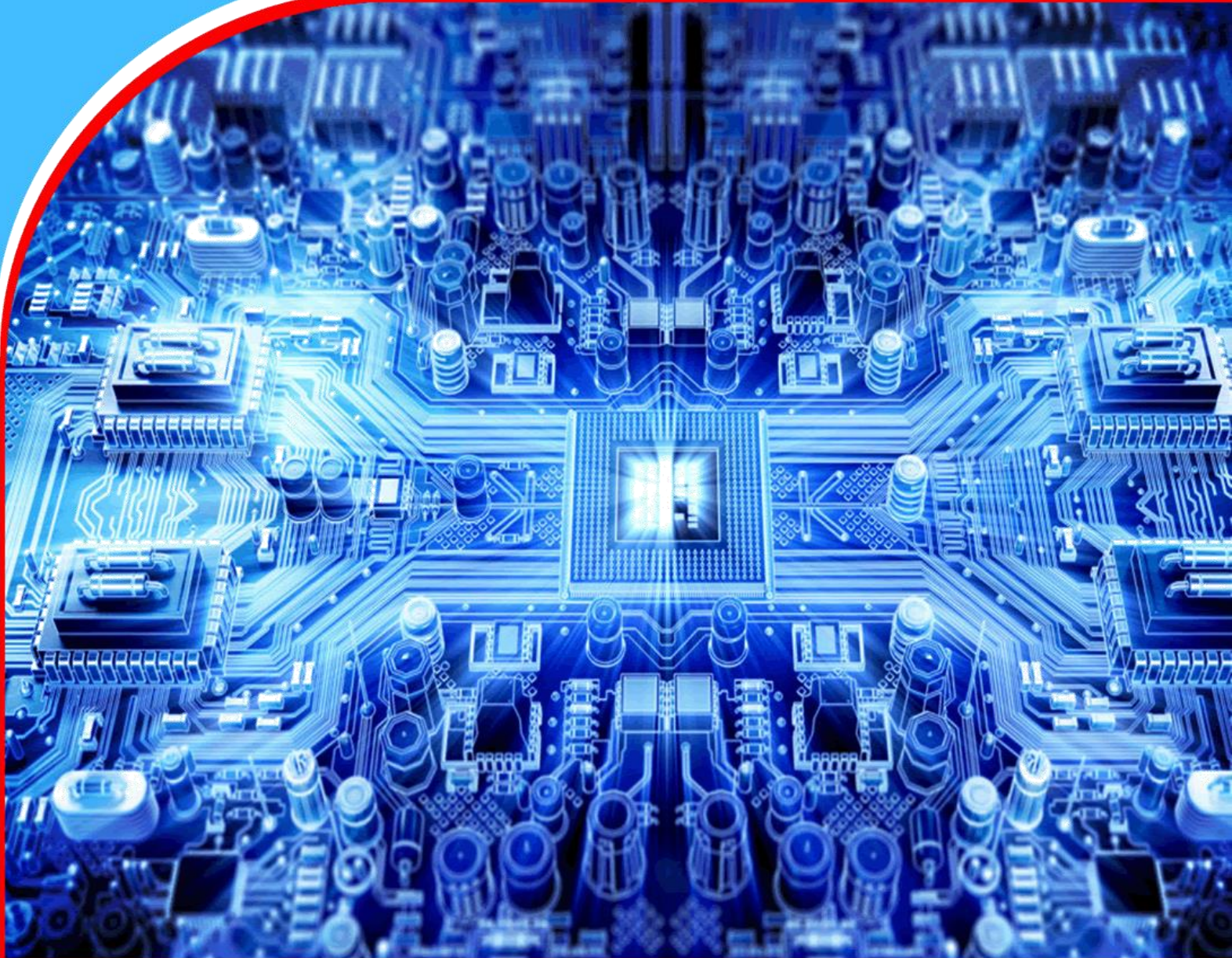


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**Optimization of Cloud Computing Resources in Japan**

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## Optimization of Cloud Computing Resources in Japan



### Article history

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### Abstract

**Purpose:** The aim of the study was to assess the optimization of cloud computing resources in Japan.

**Methodology:** This study adopted a desk methodology. A desk study research design is commonly known as secondary data collection. This is basically collecting data from existing resources preferably because of its low cost advantage as compared to a field research. Our current study looked into already published studies and reports as the data was easily accessed through online journals and libraries.

