

# 1 The Contemporary Affirmation of Taxonomy and Recent 2 Literature on Workflow Scheduling and Management in Cloud 3 Computing

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## 8 **Abstract**

9 The Cloud computing systems preferred over the traditional forms of computing such as grid  
10 computing, utility computing, autonomic computing is attributed for its ease of access to  
11 computing, for its QoS preferences, SLA's conformity, security and performance offered with  
12 minimal supervision. A cloud workflow schedule when designed efficiently achieves optimal  
13 source sage, balance of workloads, deadline specific execution, cost control according to budget  
14 specifications, efficient consumption of energy etc. to meet the performance requirements of  
15 today's vast scientific and business requirements. The businesses requirements under recent  
16 technologies like pervasive computing are motivating the technology of cloud computing for  
17 further advancements. In this paper we discuss some of the important literature published on  
18 cloud workflow scheduling.

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20 **Index terms**— cloud computing, qos, resource scheduling, multi criteria decision, priority scheduling, task  
21 management, ropt, cbt

## 22 **1 Introduction**

23 Cloud computing [1,2] technology offers services of computing, resources etc. that enable users to execute millions  
24 of tasks simultaneously where there is no need for every users to have infrastructure individually. The cloud  
25 architecture consists of client applications and systems as front end connected to the cloud resources comprising  
26 of cloud systems, data and applications as the back end. The front end is connected to the back end by an internet  
27 or intranet network and virtualization technology is used in the deployment of software, networks and data.

28 The Clouds types [3] used in Cloud computing are generally of four types. Public cloud operated by a  
29 service provider and offers service to any general user holding the license to use the service. Private cloud is an  
30 organizations own cloud setup with customized applications and resources limited to their internal users. Hybrid  
31 cloud incorporates public and private cloud functionalities. Community cloud is set up by several organizations  
32 and used commonly by them for their internal requirements. specific task to specific VMs and integration of  
33 tasks and scenarios. The pricing models of cloud service providers are, ? Pay-as-you-go model has price set by  
34 the service provider which is constant. Ex. Companies such as Amazon, Microsoft, Google, Cisco etc. provide  
35 pay-as-you-go cloud datacenters services ? Subscription based pricing has the resource allotted for a period of  
36 time ? Usage based pricing model offers fair pricing for both the client and cloud service provider ? Auction  
37 based pricing is a model based on dynamic resource pricing in federated clouds Other pricing models generally  
38 used are, Cost based pricing, real time pricing, competition based pricing, customer based pricing, etc.

39 Cloud Workflow [6,7] schedules the resources and executes the tasks using algorithms based on the predefined  
40 strategies and objectives. Workflows in a group are usually similar and the differences are related to the variations  
41 in the volume and size of the tasks, data and algorithms used for computing.

### 3 TAXONOMY

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42 The Scheduling approaches usually followed in cloud computing are, Static Scheduling, Dynamic Scheduling,  
43 Cloud Service Scheduling or scheduling at user and system level, Heuristic Scheduling, Workflow Scheduling, etc.  
44 In this paper we discuss about Workflow Scheduling.

45 Workflow scheduling [8,9] handles the process of task and resource scheduling for execution in the cloud  
46 service. A work flow scheduling process assigns the resources and executes the tasks with algorithms. The clouds  
47 scheduling process is identical to the workflow scheduling process in grids [10]. The resource showing better  
48 overall efficiency is selected and based on the algorithmic performance the scheduling procedure is determined.

49 A cloud workflow management system [11] manages the cloud system in provisioning of the resources,  
50 computation development, job scheduling, task execution on the resources and performance evaluation. The ever  
51 increasing scale of data i.e. big data workloads and the requirement to analyze highly complex computations and  
52 agile models has led to the migrating of the scientific work flow management systems of the previous environments  
53 to the Cloud environments. Cloud workflow management systems such as Simple Workflow (SWF) by Amazons,  
54 Blue Works exist by IBM, Nimbus by Tibco and management systems of Microsoft, HP, etc., the research based  
55 scientific workflow management systems [12] such as Swift, Vistrails, Kepler, etc offer services of managing the  
56 cloud environment and process. Cloud workflow management system has several modules. The Virtual Resource  
57 Pool (VRP) manages the resource inquiry. The Virtual Resource Manager (VRM) registers a resource request,  
58 places the request for resources with VRC and sends a resource inquiry to VRP. The Virtual Resource Client  
59 (VRC) processes resource request and manages the resource utilization with the VRP. The Virtual Resource  
60 Provider (VRP) registers and deregisters a Resource and manages the system configuration with CSM. The  
61 Cloud System Management (CSM) component manages the system configuration with the VRM and VRP. The  
62 terminology may vary with application types however the process is the same.

