



# CLOUD COST OPTIMIZATION FOR PROVIDERS AND USERS

Guditi Santhosh Kumar<sup>1</sup>, Dr. J Bhuvana<sup>2</sup>

Student, MCA, Jain (Deemed-to-be-University), Bengaluru, India<sup>1</sup>

Professor, MCA, Jain (Deemed-to-be-University), Bengaluru, India<sup>2</sup>

**Abstract:** The goal of this project is to provide a service mechanism that generates revenue for both the cloud provider and its numerous consumers. We analyse the interaction between the cloud service provider and its customers from a game theoretic point of view. We presume that most every user's approach is influenced by the cloud provider's plan. The goal of the cloud provider is to choose suitable servers and set up a request allocation mechanism that satisfies consumers while using less energy.

We add a controlling parameter and choose the best request allocation approach in order to approximate the server selection space. Based on the cloud provider's strategy, we create a utility function for each user that combines net profit and time effectiveness, and we aim to optimise this value. The suggested service mechanism has potential advantages for both the cloud service provider and its clients. The cloud provider may save energy expenses and boost revenue by picking the right servers and allocating requests effectively. Shorter job completion times and maximising net rewards are further advantages for users.

Our approach considers the interdependence of the strategies of the cloud provider and its users and seeks to optimize the profits of both parties. By designing a service mechanism that incentivizes users to submit requests during off-peak hours and provides appropriate server resources, the cloud provider can increase profit while providing high-quality services to its users.

**Keywords:** Cloud computing, Cloud providers, Multiple users, Game theory, Utility functions, Profit optimization, Server selection, Request allocation strategy, Energy efficiency, Time efficiency, Controlling parameter, Iterative algorithm, Numerical calculations, Service mechanism, Off-peak hours

## I. INTRODUCTION

Cloud providers generate income through service charges for the requests made by their users. When determining the per-request charge, it is essential to consider server selection and request allocation strategies, as they not only affect the cloud provider's profits but also impact the appeal of their services to potential users. Providing excess computing capacity will result in energy waste and reduce profit, while providing inadequate capacity or inefficient request allocation may lead to dissatisfied users. Rational users will choose a service that maximizes their net reward, which is the benefit received minus the payment. The urgency of tasks also affects a user's utility, meaning that completing tasks quickly generate more utility. However, it is not feasible for a cloud provider to provide enough computing resources to complete all requests in a short period due to energy and economic reasons.

Thus, users must configure their requests at different time slots to optimize their utility. Request arrivals can be approximated as a Poisson process since they are submitted randomly. The behavior of users can be analyzed as a strategic game since their decisions affect each other's payment and time efficiency. In this project, a new service mechanism will be designed to optimize the profits of the cloud provider and its multiple users. To design a new service mechanism that optimizes the profits of both the cloud provider and its multiple users, several factors must be considered. One critical aspect is to balance the computing capacity provided by the cloud provider and the number of requests from its users. The cloud provider must ensure that the computing resources are sufficient to handle the incoming requests while avoiding over-provisioning to reduce energy waste and increase profit.

Another factor to consider is the allocation of requests to servers. The cloud provider should ensure that requests are assigned to servers that can handle them most efficiently to reduce the overall response time and increase user satisfaction. The allocation strategy should be optimized to maximize the utility of each user while minimizing overall energy consumption. Furthermore, the pricing mechanism should be designed to incentivize users to submit requests during off-



peak hours when the computing capacity is underutilized. This will allow the cloud provider to optimize its resource utilization and increase profit while reducing the per-request charge for users. In contrast, users who submit requests during peak hours when computing resources are scarce should be charged a higher per-request rate. To incentivize users to submit requests during off-peak hours, the cloud provider could offer discounted rates, rewards, or credits for users who submit requests during these periods.

Additionally, the cloud provider could employ a queueing mechanism that prioritizes requests submitted during off-peak hours, allowing them to be processed faster than those submitted during peak hours. Finally, the cloud provider should provide transparent and real-time information about the status of the user's requests and the availability of computing resources. This information will enable users to make informed decisions about when to submit their requests and how much to pay for them, increasing their trust in the service and satisfaction. In conclusion, the design of a new service mechanism that optimizes the profits of both the cloud provider and its multiple users requires a balance between computing capacity, request allocation strategy, pricing mechanism, and transparent communication. By optimizing these factors, the cloud provider can increase its profit, reduce energy waste, and satisfy its users, leading to a win-win situation for both parties.