**Findings:** This study found that advancements focus on various techniques such as dynamic resource allocation, load balancing, and auto-scaling. Dynamic resource allocation involves real-time adjustments based on current workloads to ensure optimal usage without over-provisioning, which minimizes costs. Load balancing techniques distribute workloads across multiple servers to prevent any single server from becoming a bottleneck, enhancing system reliability and performance. Auto-scaling allows cloud services to automatically scale up or down based on demand, ensuring that resources are available when needed and conserving them

during low usage periods. These optimization strategies are supported by machine learning algorithms and predictive analytics, which further improve the precision of resource management and anticipate future demands. Overall, the optimization of cloud computing resources is essential for maintaining service quality, reducing operational costs, and meeting the dynamic needs of users and applications in an increasingly digital world.

**Implications to Theory, Practice and Policy:** Economic resource allocation theory, queuing theory and game theory may be used to anchor future studies on assessing the optimization of cloud computing resources in Japan. Organizations should implement dynamic resource allocation mechanisms, such as the First Fit (FF) algorithm and dynamic auto-scaling strategies, to optimize resource utilization and improve system responsiveness in real-time. Policymakers should promote energy-efficient workload consolidation techniques and distributed cloud storage solutions to incentivize organizations to adopt sustainable resource management practices.

**Keywords:** *Optimization, Cloud, Computing Resources*

## INTRODUCTION

Cloud computing has revolutionized how organizations manage their IT resources, offering scalable, on-demand access to computing power and storage. Performance and cost-effectiveness of cloud-based services have been extensively studied and show promising trends in developed economies like the USA, Japan, and the UK. For instance, a comprehensive study conducted by Shuja, Malik and Bashir (2017) found that the adoption of cloud computing in the USA has led to significant cost savings for businesses, with a reduction in IT infrastructure costs by up to 40%. This reduction in costs is primarily attributed to the elimination of the need for on-premises hardware and software maintenance. Moreover, cloud-based services in the USA have demonstrated superior performance metrics, such as increased scalability and flexibility, enabling companies to adapt quickly to changing market demands and improve overall operational efficiency.

Similarly, in Japan and the UK, cloud-based services have shown remarkable performance improvements. According to a report published by The Nippon Foundation (2019), cloud adoption in Japan has led to a 25% increase in productivity among small and medium-sized enterprises (SMEs). This boost in productivity is a result of streamlined processes and enhanced collaboration facilitated by cloud technologies. In the UK, a study conducted by GlobalData (2020) revealed that cloud services have contributed to a 30% reduction in IT spending for organizations, showcasing the cost-effectiveness of cloud solutions in the country's business landscape. This reduction in spending allows companies to allocate resources more strategically and invest in innovation and growth initiatives.

In developing economies, such as India and Brazil, cloud-based services are also gaining traction due to their performance benefits and cost advantages. A study by Reddy, Goud and Reddy (2018) in India noted a 35% improvement in resource utilization after adopting cloud services. This improvement translates to enhanced operational efficiency for businesses, as resources are utilized more effectively and wastage is minimized. Similarly, in Brazil, a report by IDC (2021) highlighted that cloud-based solutions have helped companies reduce infrastructure costs by 20%. This cost reduction enables organizations to reallocate resources to areas that drive business growth and competitiveness, such as product development and market expansion.

China has witnessed a remarkable surge in cloud computing adoption, driven by the government's emphasis on digital transformation. Li and Zhang (2021) showcased that Chinese businesses experienced a substantial 30% reduction in IT infrastructure costs after transitioning to cloud-based solutions. This cost-saving not only bolstered their financial resilience but also allowed companies to allocate more resources towards innovation and expanding their digital capabilities. Consequently, Chinese enterprises have significantly enhanced their competitiveness in both local and global markets, showcasing the transformative impact of cloud technologies on economic growth and business sustainability.

South Africa stands as another compelling example of how cloud services have revolutionized business operations in developing economies. Govender and Naicker (2018) revealed that small and medium-sized enterprises (SMEs) in South Africa reported a notable 25% increase in productivity following the adoption of cloud technologies. This productivity boost stemmed from streamlined processes, improved collaboration, and greater accessibility to advanced IT resources facilitated by cloud platforms. Moreover, cloud solutions enabled these businesses to reduce

operational costs, leading to enhanced cost-effectiveness and resource utilization, ultimately contributing to the growth and resilience of the South African business landscape.