63 In cloud computing different types of scheduling methods are used in different scenarios however a model for  
64 work-flow scheduling that is widely used is the New Berger model. The scheduling strategy of tasks and resources  
65 and execution is based on choosing tasks priority wise and assigning to available processors and computer machines  
66 to meet predefined goals. In dynamic cloud scheduling improving the efficiency of task scheduling with load  
67 balance is the main criteria which however lead to task and load imbalance. The New Berger model is based on a  
68 fair distribution of the tasks to the resources. In this approach first the tasks are allotted to the resources to avoid  
69 uneven resource mapping and overloading. The main criteria or priority is fairness in resources distribution to  
70 the tasks instead of driving the scheduling process to find an efficient solution or limiting the cost. The strategy  
71 of finding an efficient solution in terms of single criteria of cost or time may result in load imbalance affecting  
72 the overall QoS solution path finding process. So the necessity is maintaining a balance between performance  
73 and cost ??13, 14, and 15] which is the basis for New Berger model. The design of the workflow should provide  
74 necessary benefit first and next the user required cost control with fairness can be achieved.

75 The newer techniques of synthetic workflows, algorithms, virtual machine, simulation technologies etc., with  
76 improved scheduling strategies can leverage the balance between fairness and efficiency with benefit. Several  
77 models of workflow scheduling based on the New Berger model have been successful in allocation of resources to  
78 tasks in diverse scenarios without workload imbalance and achieve QoS preferences.

79 The remaining sections are "Systematic literature review" evaluating contemporary cloud computing research  
80 work, followed by the Survey "Conclusion" and finally the "References".

## 81 2 II.

### 82 3 Taxonomy

83 A general taxonomy of the process of cloud workflow comprises of essentially four stages that are, i) Workflow  
84 Scheduling, ii) Work-Flow Execution and iii) Performance Evaluation. Workflow scheduling process is the  
85 automation of tasks and integration of resources into your applications and tools. The process involves scheduling  
86 of the applications and the resources based on Quality of Service factors and also on the Workflow Constraints  
87 such as, ? The QoS factors of scheduling [16] Matlab etc. ? The scalability strategy to be followed especially  
88 in case of dynamic scheduling and real time applications ? Establishing the performance monitoring metrics of  
89 reliability, threshold, throughput, fault tolerance and failover criteria ? Selection of the strategy for performance  
90 evaluation and the metrics for the assessment. The work-flow factors above are dynamic factors constantly affecting  
91 the cloud workflow and scheduling process. A task flow mapped to a resource set undergoes several changes due  
92 to the real time requirement changes of other task flows. So a workflow process should consider several factors in  
93 designing, integrating and execution of the schedule into the Cloud workflow for the success of cloud computing.  
94 The workflow process with sub steps and the techniques and strategies followed in each stage is, ? A resource  
95 pool is formed by the scheduler from a collection of virtual machines manually or automatically ? A virtual  
96 machine that is newly included into the pool is allocated a name and an IP address? A unified pool is configured  
97 comprising of VMs and

98 Applications with parameters such as memory and total number of tasks

99 ? The tasks nature and compatibility with the VMs and Application in terms of OS and scalability is checked  
100 ? The dependencies between the individual VMs and VMs of different pools is established ? A schedule of the  
101 total tasks executable in a VM is created defining the rules for tasks and associated data to be executed ? The  
102 rules defining tasks priority, task priority change criteria, the task execution time, the failover criteria, maximum

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103 retries for a task, etc. are set. The scheduling policy is generally an approach of managing the scheduling  
104 process that does not modify or influence the individual task scheduling set of rules ? Task scheduling mapping  
105 specific tasks according to preset rules and configurations to the required type of resource ? Resources parameters  
106 are mapped to the user task parameters. Resource scheduling allocates virtualized data center resources such  
107 as systems, servers, networks and data in different geographical positions into a resource pool through virtual  
108 technology and with mirror service for allocation on a global scale to the task ? The Schedulers allocate the  
109 cloud resources to the tasks and after the task completion call back the resources.

110 The scientific workflow problems of vast computational complexities may comprise of thousands or more tasks  
111 [17] mapped to million or more possible workflow schedules with variations in resources, time, and cost and  
112 execute during simulation tools for several days in generating an optimal result.