## **II. LITERATURE REVIEW**

Several studies have focused on developing service mechanisms for cloud providers and their multiple users that optimize profit and satisfaction. One popular approach is to use game theory to model the interactions between the cloud provider and its users. For example, Al-Fares and El-Sayed proposed a service pricing mechanism based on cooperative game theory that considers the capacity of the cloud provider and the demands of its users.

Other studies have investigated the energy efficiency of cloud computing systems. Chen et al. proposed a heuristic algorithm that dynamically allocates virtual machines to physical servers based on the workload, and they evaluated its energy efficiency in a simulated cloud environment. Similarly, Wang et al. developed an energy-aware resource allocation algorithm that considers the trade-off between energy consumption and user satisfaction.

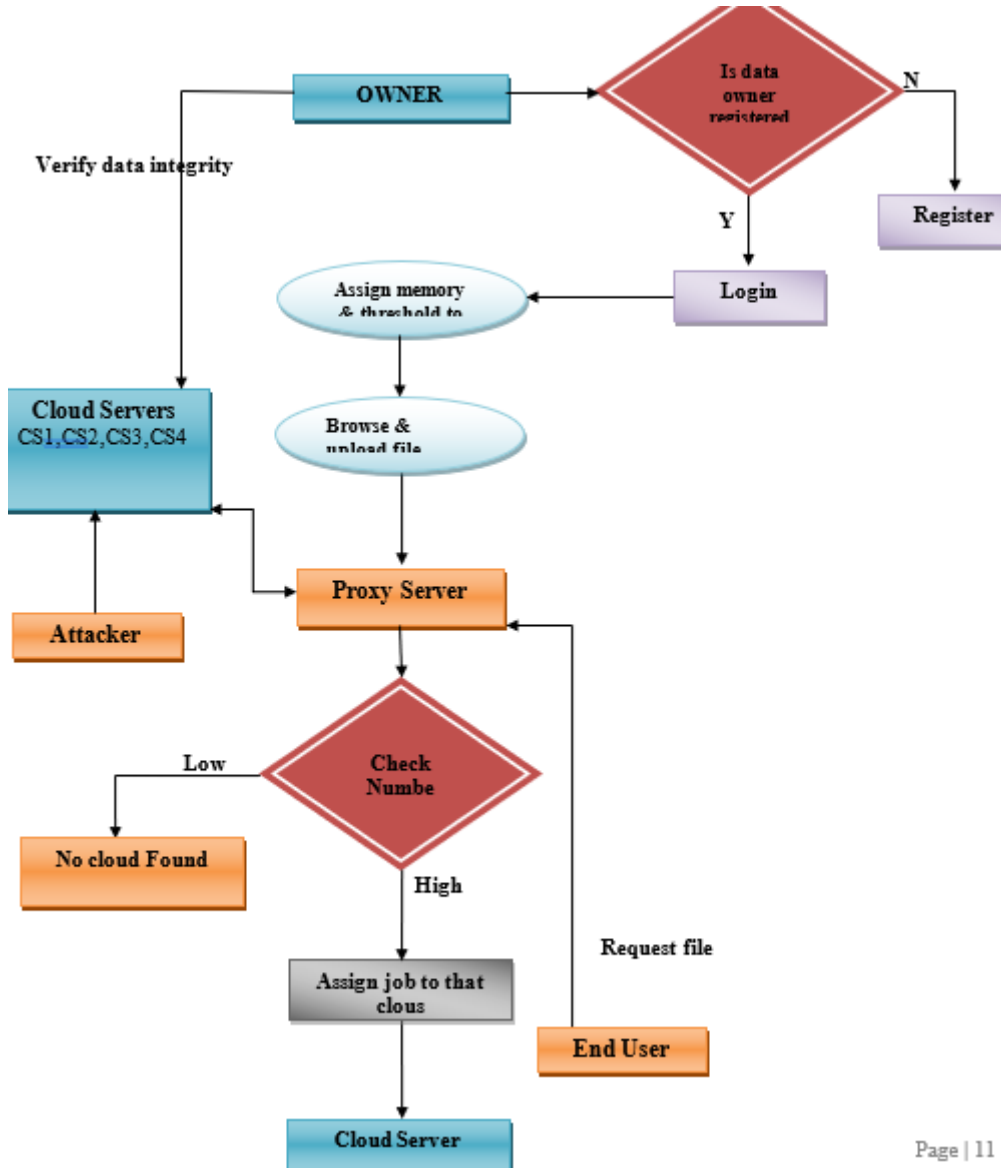
Some researchers have focused on optimizing the quality of service (QoS) for cloud users. For instance, Li et al. developed a multi-objective optimization algorithm that considers QoS metrics such as response time, throughput, and reliability. They demonstrated the effectiveness of their approach in a simulated cloud environment.