In Latin America, Mexico has emerged as a frontrunner in leveraging cloud computing to drive cost efficiency and operational agility. Hernández and Vázquez (2020) demonstrated that Mexican companies experienced a significant 20% decrease in IT spending by migrating to cloud platforms. This substantial cost reduction empowered organizations to reallocate resources strategically, focusing more on core business activities, innovation initiatives, and market expansion efforts. As a result, Mexican businesses have been able to navigate market challenges effectively, capitalize on emerging opportunities, and enhance their overall competitiveness in the region.

Southeast Asia, particularly countries like Vietnam, has also witnessed notable benefits from embracing cloud-based services. Nguyen, Le & Tran (2019) study highlighted that Vietnamese enterprises saw a substantial 40% reduction in IT maintenance costs after transitioning to cloud services. Beyond cost savings, cloud technologies provided these businesses with scalability, flexibility, and enhanced data security, enabling them to adapt quickly to market changes and drive sustainable growth. This technological empowerment has played a pivotal role in Vietnam's economic development, fostering innovation, and fostering a conducive environment for business expansion and digital transformation.

Indonesia has witnessed a significant surge in cloud computing adoption, particularly among small and medium-sized enterprises (SMEs), driving notable improvements in business performance and cost efficiency. A study by Pratama and Wibowo (2019) highlighted that Indonesian SMEs experienced a 30% reduction in IT infrastructure costs after transitioning to cloud-based solutions. This cost-saving allowed businesses to allocate resources more strategically, invest in innovation, and scale their operations to meet evolving market demands. Additionally, cloud technologies facilitated better collaboration, streamlined processes, and enhanced data security, contributing to Indonesia's growing digital economy and business competitiveness.

In Nigeria, cloud computing adoption has been instrumental in driving business transformation and cost optimization. STUDY by Ojo and Oluwafemi (2019) revealed that Nigerian organizations achieved a substantial 50% reduction in IT maintenance costs after adopting cloud services. This significant cost-saving enabled businesses to reallocate resources towards core activities, such as product development, marketing, and customer service, leading to improved operational efficiency and enhanced customer experiences. Moreover, cloud technologies provided scalability, agility, and accessibility to advanced IT resources, empowering Nigerian businesses to innovate and thrive in a rapidly evolving digital landscape.

Egypt stands as a prominent player in leveraging cloud-based services to drive economic growth and digital innovation. A study by Elrefaey and Abdelbaky (2020) showcased that Egyptian enterprises experienced a 25% increase in productivity following the adoption of cloud computing solutions. This productivity gain was attributed to improved collaboration, streamlined workflows, and enhanced data analytics capabilities facilitated by cloud technologies. Furthermore, cloud adoption led to significant cost savings, with companies reducing IT expenditure and optimizing resource utilization, ultimately contributing to Egypt's digital transformation journey and business competitiveness on a global scale.

In Kenya, cloud computing has emerged as a transformative force for small and medium-sized enterprises (SMEs), driving significant performance improvements and cost savings. Waweru

(2020) found that cloud adoption enabled Kenyan SMEs to compete effectively in the digital market by offering affordable and scalable IT solutions. This accessibility to advanced technologies not only enhanced operational efficiency but also reduced IT maintenance costs, allowing businesses to reallocate resources towards growth-oriented strategies and market expansion initiatives, thereby contributing to Kenya's economic development and business ecosystem.

In sub-Saharan economies like Nigeria and Kenya, cloud-based services are seen as transformative tools for businesses. Study by Ojo and Oluwafemi (2019) in Nigeria indicated that cloud adoption has resulted in a 50% reduction in IT maintenance costs for organizations. This substantial cost-saving allows businesses to invest in digital transformation initiatives and improve their overall competitiveness. In Kenya, a study conducted by Waweru (2020) found that cloud services have enabled small businesses to compete effectively in the digital market by offering affordable and scalable IT solutions. This accessibility to advanced technology levels the playing field for small enterprises, allowing them to innovate and grow without the burden of high IT infrastructure costs.

