A Survey on Resource Management Techniques in Cloud Computing

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Abstract: Cloud computing refers to a computer environment in which traditional software systems, installations, and licensing concerns are replaced with comprehensive on demand, "pay as you need" internet based services. In this scenario, many cloud customers can request multiple cloud resources at the same time. As a result, there should be a plan in place to ensure that resources must be prepared for the needy customer in proficient way in order complete their needs. In cloud computing systems, resource management is a critical and difficult issue. It must meet numerous service quality requirements and, as a result, reduce SLA violations. This paper survey different resource management technique for cloud infrastructures.

Keywords: Cloud, Resource management and techniques.

I. INTRODUCTION

Cloud computing is a technology in which numerous data applications, services, and infrastructures are maintained over the internet and central distant datacenters in order to supply necessities to customers. It enables users to utilize and deploy services and resources without the need for installation, and users may subscribe to different resources or access their files from any device with internet connection \([2]\). This model offers excellent computing facilities as well as ubiquitous, simple, on-demand resource access to its members. It configures a common set of computer resources, and these resources, including as services, networks, storage, applications, bandwidth, and infrastructures, may be allocated fast with right management techniques \([5, 6]\). The most fundamental element of cloud computing is that, rather of having personnel physical facilities, its subscribers and users may hire or rent resources from any cloud service provider \([8]\). Users must therefore pay for the services or resources that they have consumed. Cloud computing enables customers to get resources in a flexible and dynamic manner. The most key challenge in resource management is defining the proper number of needed resources for allocation in order to meet users' demand while simultaneously reducing costs from the users' perspective and making the optimum use of resource usage from the service providers' viewpoint. Resource management is the collective effort of interprocess communication such as resource collection, distribution, and on-demand administration to ensure the system's functionality. Consideration of Service Level Agreement (SLA) during resource management is critical. SLAs are the elementary deal among cloud users and cloud service providers \([14]\). As Quality of Service, SLA also ensures several aspects such as performance, accessibility, trustworthiness, response time, security, and energy savings (QoS). Allocation of resources may be done in both a static and dynamic way, allowing resources to be used more efficiently. This allocation should not breach any SLAs and should also adhere to the QoS requirements. High energy use, excess or insufficient resource supply must not disrupt resource management \([15]\). The primary goal of resource management may differ for customers and service providers \([17, 19]\). The goal for users is to reduce operating and maintenance costs without owning physical systems, whereas the goal for cloud service providers is to earn profits through resource allocation and management. To achieve these goals, cloud customers must advise cloud service providers about whether resources will be assigned statically or dynamically \([20]\). This information will be intake for CSPs to assure the quantity and availability of resources. Because resource allocation for various applications may change, the cloud service provider will also be able to calculate or manage resource requirements based on application.

\[\text{Resource Pools}\]

Fig. 1 Resource Management in Cloud

\[\text{Cloud Subscribers} \quad \text{Cloud Workloads}\]

\[\text{Resource Management} \quad \text{Resource Provisionor} \quad \text{Resource Scheduler}\]
Resource management includes transfer of jobs from their start to their execution and end. In the cloud, resource management works in steps. In first step, resource allocation identifies adequate resources for precise jobs on the basis of QoS requirements delivered by consumers, while resource planning plots and implements consumer jobs based on designated resources. Fig. 1 describes this process.

Remaining structure of paper contains: Section II that deals with resource allocation. Section III discusses resource allocation parameters. Section IV offers several resource management options. Section V compares several resource management approaches. Section VI contains the conclusion.

II. VARIOUS RESOURCE ALLOCATION MECHANISMS

Resource allocation may be categorized as follows based on the demands of the users and the type of requests/claims:

1) **Fixed/Static Allocation**: Fixed/static allocation is best suited for effective allocation of requests/claims with predictable and typically static requests/workloads. The consumer anticipates their resource requirements and may enter into agreements with cloud service providers [22]. The cloud service provider then prepares the necessary resources ahead of the start of the service. Billing or charges for services are due in the form of set fees or on a monthly basis.

2) **Dynamic/Active Allocation**: Dynamic/active allocation is useful when requests/claims are unexpected and may be altered, or when requests/workloads vary. In general, virtual devices may be quickly transferred by constructing new nodes in the cloud [23]. This cloud service provider may give more resources as needed and withdraw them when they are no longer required. For billing purposes, it mostly employs a "pay-as-you-go" approach.

3) **User self-allocation**: It is similar to self-service in that consumers obtain resources from any cloud service provider via any web-based service/website. For this, the consumer must sign in with an account on the website. They can order or reserve desired resources/services for the appropriate period of time after entering into the website [24]. The consumer will be charged for that resource/service.

III. PARAMETERS FOR RESOURCE ALLOCATION

A. In order to provide benefits to cloud users, costs must be kept to a minimum.

B. Profit growth should be achieved by delivering benefits to the cloud service provider.

C. In addition to low costs and large profits, the SLA parameter should be examined in such a way that its violation is minimized.

D. Power or energy consumption may be excessive during resource management and provisioning according to VM placement and migration. As a result, guarantee of decreased power consumption or low energy absorption must be taken into account.