Finally, some studies have proposed hybrid approaches that consider multiple objectives, such as profit, energy efficiency, and QoS. For example, Zhang et al. developed a multi-objective optimization framework that considers the profit of the cloud provider, the energy efficiency of the cloud system, and the QoS of the users. They evaluated their approach using real-world data and demonstrated its effectiveness in achieving a balance between the different objectives.

## **III. HOW KEYBOARD WORKS**

The costs of using the cloud are reduced via a number of ways. Users and providers of cloud services can accomplish this. By integrating cost-efficient infrastructure solutions, including as open-source software and hardware, and optimising resource usage through auto-scaling, load balancing, and predictive analytics, providers can cut expenses. Spot instances can be used to access resources for less money.

Customers can reduce expenses by choosing the best price option for their needs and usage patterns, such as pay-as-you-go, reserved instances, or spot instances. Users can keep an eye on their resource utilisation and make necessary adjustments by using auto-scaling. To limit the amount of storage needed, users might employ strategies like compression and deduplication.



[ Fig. 1: Cloud cost optimization for providers and users works]

IV. RELATED WORKS

The cloud is the subject of a lot of study and industry best practises. Many studies on cloud resource allocation, architectures for effective resource management in cloud computing environments, and cost-maximization strategies for cloud computing have been published. These papers offer valuable insights and recommendations for both cloud service providers and users.

AWS Cost Optimization: 101 Techniques for Reducing Your AWS Bill" by Tensult, which provides a comprehensive list of strategies and best practices for optimizing costs on Amazon Web Services (AWS).

Cloud Cost Optimization: Techniques to Save You Money" by Cloud Academy, which provides practical advice and strategies for reducing cloud costs across various cloud platforms, including AWS, Microsoft Azure, and Google Cloud Platform.

Impact of Keyloggers

Cost maximization in the cloud can help knowledgeable users achieve substantial cost savings and align their cloud spending with their priorities. This can significantly enhance their profitability or allow them to invest in other areas of their business. In addition, greater flexibility in their cloud infrastructure enables them to respond more rapidly to shifting market conditions and client demands, thereby achieving a competitive advantage.



Cloud service providers and users play a vital role in determining the cost of cloud services. Providers can enhance their profit margins and pricing competitiveness through a variety of means. Likewise, by enhancing the performance and reliability of cloud services, providers can improve customer satisfaction and retention rates.

## V. METHODOLOGY: KEYLOGGER SYSTEM

Businesses can cut their cloud costs by adhering to a systematic approach. It's critical to create a cost optimality strategy that supports the organisational objectives, the financial limitations, and the resource consumption trends. Monitoring cloud consumption will help you find ineffective or wasteful regions. It is possible to gain visibility into resource usage patterns using a variety of methods and technologies.

Lastly, businesses should discover opportunities to use reserve instances, right size instances, or use auto-scaling to dynamically modify capacity in response to demand.

It is vital to set up transparency and oversight by creating norms and regulations for cloud cost estimation and appointing responsibilities for better efficiency. Finally, businesses should constantly evaluate their cloud cost saving plan in light of evolving customer needs and consumption trends.

In conclusion, using these guiding principles and actions can help businesses reduce their cloud costs and enhance the efficiency, flexibility, and pricing of their cloud applications.

### Keylogger Characteristics

Cloud cost optimization has several key characteristics that make it effective in reducing cloud costs while improving the performance and scalability of cloud deployments. Some of these characteristics include:

1. Flexibility
2. Scalability
3. Automation
4. Continuous Optimization
5. Data-driven
6. Collaboration

## VI. CONCLUSION

The project proposes a service mechanism for the profit maximization of cloud providers. The mechanism incorporates game theory and utility functions to model the interactions between the cloud provider and its users, allowing for the optimization of both economic and time-related benefits. A controlling parameter can be used to approximate the server selection space. The proposed iterative algorithm provides a practical and reliable way to implement the service mechanism in a real-world cloud environment. The results of the numerical calculations conducted to verify the theoretical analyses show the effectiveness and reliability of the algorithm in predicting the behaviour of the cloud provider and its users. An effective and comprehensive approach to profit maximization for cloud providers and their multiple users is a valuable contribution to the field of cloud computing.

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