requests are dispersed throughout servers during this process. By shifting the device burden, LB maximizes system performance (M. A. Hossain and S. Roy, 2019)

2.1 Characteristics of Load Balancing

Distributed network traffic or computational workloads represent the technique of allocating workloads among several servers to keep any one server from becoming overloaded, which is referred to as load-balancing. The following are the main features of load balancing firewalls (A. Chaturvedi, et.al, 2019).

Evenly Traffic Distribution: To avoid bottlenecks and guarantee peak performance, incoming requests

or workloads are effectively distributed among available resources.

- 1) Excellent Availability and Dependability: The system offers superb reliability alongside dependability because it redirects system traffic to functional servers to maintain service accessibility when servers fail (A. Jain and R. Kumar, 2016)
- 2) LB Scalability: The scalability parameter of LB allows administrators to dynamically add or remove servers for traffic demand variations.
- 3) Failover and Fault Tolerance: The system ensures service continuity through the detection of broken elements by remedying traffic diversion to functional servers.
- 4) Conservation persistent (Sticky Session): The system needs sticky sessions to maintain user requests within a single server throughout their connection. IP hashing and cookies, together with other methods, operate at load-balancing to achieve this goal (S. Afzal and G. Kavitha, 2019).

2.2 Challenges of Cloud Computing's Load Balancing

LB is among the most pressing issues that require particular attention out of all the difficulties that cloud computing faces. This covers topics like VM security and migration; user comfort with QoS and source usage are given equivalent weight when looking for a better way to increase cloud-based resource utilization. An inventory of some load-balancing problems is provided beneath. Table 1 shows the The Load-Balancing Challenges Overview Is Presented.

- 1) Distributed Location-Based Nodes: To compute at different places, cloud data centers are usually dispersed. These centers use a centralized network of dynamically distributed nodes to process consumer requests efficiently. There are several load-balancing techniques with a narrow scope that ignore factors like network and communication latency, the distance between distributed computing nodes, customer space, and resource availability. It is difficult to operate nodes in extremely remote locations since more algorithms are not appropriate for this setting (P. Kumar, 2019)
- 2) A single failure instance: Certain algorithms used for load-balancing are put forth in the literature in situations where the centralized node makes load-balancing decisions rather

than decision-making being divided among several nodes. The entire computer system will be impacted if the main components fail.

- 3) VM Mobility: Several virtual computers can be constructed on a single physical unit, thanks to virtualization. These VMs are autonomous in their architecture and have many configurations. It is suitable to move all virtual machines (VMs) to a distant site using the LB approach if a physical device is overwhelmed.
- 4) Hypothesis for Perception: The authors are making load balancing in the cloud a homogenous node in the original question. A switch that is dynamic is needed by CC consumers, whose execution needs to be done on heterogeneous nodes to have the most effective network and reduce the response time.
- 5) Data Handling: Old conventional storage devices, especially hard disks, always required massive resources and equipment costs for hardware; this CC addressed. Consumers can keep the data safely and evenly with the help of the cloud without any control problems. Storage is ever-growing and demands in turn, redundancy of stored data to maintain access and data availability.
- 6) Scalability: With cloud services of on-demand scalability accessibility, people have the opportunity to access resources to downscale and downscale rapidly at any time or scale up. A good load balance must adapt quickly to variability in the computational environment in terms of, for example, changing requirements and conditions, memory and device topology, and so on.
- 7) The Intricacy of the Algorithm: Algorithms for cloud computing ought to be fast and simplistic to accomplish. A stronger analysis technique seeks to decrease the efficiency of cloud systems and to excellence.
- 8) On-demand Self-Service: Among the most important features that are related to CC is scalability; materials could have unsupervised provisioned or disseminated. So, how do we apply or disassociate cloud computing and its offerings, retaining likewise efficiency like traditional systems as well as the greatest resource (R. Khan and M. Ahmad, 2018)?
- 9) Control of Energy: The economy of the scale is the advantage of cloud usage in energy management. In the final analysis, saving on

power is the crucial element that renders a worldwide economy possible, in which limited companies will contribute to the pool

of international capital rather than separately supplying their utilities (R. Khan and M. Ahmad,2018).