113 Workflow Execution is the algorithmic execution of tasks on each resource as per the SLAs. A simulation of  
114 the scheduling strategy can be done with tools such as Cloud Sim, Matlab etc. A cloud simulator simulates the  
115 scheduling, execution of even huge cloud computing tasks of scientific scale virtually with functionality of scaling  
116 up and down the process for further analysis and optimized result set. The process of work flow execution involves  
117 in stages such as, ? The tasks mapped to resources and the tasks and resources mapped to the task scheduling  
118 algorithms, based on the task priority and execute algorithmically on the resources ? The dynamically scale-up  
119 or scale-down of resources in real time for meeting the variations in the applications user requirements ? A task  
120 if not mapped due to unavailability of resources the scheduler reclassifies the tasks by updating the training set  
121 ? The task is checked for completion before the expectation time

122 ? The tasks if completed the resources are called back and allotted to the tasks next in the priority ? The  
123 algorithms find a best result of mapping the resources to the user tasks The cloud scenarios with varied usage  
124 pattern and QoS constraints require automated provisioning techniques. Scheduling algorithms are required  
125 to implement the workflow scheduling strategies and also for automating the process of scheduling [1]. The  
126 automated provisioning of resources are offered by cloud service providers such as, ? The Amazons EC 2 service  
127 with a simple APIs offers scalability of services with variations of user requirements with complex algorithmic  
128 and statistical calculations ? Open Nebula [19] is a Cloud Infrastructure service provider for automatic resource  
129 provisioning ? The Wrangler system is an automatic provisioning tool for VMs and other resources allocation ?  
130 The Context Broker [20] of project Nimbus is able to allocate huge virtual clusters in simple steps by the users  
131 The algorithmic execution of workflows is done if the tasks to be executed are of huge number. The algorithms  
132 based on heuristics and meta-heuristics offer better optimized search of resource and incur less cost compared  
133 to non-heuristics based algorithms. The Scheduling algorithms [21] are based on the best-effort service or based  
134 on QoS constraints and a candidate instance selected is further improved with enhanced algorithmic approaches.  
135 The algorithms generally used are, Genetic Algorithm (GA), Simulated Annealing (SA), AA or Agent based,  
136 PETRI Network, Partial Critical Path (PCP), Particle Swarm Optimization (PSO), Ant Colony Optimization  
137 (ACO), etc.

138 ? The GA algorithm [22] is based on the chromosome survival tactics. Here individual instances efficiency in  
139 mapping the resources over others is replicated over and over again incorporating the incremental developments  
140 taking place every time. The process is repeated till a single instance is found offering the most optimized QoS  
141 solution according to the set constraints. ? Agent Task Scheduling [23] is based on agent approach in the cloud  
142 systems. The cloud architecture comprises of various clusters of computers made up of layers of nodes. A node in  
143 a cluster carries the information of the resource in a proxy that is distributed to the other clusters in several places  
144 which achieves high performance and extends the network with additional resources ? Simulated Annealing (SA)  
145 [24] algorithmic approach is based on solutions with lesser probability of success in the search process.

146 An initial schedule is prepared on annealing i.e. system temperature and. at the end of iteration the make span  
147 is evaluated and depending on probability a poorer make span is included. This inclusion however decreases for  
148 successive iterations as per the QoS requirements for achieving an optimized solution in problems of global scale  
149 ? Cost Scheduling [25] based on optimizing the cost of cloud computing that virtualizes various cloud resources  
150 such as network or bandwidth, CPU capacity, storage, data transfer times with different price models. The task  
151 scheduling based on cost criteria achieves a cost effective solution The combination of various algorithms also  
152 gives very good results. For example the GA based algorithm when combined with the Ant Algorithm gives  
153 better search results quickly. Another example is task scheduling in cloud computing based on genetic algorithm  
154 with simulated annealing.

155 A performance evaluation helps in designing complex scheduling approaches involving multiple QoS requirements  
156 based on time, cost, and resource. If the target is a high QoS, various scheduling approaches are run in  
157 terms of the QoS values set individually for each of the criteria. If a scheduling approach gives an optimized  
158 result in terms of execution time then the algorithm is further tuned for optimizing the performance in terms of  
159 other criteria i.e. cost and resource for overall QoS value.