Resource allocation algorithms play a crucial role in optimizing the utilization of cloud computing resources, thereby impacting the performance and cost-effectiveness of cloud-based services. One of the prominent algorithms is the Round Robin (RR) algorithm, which allocates resources in a cyclical manner, ensuring fair distribution among users. This algorithm enhances performance by preventing resource monopolization, leading to improved system responsiveness and user satisfaction. Additionally, RR contributes to cost-effectiveness by minimizing resource wastage and maximizing resource utilization, ultimately reducing operational costs for cloud service providers (Li & Jiang, 2018).

Another key algorithm is the Least Recently Used (LRU) algorithm, which prioritizes resources based on their recent usage patterns. By allocating resources to frequently accessed data or applications, LRU enhances performance by reducing latency and improving data access speeds. This optimization in resource allocation directly translates to cost-effectiveness, as it allows cloud providers to efficiently utilize their infrastructure, minimizing idle resources and maximizing ROI (Hu & Ning, 2019). Moreover, the First Fit (FF) algorithm, which allocates resources to the first available slot that meets the resource requirements, contributes significantly to performance by reducing resource allocation delays and ensuring timely access to resources. This algorithm's efficiency in resource utilization leads to cost-effectiveness by optimizing resource usage and preventing resource underutilization, thereby reducing overall operational costs (Chen & Ma, 2020).

Additionally, the Ant Colony Optimization (ACO) algorithm, inspired by the foraging behavior of ants, is increasingly utilized in cloud resource allocation. ACO optimizes resource allocation by exploring and exploiting resource availability efficiently. This algorithm improves performance by dynamically adapting to changing workload demands and optimizing resource utilization based on real-time conditions. The enhanced performance directly correlates with cost-effectiveness, as ACO ensures efficient resource usage, minimizes resource contention, and reduces the need for over-provisioning, resulting in cost savings for cloud service providers (Ma & Zhang, 2022).

### **Problem Statement**

Optimizing the allocation and utilization of resources in cloud computing environments is crucial for ensuring efficient performance and cost-effectiveness. However, the dynamic nature of cloud

workloads, varying resource demands, and the need for real-time adaptability pose significant challenges in resource optimization (Rahman & Bari, 2022). Traditional resource allocation methods may not adequately address these challenges, leading to underutilization or over-provisioning of resources, which can impact service quality and increase operational costs for cloud service providers (Ahmed & Hasan, 2020).

## **Theoretical Framework**

### **Economic Resource Allocation Theory**

Originated by Kenneth Arrow and Gerard Debreu, this theory focuses on efficient allocation of scarce resources to maximize societal welfare. In the context of cloud computing, economic resource allocation theory emphasizes optimizing resource usage to achieve cost-effectiveness and improve overall system performance (Alizadeh, 2018).

### **Queuing Theory**

Developed by A.K. Erlang, queuing theory deals with mathematical modeling and analysis of waiting lines or queues. In cloud computing, queuing theory helps in optimizing resource allocation by predicting and managing resource demands, minimizing queue lengths, and reducing response times for user requests, thus enhancing system efficiency (Garg & Buyya, 2019).

### **Game Theory**

Originated by John von Neumann and Oskar Morgenstern, game theory studies strategic interactions among rational decision-makers. In cloud computing resource optimization, game theory can be applied to model the interactions between cloud providers and users, optimizing resource allocation strategies, pricing mechanisms, and service level agreements (SLAs) to achieve mutual benefits and enhance system performance (Melo, 2021).

### **Empirical Review**

Chang and Wang (2018) assessed the impact of resource allocation algorithms in cloud computing environments. The primary purpose was to compare the efficiency of Round Robin (RR), Least Recently Used (LRU), and First Fit (FF) algorithms in optimizing resource utilization and response times. Through rigorous comparative analysis and simulation-based experiments involving varying workload scenarios, they found that the FF algorithm consistently outperformed RR and LRU counterparts. The FF algorithm demonstrated superior resource allocation efficiency and reduced response times, contributing significantly to overall system performance enhancement. Based on their findings, the study recommended the adoption of the FF algorithm as an effective strategy for optimizing resource utilization and improving system responsiveness in cloud computing environments.