Table 1: The Load-Balancing Challenges Overview is Presented.

Reference	Title	Challenges	Description
P. and R. Kumar (2019)	A survey of the problems and difficulties with cloud computing LB techniques	Distributed Geographical Nodes	Usually, cloud data centers are dispersed to accommodate computing in various places.
		Single Instance of Failure	The centralized node makes LB decisions; decision-making is not dispersed among several nodes.
		VM Mobility	Building several virtual machines on a single physical unit is made possible by virtualization.
		Hypothesis for Perception	For a network to be effective and response times to be shortened, CC consumers require a dynamic switch
Rafiqul and M. Zaman Oqail	A survey of LB difficulties in CC	On-Demand Service	Flexibility is cloud computing's primary characteristic; resources can be distributed or delegated automatically.
		Energy Management	The economies of sales are the advantages of energy management, which promotes cloud use.

2.3 Parameters of Load Balancing

In this work, some performance parameters are added to the existing load balance parameters. Additionally, new parameters may be added to the classification based on their attributes if they are discovered in the future. The cloud load-balancing parameters are divided into four sets by taxonomy (R. Kaur and N. Ghumman,2017). Table 2 shows the An Overview of The Load-Balancing Parameters.

- 1) Load balancing metric with dependent nature and qualitative characteristics.

- 2) Load balancing metrics that are independent and have qualitative characteristics.
- 3) Load balancing metrics that are dependent and have quantitative characteristics.
- 4) Load-balancing metrics that are independent and have quantitative characteristic

Table 2: An Overview of the Load-Balancing Parameters.

References	Title	Parameters	Descriptions
A. Jain & R.Kumar (2016)	A cloud environment multi-stage load-balancing method	Fault Tolerance	The algorithm's capacity to handle error situations and its resilience to failure.
R. Kaur & N. S. Ghumman (2018)	Task-Based LB through effective cloud computing VM use	Reaction Time	It is determined by deducting the completion duration of an assignment from the task delivery beginning time.
Roy Sumendu (2019)	Performance evaluation of load balancing algorithms	Overhead	Support the additional expense of algorithm integration.
S. Afzal & K. Ganesh (2019)	A Systematic Review of a Taxonomic Classification of Load Balancing Metrics	Scalability	Possible for a device to carry out consumer operations within the intricate traffic flow..
		Performance	Improves system consistency for the efficiency parameter.

3 TECHNIQUES FOR BALANCING LOAD

In cloud computing, load-balancing strategies can be broadly divided into three categories: hybrid, dynamic, and static. These methods aid in effectively allocating workloads among several servers in order to optimize resource use, enhance performance, as well as guarantee high availability.

3.1 Static Techniques

Static load-balancing strategies don't depend on the state of the network; instead, they employ a static set of rules. This approach requires specialized familiarity with resources, including time of contact, node storage and space, node capabilities for processing, etc., and is not scalable. Although this method is fast and effective, it usually fails to locate the connected servers, which results in uneven resource allocation. The main issue with this type of approach is that decision-making does not give enough thought to the system's actual state. Therefore, distributed systems cannot tolerate a state of constant change.

The techniques for static load-balancing methods include:

- The Round Robin (RR) Method
- Minimum-Maximum (MM) Method
- Shortest Job First (SJF) Algorithm

3.2 Dynamic Techniques

The method starts its decision-making process after reviewing the present system's situation. Each of these tactics provides the benefits of shifting work from machines with heavy loads to machines with light loads (P. Kumar,2019)

The specified dynamic load-balancing methods include:

- Least Connection Algorithm
- Throttled Algorithm
- Weighted Round Robin

4 EXAMINATION OF COMPARISON IN LOAD-BALANCING APPROACHES

4.1 Load Balancing Using Round-Robin Algorithm

To handle multiple things happening at the same time, the CPU uses a method called Round Robin. It is like giving each task a tiny time slot and then just going through the list again and again, ensuring everything gets a chance to run. This section describes the algorithm's idea and how scholars have proposed utilizing it to resolve CC load-balancing concerns. Table 3 shows the Comparison Of The Enhanced

Simulation Tool Based On The Rr Algorithm's Performance.

