

The Fuzzy Grounded Ant Colony Optimization Scheduling in Cloud Computing

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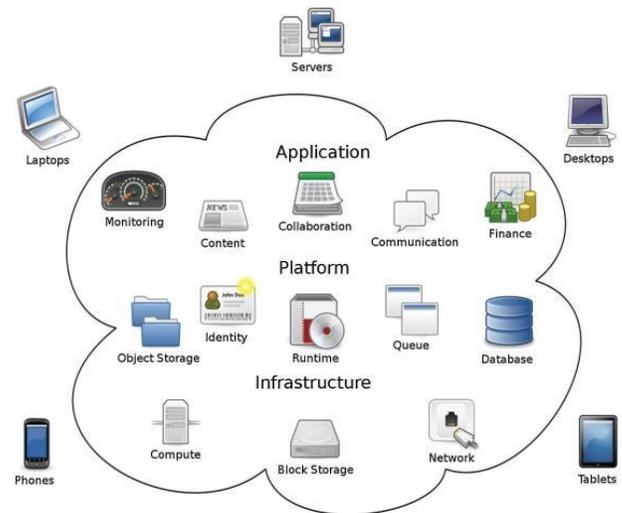
Abstract

Cloud computing is an Information Technology deployment mannequin installed on virtualization. Task scheduling states the set of rules for mission allocations to an exact digital computing device in the pall calculating terrain. still, mission scheduling challenges comparable as greatest undertaking scheduling overall performance results, are addressed in pall computing. First, the pall computing performance due to challenge scheduling is bettered via proposing a Dynamic Weighted Round-Robin algorithm. This endorsed DWRR algorithm improves the mission scheduling overall performance via considering aid capabilities, mission precedence's, and length. Second, a heuristicalgorithm referred to as Hybrid flyspeck mass analogous Ant Colony Optimization is proposed to damage the challenge prosecution detention hassle in DWRR rested venture scheduling. In the end, a fuzzy feel device is designed for HPSPACO that expands mission scheduling in the pall terrain. A fuzzy machine is proposed for the indolence weight replace of the PSO and pheromone trails replace of the PACO. therefore, the proposed Fuzzy crossbred speck mass analogous Ant Colony Optimization on pall computing achieves bettered task scheduling with the aid of minimizing the prosecution and staying time, machine proliferation, and maximizingresource operation.

INTRODUCTION

Cloud computing is a fast-growing technology that allocates dispensed present day computing structures and resources to hardware and software, permitting for extra efficient aid utilization. Cloud computing discharges with minimal downtime efficiently. Cloud computing possesses dynamic provisioning, and this technique is now not only applicable for the cloud service, however it can also compute the capability, storage, networking, and information technological know-how infrastructure. Cloud services are useful to every body the usage of a digital wallet, with the framework helping the growing

and diminishing software program in response to requests and charges for the length at some stage in which these tools have been used. Fig. 1 depicts the extraordinary sorts of interconnected computers, consisting of laptops, computer computers, mobiles, tablets, servers, and databases. Data will be stored, and packages will be run so that Internet-connected purposes will get admission to cloud data. Several research areas have been developed to categorical complex computational problems that can be efficiently



Task scheduling can't be carried out through way of focusing on a single criterion. Still, task scheduling can be carried out by means of preparations such as Quality of Service (QoS) to the consumer primarily based on the cloud carrier providers' key roles.

However, a extremely good volume of obligations is running on the cloud provider provider's role. Task scheduling ought to be idea of as a quest for the optimization to assign a sequence of subtasks from one of atype duties to the handy digital servers to reap the meant assignment scheduling goal. As a result, this paper investigates traditional undertaking scheduling

algorithms, which are accelerated the utilization of a hybrid gadget that can assist cloud choices achieve an tremendous cloud service level.

As in the Information Technology field, cloud computing is a new booming difficulty that has emerged as an authenticity. There are a few factors of the cloud that can be changed. Task scheduling is amongst the fundamental sources of concern. Since huge archives administration is turning into larger well-known in the cloud computing environment, it ought to method the records effectively. The assigning of responsibilities to versatile sources in accordance to bendy time helps decide a logical sequence in which the obligations are completed.