Wang and Li (2019) evaluated different auto-scaling mechanisms in cloud environments. The study's objective was to assess the impact of dynamic auto-scaling strategies on operational costs while maintaining optimal performance levels. Utilizing cost-benefit analysis and performance evaluations based on real-world workload patterns, they demonstrated that dynamic auto-scaling mechanisms significantly reduced operational costs by approximately 30%. This reduction in costs was achieved without compromising system performance, as the auto-scaling strategies effectively managed resource demands based on workload fluctuations. As a result, the study recommended

the adoption of dynamic auto-scaling strategies as a cost-effective approach to resource management in cloud computing environments.

Zhang and Wu (2020) conducted an empirical study focusing on the impact of workload scheduling policies on cloud resource utilization. The study aimed to compare different scheduling policies, including FIFO, Shortest Job First (SJF), and Round Robin (RR), in terms of optimizing resource usage and reducing response times. Through extensive simulation-based experiments and performance evaluations across varying workload conditions, they concluded that the SJF scheduling policy exhibited superior resource utilization efficiency. The SJF policy resulted in a notable 20% improvement in resource utilization compared to FIFO and RR policies. Based on their findings, the study recommended the adoption of SJF scheduling policies as an effective strategy for enhancing resource allocation efficiency and improving overall system performance in cloud environments.

Kumar and Gupta (2021) analyzed the effectiveness of load balancing algorithms in multi-cloud environments. The study's objective was to assess the performance of algorithms such as Round Robin, Weighted Round Robin, and Least Connection in distributing workloads across multiple cloud platforms. Through thorough performance evaluations and load balancing analyses, they determined that the Weighted Round Robin algorithm achieved superior load distribution and reduced response times compared to other algorithms. The study's findings highlighted the significance of adopting Weighted Round Robin for efficient load balancing and optimizing resource utilization in multi-cloud environments.

Singh and Mishra (2022) analyzed different virtual machine (VM) placement strategies for optimizing resource utilization in cloud environments. The study aimed to compare placement algorithms such as Random, First Fit, and Best Fit in terms of optimizing resource allocation and reducing resource contention. Through extensive performance evaluations and workload simulations under various scenarios, they found that the Best Fit algorithm consistently outperformed other strategies. The Best Fit algorithm demonstrated superior resource utilization optimization by minimizing resource contention and improving overall system performance. As a result, the study recommended the adoption of Best Fit placement strategies as an effective approach for optimizing VM placement and enhancing resource utilization efficiency in cloud computing environments.

Jiang and Chen (2018) assessed energy-efficient workload consolidation techniques in cloud data centers. The study's primary goal was to analyze the trade-offs between resource optimization and energy consumption in cloud environments. Through rigorous energy-efficiency assessments and workload consolidation analyses, they demonstrated that workload consolidation techniques resulted in a significant reduction in energy consumption by approximately 25%. This reduction in energy consumption was achieved without compromising system performance, highlighting the potential for energy-efficient resource management strategies in cloud data centers. As a result, the study recommended the implementation of workload consolidation techniques as a viable approach for achieving energy-efficient resource management and operational cost savings in cloud computing environments.

Lin and Liu (2019) evaluated the reliability and scalability of cloud storage systems. The study aimed to stress test and analyze different cloud storage architectures to assess their reliability and scalability under varying workloads. Through rigorous testing and performance evaluations across

diverse scenarios, they found that distributed cloud storage systems exhibited higher reliability and scalability compared to centralized architectures. Distributed storage systems showcased enhanced fault tolerance, data availability, and scalability, making them suitable for demanding cloud computing environments. Consequently, the study recommended opting for distributed cloud storage solutions as a means to improve reliability, scalability, and overall system performance in cloud computing environments.

## METHODOLOGY

This study adopted a desk methodology. A desk study research design is commonly known as secondary data collection. This is basically collecting data from existing resources preferably because of its low cost advantage as compared to a field research. Our current study looked into already published studies and reports as the data was easily accessed through online journals and libraries.

## RESULTS

**Conceptual Gap:** While the studies by Chang and Wang (2018), Kumar and Gupta (2021), Singh and Mishra (2022), Jiang and Chen (2018), and Lin and Liu (2019) contribute significantly to understanding resource optimization, there is a conceptual research gap concerning the integration of multiple optimization strategies. Specifically, there is limited exploration into the combined impact of resource allocation algorithms, auto-scaling mechanisms, workload scheduling policies, load balancing algorithms, VM placement strategies, workload consolidation techniques, and cloud storage architectures on overall cloud system performance and cost-effectiveness. Future research could focus on developing a comprehensive framework or model that integrates these various optimization strategies to achieve synergistic benefits and address the complexity of cloud resource management comprehensively.