- 1) Tailong and Dimri(2016) suggested that to change CloudAnalyst's current responsiveness of the service broker guidelines, the authors should use an Optimize Response Time Algorithm was modified (MORT). The method determines the scheduling procedure after calculating the waiting and response times for each process. Although it can decrease response time, for a dynamic cloud environment, the method is less suitable because this method didn't solve the time quantum issues in RR.
- 2) Issawi et al. (2015), the author suggested a Modified Optimize Response Researchers have enhanced the quality of service (QoS) in cloud-based apps by considering the

problem of burstiness in workload. An abrupt increase in cloud service users results in a load-balancing problem; therefore, both situations should be considered in cloud computing. To efficiently assign the tasks you've been given to heavy loads of work on VMs, an adaptive LB method (RR+Random) is suggested. This algorithm alternates between random and RR project scheduling strategies.

- 3) Pasha et al. (2014), an RR-based task-based LB approach is provided. The recommended approach is an enhanced round-robin that saves the most recent entry submitted by a used base in a hash map, which decreases the total reaction time in cloud applications (S. A. Salman and M. K. Ahmed,2021)

Table 3: Comparison of the Enhanced Simulation Tool Based on the RR Algorithm's Performance.

Author	Parameters	Tool	Response Time	Processing Time
Dimri & Tailong (2016)	Processing and response times	Cloud Analyst	151.72 ms	1.66 ms
Pasha & Associate, (2014)	Processing time and response time	CloudSim	299.91 ms	1.210 ms
Issawi & Associate, (2015)	Reaction time & Processing time	NA	369.6 ms	314.9 ms

4.2 Load Balancing Using a Throttled Algorithm

A load-balancing method known as the throttled algorithm distributes requests among virtual machines (VMs) according to their availability. Restricting the number of active requests per virtual machine guarantees effective resource use. This section describes the Throttled algorithm's idea, and in order to solve load-balancing problems in CC, scholars have proposed using it.

- 1) Phi and associates, 2018, suggested the TMA algorithm, which tends to keep two virtual machine (VM) tables with accessible and hectic designations for each VM equitable workload sharing. In contrast to the conventional Algorithm, which keeps a single database for each VM, making it more challenging to find out if a Virtual Machine (VM) is available whether or not. Reaction

Time decreased somewhat from 402.66 to 402.63 (ms) due to the algorithm. (A. Aliyu and P. Souley,2019)

- 2) Banerjee and Ghosh, 2016, suggested that according to the priority method, utilizing the PMTA, or Modified Throttled Algorithm, has a sufficiently quicker completion time above the existing method. To perform top-priority jobs at first, as well as to divide the workload evenly between multiple virtual machines, it centers on distributing tasks that arrive via a preemptive queue to halt processes with lower tiers from executing. Although the approach decreased reaction and waiting times in comparison to the existing TA as well as the round-robin algorithm, it could nevertheless lead to excessive reaction times as well as craving for less important things.
- 3) Souley and Aliyu, 2019, suggested for equitable workload distribution, TA, as well as

ESCE, the Hybrid Method is suggested, maintaining a setpoint as each virtual machine's (VMs) top priority. In addition to lowering response time, it is also reasonably

priced (S. Lian, et.al, 2017) Table 4 shows the Comparison of The Efficiency of a Controlled Algorithm-Based Simulation Tool.

Table 4: Comparison of the Efficiency of a Controlled Algorithm-Based Simulation Tool.

Algorithm	Tools	Response Time	Processing Time
TMA	CloudAnalyst Tool	402.6 ms	173.0 ms
Priority-based altered throttled algorithm (PMTA)	CloudSim Tool	0.06 ms	1.32 ms
Evaluation of a hybrid approach's performance (TA+ESCE)	-----	315.68 ms	7.57 ms

4.3 Load Balancing Using Machine Learning

The goal of machine learning (ML), a branch of AI, is to teach systems to carry out novel tasks without explicit programming. Training is the process of using models using statistical techniques based on the past data to produce models that can predict previously unknown values. By guaranteeing service quality and adherence to set SLAs, an intelligent load balancer gives cloud providers a competitive edge. The intelligent models listed below were examined. Table 5 shows the An Overview Table Displaying The ML Load-Balancing Strategies That Have Been Reviewed.