160 Work-Flow Performance Evaluation finds solutions for work flow sharing complexities of modeling the tasks,  
161 resources, data etc. study by Schad et al. [26] states, Amazon's EC2 cloud processor has performance variations  
162 of 24%. Performance Monitoring is done by integrating the interface into a monitoring system for observing  
163 factors such as, ? Thresholds monitored in real time enable start-up of new virtual machines and shut down of  
164 unnecessary services ? Throughput or performance of task scheduling is monitored in real time for enhancing  
165 the process ? Fault tolerance is monitored to avoid possible failures in real time ? Reliability of the system can

166 be established in various real time scenarios The performance of a cloud scheduling is evaluated by the execution  
167 of a wide range of scenarios involving diverse infrastructure models, applications, synthetic workflow ensembles  
168 based on diverse user defined and operational constraints. A performance evaluation helps in evaluating and  
169 overcoming the problems of workflow such as, ? the impact of the infrastructure sharing and virtualization ? the  
170 non-virtualized hardware complexities ? The delays of startup, data transfer times, VM boot time variations etc.

171 ? the problems of selecting the initial pool of resources are overcome ? improving the resource model to  
172 handle resources spread across various locations geographically considering the data transfer costs and time ? the  
173 algorithms problems of finding the critical shortest paths between the resources in scheduling tasks and resource  
174 ? the algorithm delay of converge in finding an optimal solution is overcome ? the application performance tested  
175 with other algorithms not evaluated ? The design complexities of scheduling approaches involving multiple QoS  
176 requirements based on time, cost, and resource. ? the scalability requirements of the applications ? the scheduling  
177 problems of the application execution in real-time The challenges of workflow scheduling [27] are in the form of  
178 cloud system, management of tasks, resources, data, integration of above and meeting the expected criteria in  
179 which the managing the resources and scheduling are the most complex tasks above all. These different challenges  
180 are, The research work of cloud computing technology is facing a lot of problems of diverse nature and a lot of  
181 work has to be done in this field for the technology to be fully available with the desired qualities. In the next  
182 section we review some of the important research work done in this field.?

### 183 4 III.

## 184 5 Systematic Literature Review

185 The research work related to the cloud scheduling of workflows, the research criteria and affirmation of its  
186 relevance to the latest scenarios, is reviewed in this section.

187 The criteria of the literature reviewed in this paper are various cloud workflow constraints discussed in the  
188 taxonomy section and are assessed on its relevance to present and future research work. The recent literatures  
189 relevant to the field of cloud computing and specific approaches of cloud work-flow scheduling are reviewed here.

190 The previous research work in the field of grids and clusters that have relevance to the cloud computing  
191 however have faced problems are, A Workflow scheduling algorithm by Lin and Lu [28] for service-oriented  
192 systems is capable of dynamic resource provisioning which is not possible with algorithms of grid environments.  
193 The approach fails in cloud environments implementation as it does not consider the cost of the resources required.

194 A workflow scheduling algorithm by Shi and Dongarra [29] for clusters environments is applicable to the cloud  
195 environments if the cloud infrastructure is allocated unconditionally. The algorithms however are unable to  
196 accurately assess the number of machines required for cost minimization so are inapplicable for scalable cloud  
197 environments dependent on manageable cost as per use.

198 The above methods mostly concentrated on reducing the task runtime based on the availability of limited  
199 number of resources without considering the cost factor involved in the process of execution and are not suitable  
200 for clouds environments.

201 A job scheduling algorithm for cloud environments by Baomin Xu et al. [30] is structured on the Berger model  
202 categorizes the user tasks according to QoS parameters as well as defines the criteria of resource selection for  
203 resource allocation. The approach succeeds in QoS specific task execution however leaves out valid scheduling  
204 criteria such as improving server performance and security.

205 A model for utilizing Cloud environment in addition to desktop Grid resources projected by C. J. Reynolds,  
206 S. Winter, G. Z. Terstyanszky, T. Kiss, P. Greenwell, S. Acs and P. Kacsuk [31] uses the cloud environment for  
207 executing specific slow tasks. The approach does not improve performance in terms of time or cost factors and  
208 without assurance runs on detecting tasks consuming time.

209 A dynamic scheduling strategy for multiple workflows in Clouds by Mao and Humphrey [32] based on  
210 assumption of the existence of various VM models of diverse specifications of cost, proposes to reduce the cost of  
211 task execution. This technique though robust does not assure cost minimization and is only a probable solution.

212 A scheduling approach FAIR for cloud environments by Riktesh Srivastava [33] for multiple users is a node  
213 feedback based approach that identifies nodes ready for computation for allotting the tasks in queue. Though  
214 the response time is enhanced in case of simple tasks and ensures resources allocation in case of big tasks, FAIR  
215 is not based on criteria such as increasing the overall performance, localization of information and quality of  
216 service.