Assignments want to be deliberate efficaciously in a cloud computing environment to reduce response time, ready time, going for walks time, computation time, and beneficial useful resource usage. The undertaking scheduling trouble is imperative now not totally for attaining most cloud effectivity on the other hand moreover for assembly the needs of great Cloud customers in an excellent way to beautify the frequent common overall performance of cloud computing. The quintessential purpose of project scheduling in mission administration is to prioritize the duties in the cloud computing surroundings to decrease time, keep away from shedding work, and be profitable in the task's deadlines. Task scheduling improves the cloud computing device to optimize the benefit from high-performance computing and high-quality computing device performance. The scheduling algorithm distributes the workload throughout processors, maximizing their effectively whilst reducing standard task.

Dynamic Weighted Round Robin

DWRR was proposed to support VBR and CBR traffic sources by assigning an additional dynamic weighted value to each source. In this algorithm, the peak cell rate (PCR) and average cell rate (ACR) characteristics are considered.

This scheduling algorithm calculates the cycle size on the basis of VBR and CBR traffic sources. In this algorithm, ABR and UBR traffic are not considered. In addition, valid ABR cells may not be serviced because of invalid VBR and CBR cells that violate the established QoS parameters.

The ATM output port can be thought of as a train of time-slots. The time-slots in the ATM output are structured into a sequence of fixed-length time intervals called cycles. Each cycle contains a fixed number of time-slots. In each cycle, the time-slots are further divided into several rounds, as shown in Fig. 2. Each round may contain a variable number of time-slots. A circular scan on the traffic sources is done within every round. When a source is visited, a cell of this source is allowed to be transmitted by one of the time-slots in this round.

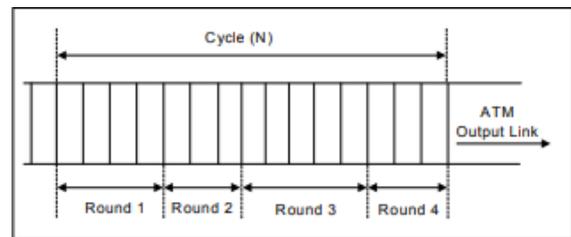


Fig. 2. The cycle and rounds in the ATM outgoing time-slots.

As shown in Table 1, the peak cell rate of a traffic source, B_p , and the average cell rate of a traffic source, B_m , have a special relation to the characteristics of each class of traffic; in AR traffic, B_m represents the minimum cell rate of a traffic source. For CBR sources, the value of B_p is set to the same value as B_m . For VBR and ABR sources, the value of B_p is set to a value larger than B_m . For best-effort sources, the values of B_p and B_m are set to 0.

Table 1. The relation between B_p and B_m according to the characteristics of traffic.

The class of traffic sources	The relation between B_p and B_m
CBR traffic	$B_p = B_m$
VBR traffic	$B_p \geq B_m$
ABR traffic	$B_p \geq B_m$
UBR traffic	$B_p = B_m = 0$

As shown in Fig. 3, the DWRR algorithm handles the VBR and CBR and the best-effort traffic sources dynamically. To assure QoS, VBR and CBR sources have to capture a guaranteed portion of the time-slots in every fixed time interval. Then, the remaining time-slots are for the best-effort traffic. The cells from each source are assumed to arrive at a random time slot within a cycle. Cells that do not arrive at the time of their assumed arrival are stored in the temporary buffer. However, they should be

transmitted as soon as possible in the next round's scan if the sources are VBR or CBR services.

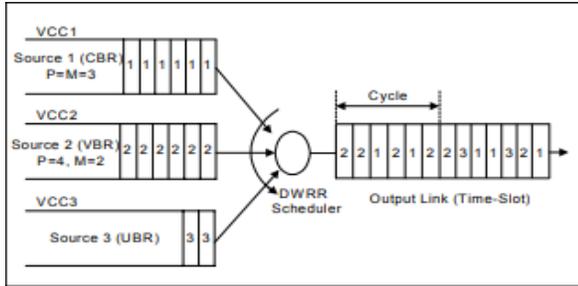


Fig. 3. DWRR cell scheduling.

The concept of the DWRR algorithm is stated as follows. Each VBR or CBR source is associated with a pair of counters, namely P and M. At the beginning of each cycle, the counter P is set to the peak cell rate BP(s) of the corresponding source s and the counter M is set to the average cell rate BM(s). The counters P and M are decreased by 1 when the source is visited and a cell is transmitted successfully.