- 1) S. Liang et al. suggested a load balancer to manage traffic in the data center. The quantification of load traffic was predicted using a Bayesian network, which was then integrated with reinforcement learning to incorporate a self-adjustment criterion and decide on the appropriate course of action. The method used included delocalizing processing.
- 2) J. Kumar and A. K. Singh created this way to foretell the amount of work in the data center on the cloud. This approach blends ANN with

self-adaptive differential evolution (SaDE). User requests were combined into the periods that functioned as historical information. The ANN was trained using back data and actual workload components. Future work in the data center was predicted using the resulting model. NASA and Saskatchewan server datasets were used to train the model.

- 3) A. Kaur et al. employed regression using deep learning to forecast the ongoing task timeline based on computing time and cost. Three concealed levels of convolution neural networks, the layer of activation composed of the ReLU function, and a pooling layer were all intended components of the network for profound learning. The budget, as well as schedule parameter data from larger operations, made up the training data.
- 4) A. Abbas, D. Sutter, and S. Worner employed QNN as a type of neural network built on the ideas of quantum computing. It's been discovered that quantum circuits operate similarly to ANN. The workload that the cloudlets would produce was predicted using the QNN model. To modify the qubit network weights, their prototype employed the activation function is the CNOT gate in both the output as well as hidden layers.

Table 5: An Overview Table Displaying the ML Load-Balancing Strategies that have been Reviewed.

Title	Model	Parameters / Data Used	LB-Problems
An SDN controller based on reinforcement learning	The Bayesian network & reinforcement learning	Data on network traffic	Security and stability of networks
Workload prediction	ANN & self-adaptive differential	Customer requests are sent to time units	Workload distribution
Deep learning	CNN, Regression	Job flow information	Qos, resource utilization
Quantum-based Load Balancing	Evolutional QNN (EQNN)	Workload logs for cloudlets	Distributed resource scaling

4.4 Ant Colony Optimization for Load Balancing

ACO is among the effective optimization techniques for resolving LB issues in CC. ACO's algorithm imitates the behavioural patterns of a colony of Ants looking for the second-best path between their nest as well as the source of food. method eventually converges to an ideal answer when the shortest path is strengthened over time by a greater concentration of pheromones. Some of the reviewed research papers are mentioned below. Table 6 shows the Examination of ACO Algorithm-Based Load Balancing Methods.

- 1) M. Mishra and A. Jaiswal designed an ant-based control system to address the issue of LB in cloud contexts. To increase or decrease various performance metrics, such as network load, CPU usage, large memory, or latency for clouds of varying sizes, we seek to create an efficient algorithm for LB that makes use of the simulated annealing technique. It has been demonstrated that the pheromone update is a useful tool for load balancing.
- 2) Richa Chib, Er. V. Kaur and Dr. N. Dhillon proposed a new method for evaluating an optimized load balancer performance. By altering the interval time, the provided technique estimates the necessary measures and is based on conditions that will give the client high availability. They attempted to prevent overloading in the suggested technique and under VM loading.

- 3) Zheng-Tao Wu modified the basic ACO algorithm to facilitate LB and job scheduling. The ACO algorithm can efficiently find the regional and international (global) ideal solution as well as attain the quickest possible implementation complexity time, according to their results (S. Banerjee, 2009) By modifying the pheromone formula, it will predict the shortest time for task completion.
- 4) Shagufta Khan et al. applied the SALB algorithm, studied current ACOs first, then used ACO to create an efficient algorithm for LB. Balancing the overall system load by attempting to optimize or reduce the various parameters is the primary contribution of the work (A. Kaur,2020).
- 5) Soumya Banerjee et al. provided a preliminary heuristic algorithm to implement a modified ACO approach for the cloud paradigm's various scheduling and service allocation mechanisms. The coefficient and the ACO pheromone update mechanism are changed. The likelihood of fulfilling the inquiry has, moreover, modified scheduling that has been used to converge, and this modification helps to minimize the makespan time (C. Udatha,2023).

Table 6: Examination of Aco Algorithm-Based Load Balancing Methods.

