

# Efficient Cloud Server Job Scheduling using NN and ABC in cloud computing

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ABSTRACT: High work load not only translates to high operational cost, which reduces the marginal profit of Cloud-providers, which also direct towards low throughput of system. Henceforth, efficient job execution solutions are required to minimize the impact of Cloud computing on the environment. So, to generate such solutions, deep analysis of Cloud is obligatory with regard to their job-efficiency. Otherwise, Cloud computing with progressively persistent front-end client-devices interrelating with back-end data centers will cause an enormous load of jobs. To address this problem, data centre servers need to be managed in an efficient manner to drive computing. In proposed work, hybrid algorithm which is based on neural network with artificial bee colony and some concepts of scheduling has been introduced in work. On the basis of them, tasks execution will be enhanced and load will be reduced. The results evaluation will be done on the basis of total jobs completed in the MATLAB 2010a environment.

KEYWORDS: Job Scheduling, Cloud Sever, Workload, Cloud Computing, Users, Efficiency, Neural Network.

#### 1. Introduction

Cloud Computing is a pay as per usage model (e.g., gas, electricity etc.) and provide resources as per needs. The characteristics of cloud computing are rapid elasticity, measured service, on demand self-service, ubiquitous network access, location-independent resources polling [1] [2] [3] [4]. Cloud computing is a business paradigm. It is also enhancing highly use of data, resources and promote for data sharing and storing through internet. It can also focuses on scheduling job and resources, providing security for storing data. Cloud Computing is service oriented model not user oriented one.

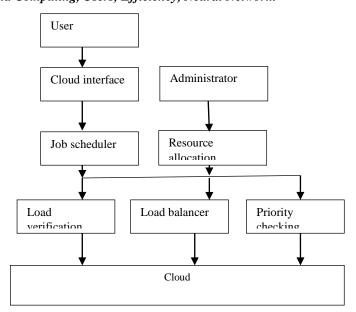


Figure 1: Scheduling in cloud Computing

Cloud computing can mainly provide four different service like Infrastructure as a service (IaaS), Software as a Service (SaaS), Platform as a Service (PaaS) and Communication as a service (CaaS). At the time of providing different services, there are several problems addressed [5] [6] [7].

Among those addressed problems, job scheduling is one of the major problems for improving job utilization. Generally scheduling can be followed for giving equal preference to jobs and promote the efficient use of resources. There are many open source tools available for developing Cloud Computing like Eucalyptus systems, open Nebula etc., Job scheduling in cloud computing is mainly focused for improving the resources utilization such as bandwidth, memory and reduce completion time [8] [9]. In the below Fig. 1, describes that nodes or servers establish

communication with each other through two different links namely variable link and fixed link.

Nodes are connected through wireless medium (i.e., Variable link). Nodes are physically connected using fixed link (wired medium). The nodes are connected through Internet [10].

## 2. A Glance of existing algorithms

Each algorithms address one or two of the parameters on the basis of which scheduling is improved. The details of each method has been shown in form of table [11-26]

Table 1: Comparison of existing algorithms

Sl.	Algorithm	Target	Criterion	Simulation	Testing	Load	Comparisons and
No.		environment		environment	Workload	Size	Comments
1.	Context	Mobile	QoS	Event Driven	Requests for	Peak	Successful for a normal
	Aware	application	improvement	Simulator	applications	load	workload as well but
	Scheduling	requests	and reduction		such as	keeps	works best for peak days.
		running on	of resource		shopping	tipping	
		cloud	wastage		assistants	during	
					and voice	work	
					assistants.	hours	
2.	Energy	Public Cloud	Reduction in	Runtime	Deadline	Large	Better in terms of energy
	Aware		soft errors	Simulation	Sensitive		efficiency as compared to
	Fault		and	Engine	Workloads.		Triple Modular Redundant
	Tolerant		maintenance				System.[3]
	Framework		of energy				
			efficiency				
3.	Green	Public Cloud	Improve	CloudSim	Generated	Copies a	Four modes are introduced
	Scheduling		Energy	and GridSim	Workloads	normal	among which prediction
	Algorithm		Efficiency		same as	day's	along with additional
	with neural		and Minimize		requests to	trend	servers is most successful
	predictor		Drop Rate.		NASA and	i.e. large	for all workload tested.
					ClarkNet	during	
					web	working	
					servers.	hours	
						and	
						small at	
						the start	
						and end.	