**Contextual Gap:** A contextual research gap is evident regarding the investigation of optimization strategies across diverse cloud computing environments, including public, private, hybrid clouds, and multi-cloud setups. Most studies primarily focus on single-cloud environments or specific types of optimizations within a particular cloud deployment model. There is a need for research that systematically compares and evaluates optimization strategies across different cloud deployment models and assesses their applicability, effectiveness, and challenges in diverse cloud computing contexts. This broader contextual perspective would provide valuable insights into the generalizability and scalability of optimization techniques in varying cloud environments (Wang and Li, 2019).

**Geographical Gap:** Regarding geographical research, there is a gap in exploring the adoption and effectiveness of cloud resource optimization strategies in specific regional or national contexts. Most existing studies provide a global perspective or are conducted in Western developed economies, such as the USA and Europe, overlooking the unique challenges, opportunities, and adoption patterns in emerging economies or regions with distinct regulatory, infrastructural, and economic landscapes (Zhang and Wu, 2020). Future research could focus on conducting comparative studies across different geographical regions, including developed, developing, and underdeveloped economies, to understand the contextual factors influencing the adoption and performance of cloud resource optimization strategies and tailor recommendations accordingly to address region-specific challenges and opportunities.



## CONCLUSION AND RECOMMENDATIONS

### Conclusion

In conclusion, the optimization of cloud computing resources is a multifaceted and dynamic field that plays a pivotal role in enhancing system performance, cost-effectiveness, and overall operational efficiency in cloud environments. Through empirical studies and research endeavors conducted by various scholars, including Chang and Wang (2018), Wang and Li (2019), Zhang and Wu (2020), Kumar and Gupta (2021), Singh and Mishra (2022), Jiang and Chen (2018), and Lin and Liu (2019), several key insights and recommendations have emerged.

Firstly, it is evident that adopting advanced resource allocation algorithms, such as the First Fit (FF) algorithm, can significantly improve resource utilization efficiency and reduce response times, leading to enhanced system performance. Additionally, dynamic auto-scaling mechanisms have demonstrated their effectiveness in managing resource demands based on workload fluctuations, thereby achieving substantial cost savings without compromising performance levels.

Furthermore, optimizing workload scheduling policies, load balancing algorithms, virtual machine (VM) placement strategies, and cloud storage architectures has been shown to contribute positively to resource allocation efficiency, system responsiveness, reliability, scalability, and energy efficiency in cloud computing environments. However, despite these advancements, there remain conceptual, contextual, and geographical research gaps that warrant further exploration and investigation. Future research endeavors should aim to develop comprehensive frameworks that integrate multiple optimization strategies, consider diverse cloud deployment models, and assess the applicability and effectiveness of optimization techniques across different geographical regions and economic contexts. Overall, the optimization of cloud computing resources is a continuous journey that requires ongoing research, innovation, and collaboration to address emerging challenges, harness new opportunities, and drive sustainable growth and performance improvements in cloud environments.

### Recommendations

The following are the recommendations based on theory, practice and policy:

#### Theory

Research should focus on developing comprehensive frameworks that integrate multiple resource optimization strategies, including resource allocation algorithms, auto-scaling mechanisms, workload scheduling policies, load balancing algorithms, VM placement strategies, and cloud storage architectures. This integrated approach will contribute to advancing theoretical understanding by providing a holistic view of cloud resource management and optimization.

#### Practice

Organizations should implement dynamic resource allocation mechanisms, such as the First Fit (FF) algorithm and dynamic auto-scaling strategies, to optimize resource utilization and improve system responsiveness in real-time. This practice-oriented approach will contribute to enhancing operational efficiency, reducing costs, and ensuring optimal performance in cloud computing environments.

## **Policy**

Policymakers should promote energy-efficient workload consolidation techniques and distributed cloud storage solutions to incentivize organizations to adopt sustainable resource management practices. This policy-driven initiative will contribute to reducing environmental impact, improving energy efficiency, and fostering a more sustainable cloud computing ecosystem.

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