E. If a VM or other service fails, the allocation should not be affected and the service should continue to be delivered.

F. When completing a job, the resource provisioning approaches mentioned fundamentally take little time to reply.

IV. RESOURCE MANAGEMENT APPROACHES

Cloud resources may be used and managed more efficiently by using resource management and allocation. Datacenters employ a variety of resource management methodologies, including static and dynamic management. Each has its own set of advantages and disadvantages. Management techniques are used to enhance QoS parameters [1,3,5,13], reduce costs and optimise profits [17,21,24], improve response time [2], offer services even when there is a failure [11,14,15], and improve performance [21,24,25]. reduces SLA violations [20], effectively uses cloud assets [4,6,9,10,16,19,20,22,24], and reduces power consumption [7,21].

The overall process of resource management uses elementary taxonomy of resources in cloud as shown in Fig. 2. According to previous research, minimizing execution time is a hot topic in cloud resource management. Management of appropriate resources to jobs is always a challenging task, and based on QoS requirements, recognizing the finest job and its resource combination is an significant research in the cloud.

When a workload is supplied to a resource management algorithm, it accesses the resource information, which comprises information related to all resources available at datacenters, and gathers scheduling result based on the job needs[16]. The resource management output is returned to the cloud customer via the algorithm. If the required resources are not available in line with the QoS requirement, the algorithm wishes to resubmit the workload as a SLA document with amended QoS criteria [18].

Various issues like resource distribution, ambiguity, and heterogeneity, are not addressed by classic resource management techniques. As a result, we can increase the effectiveness of services and applications of cloud by addressing these cloud environment aspects.
During management of resources, their plotting, implementation, and observation can be performed. After resource management, comes resource management. To begin, the cloud consumer sends the task to be executed. Following that, jobs are mapped to appropriate resources depending on the quality of service criteria stated by the cloud customer. Quality of Service may include throughput, CPU, memory use, and so on are often evaluated for every customer in the cloud.

V. SURVEY AND COMPARATIVE STUDY OF RESOURCE MANAGEMENT TECHNIQUES (RMT)
Survey and comparison between RMTs is quite tough task due to multiple resource scheduling techniques and also there is lack of standards. Hence, evaluation and study of RMTs is noteworthy to discover the cost and performance effective resource scheduling algorithms. Following table shows the comparison of different RMTs found in literature studied for this paper.

| Resource Management Technique | Application Used | Specific purpose of mechanism | Management Criteria | Tool Used | Challenges          |
|-------------------------------|------------------|-------------------------------|---------------------|-----------|---------------------|
| Bargaining Based RMT          | Blend of workflows | To increase user satisfaction | Bid Density         | Green Cloud Simulator | Execution time is high |
| Compromised cost and Time based RMT | Identical Workloads | To forecast future expense | Budget and deadline | Cloudsim | Not considered heterogeneous workloads |
| Dynamics and Adaptive based RMT | Computing of intensive workloads | To eliminate running time and price | Revenue maximization | Amazon EC2 | Lack of user satisfaction |
| Energy based RMT             | Identical Workloads | To reduce execution time | Execution time and Power consumption | Cloudsim | Performance degradation |
| Hybrid based RMT             | Fork and Join | Minimize cost and meet deadline | Execution time | Cloud based simulator | Budget increased |
| Nature inspired and Bio Inspired RMT | Computation workload | Minimize makespan | Cost, makespan and degree of imbalance | Cloudsim | Time complexity increased |
| Deadline-driven RMT          | Scientific workloads | Efficiently allocation of resources | Reduces execution time. | CloudAnalyst | Not for high data demanding applications |
| Active provisioning in many tenant RMT | Dynamic workflows | Service includes multiple tenants | Cloud based simulation | Aneka | No real world testing |
| Lightweight Approach RMT     | Composite or random workloads | Flexibility and resource efficiency. | Elastic Application Container | Test bed | Not suitable for web applications |
| Failure-aware RMT            | Hybrid workloads | Quality of service improved | Able to improve the users’ QoS | Java based simulator | Not able to run real experiments |
| Profit based RMT             | Composite workloads | To fulfill SLA | Processing time and communication cost | Monte Carlo Simulator | Only support for single tier applications |
| Priority based RMT           | Composite service applications | To reduce processing time and improved revenue | Priority and processing time | Java based simulator | Cost/SLA is not considered |
| Time based RMT              | Deadline constrained workload | To meet deadline | Improve data transfer, computational cost and network bandwidth | Java based discrete time simulation | Performance degrade |
VI. CONCLUSION

The results of this research study have been examined in different of methods, including resource categorization and resource scheduling evolution. It is also difficult to determine the appropriate workload-to-resource mapping without an effective resource provisioning strategy.

The following facts can be deduced:

1) If resources are booked in advance, the cost of the given cloud service may be decreased.
2) A comparison and valuation of RMTs in the cloud can help to preference the scheduling algorithm based on the job's QoS essentials.
3) Allocating resources depending on workload type (homogeneous or heterogeneous) might increase resource usage.

The appropriate matching between workload and resource can greatly boost performance. It is difficult for providers to precisely determine the quantity of resources necessary for a particular task. We think that this survey will be useful to researchers of cloud resources scheduling.

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