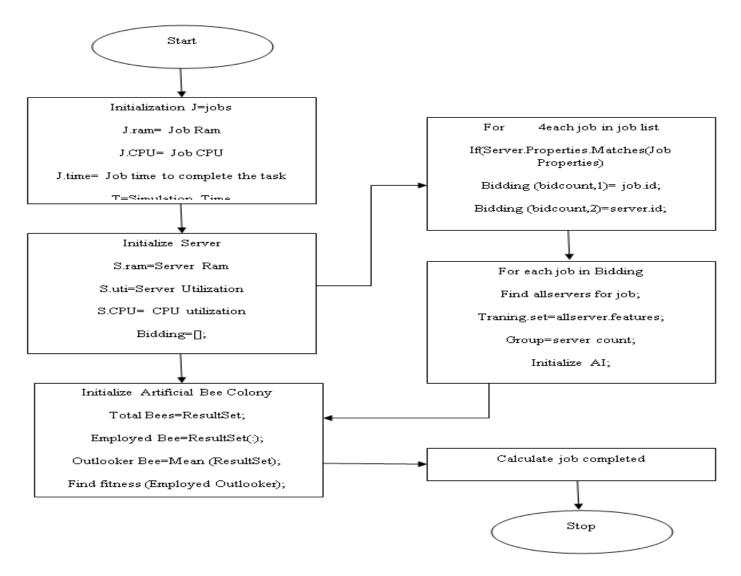
4.	EnaCloud	Public Cloud	Improve	iVic and Xen	Web server	NA	10 % more energy efficient
			Energy	hypervisor	and Data		as compared to FCFS and
			Efficiency	71	Servers		13% to Best Fit
			,		,Compute-		
					Intensive		
					,Common		
					applications		
5.	Energy	Public Cloud	Improve	Real System	Tasks which	NA	It works efficiently in case
	Efficient		Energy		don't use the		of I/O intensive tasks
	Optimizatio		Efficiency		whole of the		wherein
	n Method				hardware		CPU usage is negligible.
6.	Least first	Public Cloud	Energy	Numerical	NA	small	A higher priority is given
7.	Priority		efficiency	Study		medium	to servers with a larger
			keeping QoS				capacity.
			in mind				
8.	IC-PCP	Public Cloud	Cost		Scientific	NA	SCS and PSO are almost
9.	PSO		Minimization	CloudSim	Workflows		equally efficient in
10.	SCS		within				meeting deadlines but PSO
			Deadline				incurs a smaller cost as
							compared to SCS. Both of
							them perform better when
							deadlines are relaxed.
							ICPCP misses out on
							deadlines in most cases.
11.	Hyper-	Public Cloud	Reduction of	CloudSim	Workflow	NA	Better as compared to
	heuristic		make span	and Hadoop	and		hybrid heuristic algorithm
	Scheduling				hado		due to less computation
	Algorithm				op map-task		and case specific
							algorithm.
12.	Load	Public Cloud	Maximize	CloudSim	Computation	NA	Attains maximal utilization
	Balancing		resource		al Tasks		of available resources
	Scheduling		utilization and				
			meet QoS				
13.	Earliest	Hybrid Cloud	Cost	Java Based	Batch Type	Large	EDF overpowers FCFS as
	Deadline		Minimization	Discrete	Workloads		it successfully does the
	First (EDF)		with Deadline	Time	with		task within Deadlines
14.	FCFS		Constraints	Simulator	Deadlines	Small	taking into consideration
							the data transfer speeds and
							also handles runtime
							estimation errors
							successfully.
15.	Dynamic	Multiple	Energy	Cloud	Workloads	Large	DCMMS handles
	min-min	Clouds with	efficient	Simulation	same as		situations with large

	Scheduling	Heterogeneo	resource	Environment	requests in		workloads resource
		us Resources	scheduling	(developed	Parallel		contention occurs due to
16.	Dynamic		among	by the	Workload	Large	greater number of
	Cloud List		multicloud	author)	Archives.		Advanced Reservation jobs
	Scheduling		heterogeneou				better. Its also more
			s system				efficient in terms of
							energy.