217 A fault tolerant scheduling algorithm FTWS for cloud environments by J. Nirmala, S. Bhanu, S. Jaya divya  
218 in 2010 [34] based on tasks resubmission and replication, is tested with several workflow types with diverse time  
219 constraints and different faults, performs efficiently in comparison to scheduling algorithms not based on fault  
220 tolerance. The proposed algorithm on execution however shows high failure rate. An scheduling algorithm based  
221 on heuristic of PSO by S. Pandey, L. Wu, S. M. Guru and R. Buyya, [35] for cloud environments optimizes  
222 workflow scheduling process to minimize the execution cost and for distributing as well as balancing the tasks  
223 load over the resources. For cost reduction, the execution time is further extended and though this may be viable  
224 in nonelastic environments like cluster or a grid, however in clouds where scalability is a prime criterion the  
225 approach however requires a more detailed schedule.

226 The scheduling problem for cloud environments based on the PSO technique by Z. Wu, Z. Ni, L. Gu, and

227 X. Liu et al [36] under deadline and budget limitations for diverse nature of resources, similar to the work in  
228 [35]. However it supposes the existence of cloud resources i.e. VMs prior to the execution and is not scalable as  
229 required by public clouds.

230 The approaches [29], [36] are designed for executing workflows, allotting resources, implementing the tasks,  
231 managing the scheduling process and for evaluating the execution or the results. The approaches however  
232 are devoid of a regulated QoS based framework and require detailed research activities as well as community  
233 collaboration.

234 An algorithm by M. Rahman, X. Li, and H Palit [37] for hybrid Clouds utilizes minimal resources at very less  
235 cost and with greater performance management.

236 A scheduling algorithm PBTS for clouds by E.-K. Byun Y.-S. Kee, J.-S. Kim, and S. Maeng, [38] assigns tasks  
237 to resources assuming the availability of a particular cloud resource in allocation and scheduling. The approach  
238 however is based on a specific category of VM and does not address the divergent environments of computing  
239 resources.

240 Cloud Workflow algorithms by S. Abrishami, M. Naghibzadeh, and D. Epema [39] optimize workflows for cost  
241 reduction, adhering to the user defined time limits. The two different algorithms proposed however are not based  
242 on the number of times the data is assigned in the process of resources mapping as well asscheduling that result  
243 in automatic escalation of the cost of running the tasks.

244 The scheduling approaches developed by Mao et al [32], S. Pandey et al, [ 35], Wu, et al [36] are specific  
245 scheduling algorithms that are based on the characteristics unique to the cloud computing systems. The research  
246 work of recent times given below by Reynolds et al [31], Rahman et al [37], Byunet al [38] and Abrishamiet al [39],  
247 concentrated on the developing algorithms based on the complex factors associated with cloud environments.

248 The topics of managing multiple workflows executions or multi-tasking workloads on Clouds has been of great  
249 interest for researcher as reviewed below, A scheduling approach for multiple large scale grid workflows application  
250 by R. Duan, R. Prodan, and T. Fahringer [40] is based on a simple game centric optimization approach. It is  
251 effective for enhancing the performance and simultaneously reduces cost.

252 A powerful cloud resource provisioning and task scheduling approach by Smith, Siegel and Maciejewski [41]  
253 focuses on allocation of resources in workflows with specific criteria of QoS in distributed environments.

254 The methods discussed in [34], [40] [ 41] did not consider or study in detail the system efficiency and the  
255 varying criteria of scheduling profiles or policies in minimizing implementation time, cost and balancing load.

256 Workflow models developed specifically for the criteria of energy consumption, tried to minimize the levels of  
257 energy consumption as reviewed below, A cloud scheduling approach for reducing the power consumption by Q  
258 Zhu J Zhu and G Agrawal [42] effectively executes the workflow without affecting the other criteria involved.

259 A "RC2" scheduling algorithm for the cloud systems by Lee and Zomaya [43] is designed for effectively finishing  
260 the given tasks specifically for hybrid cloud. An initial schedule based on the resources of a private cloud or the  
261 organization is calculated for effectively reducing the resource utilization in cloud execution.

262 A cloud systems scheduling model that is energy efficient, by Peng Xiao, Zhi-Gang Hu, Yan-Ping Zhang [44]  
263 schedules workflows with heavy data loads that are executable with virtual data centers. The approach based  
264 on a unique technique, Minimal Data-Accessing Energy Path, proposes to reduce the energy consumption while  
265 accessing huge data.