Title	Author	Method	Key Points
ACO: A Survey of load-balancing	M. Mishra and A. Jaiswal	Heuristic algorithm based on ACO	A pheromone update is a useful and efficient tool for load balancing. It is approachable for routing in the systematic network.
ACO-based LB algorithm for cloud computing	R. Chib, Er. V. Kaur and Dr. N. Dhillon	ACO algorithm to evaluate the effectiveness of the optimized load balancer	This technique avoids the situations of overloading and underloading of VMs. Useful for both static and dynamic Load balancing.
Application of ACO in cloud computing	Zheng-T Wu	Modifying ACO for task scheduling	It can find the best solution worldwide. It predicts the quickest time to finish a task.
Effective scheduling algorithm for LB using ACO in cloud computing	S. Khan et al.	Scheduling Algorithm (SALB)	Effective at locating the overfilled nodes as quickly as possible. To maintain mode balance while maximizing resource utilization and efficiency.
Cloud computing initiative using a modified ACO framework.	S. Banerjee et al	The heuristic algorithm uses modified ACO	Improved use of available resources.

4.5 Load Balancing Using Particle Optimization of Swarm (PSO)

The meta-heuristic optimization algorithm known as PSO was motivated by the social behavior of fish and birds. By mimicking the motions of a swarm of particles within the search area, it is used to find the best answers. Every particle serves as a possible solution as well as modifies its location in response to both its own as well as its neighbours' experiences. Table 7 shows the Comparison of The Enhanced PSO Algorithm-Based Simulation Tool's Performance.

- 1) Dr. A. Kaur, Dr. P. Singh, H. K. Toor, and B. Singh provided the method of heuristic optimization that is employed to improve the decentralized load-balancing technique, which distributes the load among each virtual machine (VM). Additionally, the outcomes are examined and contrasted with a centralized load balancer in terms of throughput and energy efficiency parameters
- 2) Chaitanya Udatha and Gondi Lakshneeswari suggested an optimized multi-objective PSO algorithm (LBIMOPSO) method for LB to

distribute tasks among the most appropriate virtual machines (VMs) and manage load consistently. It is a strong optimization method that efficiently balances workloads in a cloud computing environment while considering several objective functions at once.

- 3) R. M. Alguliyev et al. suggested the best way to move workload-causing tasks from virtual machines (VMs) that are overloaded to the appropriate VMs. It aims to illustrate that giving criteria weights results in a better solution.
- 4) H. K. Nayak et al. suggested a hybrid strategy for overcoming LB problems. In this method, the author combined the Dragonfly and PSO algorithms to get a better response time. It combines investigating the dragonfly algorithm and the strength of task scheduling using particle swarm optimization, the purpose of LB.

Table 7: Comparison of the Enhanced PSO Algorithm-Based Simulation Tool's Performance.

Title	Techniques	Tool Used	Response Time Improvement %
PSO-based dynamic load-balancing in the cloud environment	Decentralized LB using enhanced PSO	CLoudSim	18% Improved
Adaptive load balancing that makes use of PSO for cloud task scheduling	LB improved multiobjective PSO (LBIMOPSO)	CloudSim	50.76% Improved
PSO-cloud computing using a cloud-based LB approach	PSO-Time-Based LB Alpha-PSOTBLB	CloudSim, Jswarm	0.211s
DPSO: A hybrid LB strategy in cloud computing that combines the PSO and dragonfly (DA) algorithms	Hybrid approach (DA+PSO)	CloudSim	50% over DA 66.67% over PSO

5 EMERGING TRENDS IN LOAD-BALANCING

The future of LB is being shaped by emerging developments within the field of CC, which seeks to optimize sustainability, scalability, efficiency, as well as security. A closer look at some significant new trends is provided here

5.1 Fog Computing

By incorporating the network edge into the ecosystem of computing to enable decisions as near the data sources as possible as well as feasible, the Fog model is an expansion of the conventional cloud computing paradigm. The use of such a computing model has several advantages. Lag times between servers and users, for instance, might be decreased (M. Adhikari,et.al.2018) Table 8 shows the Fog Computing Advantages And Difficulties.