#### 3. Simulation Work

In the figure 2, given below, we presented the proposed work in the step- wise form. Firstly, we initialized the total number of jobs J and initialised the performance parameters to all jobs like Job RAM, Job CPU utilization, Job time (to complete the task) and define the simulation time (T) of proposed work. After that, we initialized the Servers and set

performance parameters to all servers like Server RAM, Server CPU utilization, Server utilization (to complete the task) and define the bidding parameters with bidding count in proposed work. If server properties match with job properties then we create Server ID with count. For each job in bidding, we apply hybrid technique of artificial intelligence Neural Network with optimization technique like Artificial Bee Colony (ABC) optimisation technique.



90Figure 2: Proposed flowchart

We use ABC algorithm because ABC algorithm has high performance and easy for numerical optimization problems. The ABC algorithm is a recently introduced population-based meta-heuristic optimization technique inspired by the intelligent foraging behaviour of honeybee swarms with the references of AI. In ABC high numbers of objective functions are used so we can get an optimal solution according to bidding. According to proposed algorithm we complete our task and calculate the performance parameters like job completed count.

### 4. Simulation Result and Analysis

The whole simulation of work has been done in MATLAB environment in which various functions has been initialized as shown below in table.2 using neural network. In the end the proposed algorithm performance is measured using various metrics like jobs completed.

**Table 2: Simulation Table** 

Tool	MATLAB 2010a
Metrics	Min. jobs, Max jobs.
Iterations	5
Algorithm	Hybrid (ABC+NN)

The above Table 2 describes the various features used for simulation of proposed work. The tool used for the implementation of proposed work is MATLAB, the results are calculated on the basis of Minimum number of jobs and maximum number of jobs required for scheduling during resource utilization in cloud computing. The experiment is done through 5 iterations and the average results are calculated. The algorithm used for proposed work is the combination of neural network approach and artificial bee colony algorithm. At first step, the simulation environment is generated and then applies the neural approach. After that Artificial bee colony algorithm is apply on it to get the better

results. At last comparison between Proposed Hybrid approach and neural approach is taken. The hybrid approach gives better results as compare to previous neural approach.

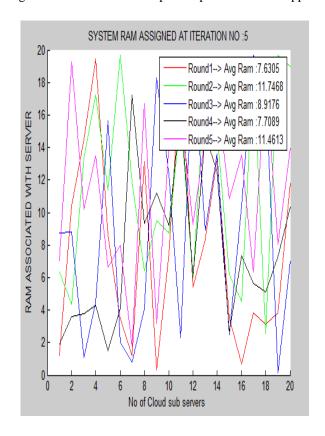


Figure 3: RAM associated with server

Above figure shows the RAM values associated with servers. For round 1 value of Average RAM is 7.603, round 2= 11.7468, round 3= 8.9176, round 4= 7.7089 and round 5 = 11.4613. For large no. of servers the RAM utilization is high.

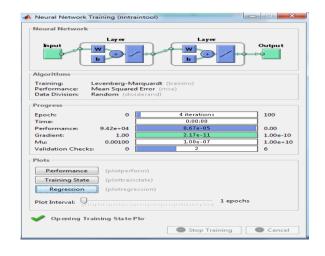


Figure 4: Neural Network Training

The above Figure describes the structure that contains all of the information concerning the training of the network. The Plots like Performance, Training State and Regression are shown through this Figure.