In each round, each source falls into one of the following states:

- State 1: $P > 0, M > 0$ and there are cells stored in the temporary buffer. In this case, the source is visited in the next round.
- State 2: $P > 0, M \leq 0$ and there are cells stored in the temporary buffer. In this case, the source is not visited as long as there are other sources staying in State 1. The arriving cell is pushed into the temporary buffer.
- State 3: $P \leq 0, M \leq 0$ or no cell arrival at the moment. In this case, the source will not be visited in the subsequent rounds.

The state of each source is examined at each time-slot of the cycle to ensure that each source can perform the required state transition upon the new arrival of a cell. The algorithm selects the sources which are currently in State 1 at the beginning of the time-slot and forms a round-robin scan in the cycle. It then waits until the next time-slot and performs the same action. If there is no source in State 1, then the algorithm performs the same steps and forms a round. If there is no source in States 1 and 2, then one of the sources belonging to the best-effort type is visited and a round is formed. To ensure a fair bandwidth among the best effort sources, the visits in a cycle are scheduled in a round robin manner

Hybrid Particle Swarm Parallel Ant Colony Optimization Algorithm

To reap optimized venture scheduling in cloud computing, a hybrid PSO and PACO are proposed in this paper. Here, the hybrid algorithm combines Particle Swarm Optimization and Parallel Ant Colony Optimization [7]. For 'n' duties and 'm'

VMs, every particle represents a sensible scheduling method. PSO evaluates every particle's health fee the use of a parallel ACO algorithm and then finds particle first-class and global fantastic for the topof the line solution. Initially, PSO randomly initializes tasks, position, and velocity vector. Then PSO computes the cost, make span time, and processor utilization. In this lookup work, parameters are viewed as health values.

The health price of every particle is estimated byway of the usage of a parallel ACO algorithm. In the parallel ACO algorithm, the ant colony populace is partitioned into sub-ant colonies. Each ant sub- colony reflect on consideration on the particle satisfactory and world best. Based on the fitness value of every ant sub colonies, the particle and world high-quality of every particle in PSO are updated. The fitness price is measured by means of the weighted sum of the value of computing, make span time, and processor utilization. . It is represented in Fig. 2.

In Eq. 1, c represents the cost of computing, T is makespan, and U is processor utilization.

$$Fitness = \sum w_1 \frac{1}{C} + w_2 \frac{1}{T} + w_3 U \quad w_1 + w_2 + w_3 = 1 \quad (1)$$

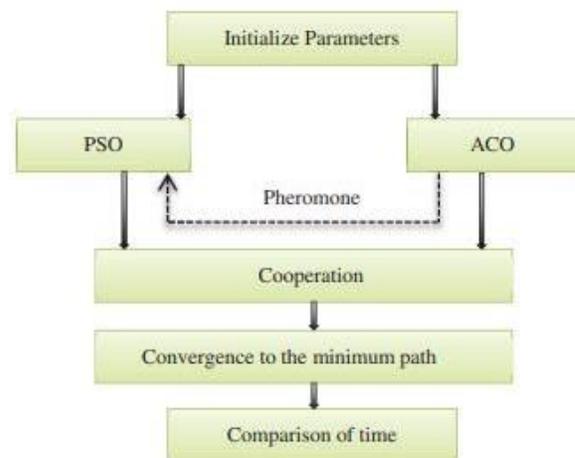


Figure 2: System Architecture of HPSPACO

Conclusion

This paper's primary objective effectively to enhance the scheduling of tasks in cloud computing, which is according to fuzzy rules. Different Meta-heuristic algorithms such as Genetic Algorithm, Particle Swarm Optimization, Ant Colony Optimization, Artificial Bee Colony, and Parallel Ant Colony Optimization are developed for task scheduling in the cloud computing environment. Using such algorithms, the tasks are scheduled efficiently in terms of minimized execution, waiting and response time, make span, and resource utilization. On the other hand, task scheduling is essential for attaining maximum cloud service efficiency and frequent cloud users' demands while also enhancing the complete cloud computing Quality of Service. As an outcome of the improved optimization algorithms used in this investigation, task scheduling performance and efficiency are improved.

Future Work

The future extension of this research would be included in optimizing a more significant number of objectives such as task availability, energy efficiency, user's comprehensive QoS, etc. Moreover, an efficient optimal task scheduling could be enhanced by novel optimization methods such as Cat Swarm Optimization, Cockroach Swarm Optimization, Bean Optimization, etc., which may provide the best and optimized results.

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