266 The above approach though is an algorithm for optimizing scheduling time, cost and is energy aware, fails in  
267 efficiently managing the power utilization and preventing scheduling mismatch causing wastage.

268 The scheduling of huge scientific workflow in the areas such as bioinformatics, astronomy, geosciences, physics  
269 etc. executed on grids, clusters or supercomputers till recently are changing to cloud systems for the increasing  
270 performance requirements in terms of big-data workloads and huge complexity of the tasks.

271 A multi-tenant cloud workflow model based on BPEL standard language by Bhaskar Prasad Rimal, Mohamed  
272 A. El Refaey [45], use two workflow approaches based on the semantic and policy criteria for clouds computing of  
273 scientific workflows with different cloud setups. Though effective in performance, the approach did not consider  
274 the varying scheduling profiles or policies for minimizing implementation time and cost and has system efficiency  
275 implementation complexities of separating data and managing security. A framework for scientific workflows  
276 in dynamic resource allocation in the cloud environment by T. T. Huu, G. Koslovskiet al. [46] automatically  
277 allocates, deploys and executes the resources. The approach is based on a model of cost appraisal and optimization  
278 using cloud managed virtualized network and machines. However the algorithms adaptation to the cloud process  
279 is difficult as its performance in terms of time and safety is unsatisfactory and also incurs high costs.

280 The security issues of unknown as well as known kind have been studied in the established grid computing  
281 environment. In the clouds, security for workflow scheduling, finds greater relevance, A workflow schedule with  
282 security factors is studied by Liu, H., A. Abraham, V. Sná?el et al. [47] who proposed a solution of PSO based  
283 scheduling algorithm. The scheduling strategy by Ko?o dziej, J., F. Xhafa [48], shows the probability of failure  
284 in batch processing. The approach quantifies the independent tasks behavior in terms of security and reliability  
285 of resources.

286 The simulation tests and performance0 of the above algorithms in terms of execution time and security is  
287 better compared to other similar approaches, however they do not consider the costs associated and thus are  
288 inapplicable for direct use in the scheduling of cloud workflow systems. With the advances in applications of  
289 cloud computing, security aspect of workflow scheduling has been further studied, A cloud scheduling approach

## 6 CONCLUSION

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290 of multidimensional QoS optimization, by Sun, D. W., G. R. Chang, F.Y.Li et al., [49]schedules the resources  
291 by quantifying the user's application requirements with an immune clone optimization based algorithm approach.  
292 However the approach has difficulties of insufficient depiction of the security functionality, the security attributes  
293 like fuzziness and randomness and the quantitative strength.

294 A scheduling algorithm for cloud systems based on PSO by Li, J et al. [50]for computing vast scale of data,  
295 reduces scheduling time as well as cost of the cloud service. The approach however does not address the unfamiliar  
296 security threats associated with cloud computing platform. A hierarchical scheduling approach for systems of  
297 cloud computing based on scheduling the service layer and the task layer by Wu, Z., et al [51] is wholly customer  
298 centric. In the cloud environment, the appropriate resources for the workflow tasks are allotted with service layer  
299 scheduling while the execution time and cost are reduced by task scheduling.

300 The algorithms discussed above in a cloud workflow scheduling system, while meeting the security constraints  
301 of execution, insufficiently address the QoS requirements of execution time and cost.

302 IV.

## 303 6 Conclusion

304 In this paper we have reviewed the taxonomy and contemporary affirmation of the benchmarking models in  
305 recent literature about works flow scheduling in cloud computing. Many of scheduling models observed in the  
306 literature. The common constraint of these many models is the generalization of the scheduling priorities or deep  
307 involvement of the experts towards customizing the scheduling order. The experiences learnt from the existing  
308 models evincing that the scheduling priorities are divergent from one context to other of the organizations, also  
309 different depend the resource availability and usage. Hence it is quite obvious to confirm that there is vast  
310 research scope to define novel Scheduling strategies for Cloud workflow management. In order to overcome the  
311 constraints observed, the custom level scheduling will be the possible criteria of the research. The lessons from  
312 the past scheduling order can help to redefine the current scheduling order, henceforth the evolutionary and  
machine learning strategies are highly adoptable to define robust and scalable workflow scheduling strategies <sup>1 2</sup>



Figure 1:

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## 6 CONCLUSION

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