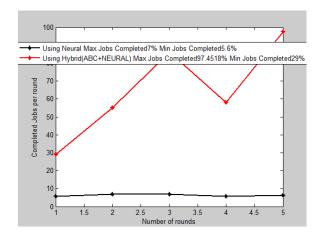


Figure 5: Completed Jobs per Round

Above figure shows the no. of jobs completed per round. By using proposed Hybrid algorithm (Neural + ABC) after 5 rounds, average of maximum completed jobs per round is 97.45% and the average of minimum jobs completed is 29% while by using Neural network alone average of maximum jobs completed is 7% and minimum jobs completed is 5.6%. Number of completed jobs of hybrid approach are more as compare to NN which has less number of completed jobs.

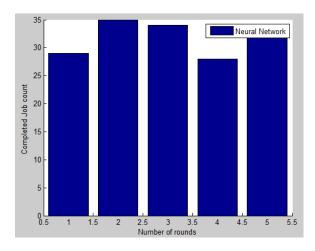


Figure 6: Jobs completed by neural approach

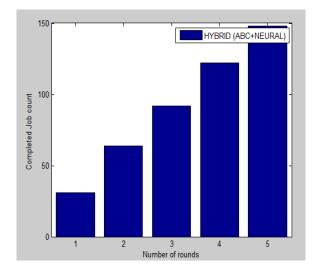


Figure 7: Jobs completed by Hybrid approach

The above Figure 6 and Figure 7 describe the comparison between Hybrid approach (ABC + Neural) and Neural technique used for resource utilization. The proposed Hybrid approach completed 150 jobs per given time while the neural network completed 35 jobs per given time. Hybrid approach gives better results as compare to neural approach.

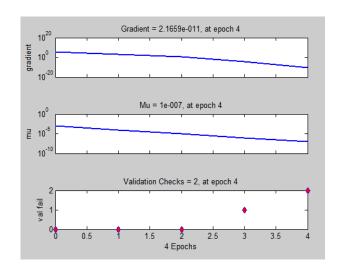


Figure 8: Training of dataset

Above graph shows the different types of parameter which are generated during training of dataset. We check whether we find best gradient value, mutation value and validation checks.

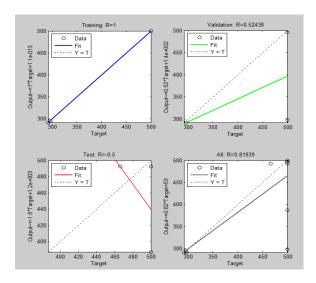
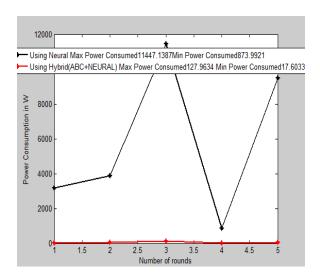


Figure 9: Datasets for training

Above graph shows the description of datasets which are used for the training purpose of dataset. There are total four graph first for training data, second for validation, third for test data which are automatically taken from the training dataset and last for outure of training. In the graph tow lines are present first is solid line and second is doted line which represents the accuacy of training. The three plots represent the training, validation, and testing data. The dashed line in each plot represents the perfect result — outputs = targets. The solid line represents the best fit linear regression line between outputs and targets. The R value is an indication of the relationship between the outputs and targets. If R = 1, this indicates that there is an exact linear relationship between outputs and targets. If R is close to zero, then there is no linear relationship between outputs and targets.



#### Figure 10: Comparison of Power consumption

Figure 10 is showing the difference of power consumption by means of number of jobs. It can be seen from the graph that by using neural network, power consumed is more but by using neural with ABC, the power consumption is very less.

#### 5. Conclusion

With the growing cost of the power in the computer systems, more attention is given to the power management by the researchers as it is directly proportional to the no. of total jobs completed. High the jobs completed, high will be the efficiency of the system. High job execution mainly depends on the scheduling of the jobs according to the proposed algorithm. So, in this work, hybrid algorithm (Neural network with artificial bee colony) for scheduling of jobs has been proposed that will work on weight matrix.

It is concluded that the hybrid approach has less power consumption as compare to the neural network alone by means of number of jobs. The number of jobs completion is more in case of Hybrid algorithm that is for neural and for ABC. Hybrid is completing 150 jobs while neural network alone is completing 35 jobs.

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