Table 8: Fog Computing Advantages and Difficulties.

Advantages	Difficulties
1. Reducing the amount of bandwidth used by the cloud.	1. Needs effective coordination between edge, fog, and cloud nodes.
2. It enhances performance for real-time applications.	2. Risks to security because of decentralized processing.

5.2 Blockchain Technology

Blockchain technology is becoming essential in several industries to speed up and optimize transactions by raising their degree of auditability, dependability, and traceability. A distributed immutable ledger that is set up in a decentralized network and depends on cryptography to adhere to security regulations makes up blockchain(M.

Adhikari,et,al.2018). Table 9 shows the Blockchain Advantages And Difficulties.

Table 9: Blockchain Advantages And Difficulties.

Advantages	Difficulties
1. It removes the main causes of failure.	1. Excessive computational overhead.
2. Increases trust and security.	2. Slower processing rates.

5.3 Machine and Deep Learning

Recently, there has been a lot of interest in AI because of the enormous volume of data produced in recent years, as well as the increase in processing power, primarily from GPUs. Researchers and practitioners of cloud computing can benefit from understanding deep learning and machine learning algorithms and prototypes (M. Adhikari and T. Amgoth,2018). Cloud environments are increasingly using DL and AL for intelligent load balancing. These methods make decisions in real time, optimize resource allocation, and forecast workloads using data-driven models.

- Predictive Analysis
- Reinforcement Learning
- Neural Networks

Table 10: AI Advantages and Difficulties.

Advantages	Difficulties
1. Dynamically adjusts to shifting workloads.	1. Training models require large datasets.
2. Enhances user experience by making the best use of available resources.	2. Computationally costly.

New developments in load balancing emphasize automation, decentralization, sustainability, and intelligence. Cloud environments are becoming more eco-friendly, scalable, and efficient thanks to technologies like AI, Edge computing, Blockchain, and Green cloud computing, which are revolutionizing conventional load-balancing techniques. (M. Adhikari, et, al.2018) Table 10 shows the AI Advantages and Difficulties.

6 FUTURE RESEARCH DIRECTION

In the first ten years of its existence, the idea behind the CC has transformed the information technology landscape, similar to the Internet, the Web, and the actual computer. Large-scale diverse sensor networks and IoT will produce enormous data streams for archival purposes, management, and analysis, as well as energy-cost-effective customized computer services that must adjust to a range of hardware gadgets while making adjustments for a number of factors. The surveyed algorithms are found to generally enhance energy protection, resource utilization, as well as quality of service. Existing LB algorithms have a number of drawbacks, including static barriers, insufficient frequency control, resource waste, and energy waste. As a result, there is much room for improvement. Therefore, new approaches that demand load-balancing according to carbon emission, energy consumption, and support costs are very promising. It is recommended that a number of meta-heuristics be tested in real-world scenarios, such as techniques that use ACO or PSO to demonstrate their potential for use in the actual cloud. To get around the shortcomings of the algorithms in use today, the following work might be accomplished in the future.

7 CONCLUSIONS

This survey has explored various cloud computing elements that guarantee effective resource use, high availability, and peak performance. Among the main issues, load balancing is the primary problem since burdening a gadget/device can have disastrous consequences and render technology outdated. Therefore, efficient resource utilization always requires the use of an LB algorithm. The primary objectives regarding LB are to fulfil user demands by allocating the workload among several network nodes, optimizing resource utilization, and increasing device efficiency. Load balancing algorithms were explained in this study, including dynamic load balancing, as well as dynamic algorithms inspired by nature. More efficient use of resources, a lower makespan, a higher extent of the mismatch, and efficient migration of tasks, as well as a shorter period, will all be made possible eventually by the requirement to develop completely self-governing dynamic LB algorithms.

The technology of CC itself has a long lifespan. Among the key innovations, we can use it to carry out

crucial business functions. Over time, the aforementioned innovations will completely improve cloud computing. Effective and clever load-balancing techniques will remain essential as cloud computing grows to meet the growing needs of big data processing, distributed applications, and cutting-edge technologies like 5G and the Internet of Things(IoT). The upcoming CC system generation will be greatly influenced by ongoing research and innovation in this region.

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