Virtual Machine Migration using Particle Swarm Optimization Algorithm in Mobile Cloud Computing

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Abstract
Mobile Cloud Computing (MCC) can play a vital role by offloading Big Data related tasks such as sharing, processing and analysis, from mobile applications to cloud resources, ensuring Quality of Service (QoS) demands of end-users. Such resource migration, is also termed Virtual Machine (VM) migration. The emergence of Mobile Cloud Computing (MCC) facilitates reduction of task-execution time and real-time communication latency.

Introduction
The emergence of mobile cloud computing (MCC) facilitates reduction of task-execution time and real-time communication latency. The [1], proposes how cloud computing is integrated into the mobile environment (MCC). MCC effectively utilizes distributed cloud server resources such as CPU, memory, network, and ports to execute the mobile Smart healthcare applications. Using MCC, mobile devices (MD) can offload these applications to a resourceful cloud server for faster execution. However, the long distance between cloud servers and the MDs may increase the response time for interactive applications, increasing the total execution time.

The [5] proposes that a dynamic server migration and consolidation algorithm is introduced. The [6] proposes identification of various computing paradigms promising to deliver the vision of computing utilities. The [7] proposes a CloneCloud i.e. a system that automatically transforms mobile applications to benefit from the cloud. The [8] proposes a cloud-based solution that is based on the client running on the smartphone and the server side running in the cloud.

User mobility is not the only reason forcing a VM to migrate. Migration can be initiated to minimize the overprovisioned resources and thus improve the overall system objectives. For instance, if a VM is required to be migrated from a cloudlet to any of the candidate cloudlets, the new cloudlet may not have the same type of VM. In that case, a VM with more resource than the current one must be chosen and provisioned in order to migrate the VM and thus minimize task-execution time. However, in this approach, the VM migration is provisioned more resources than the required. Therefore, this over-provisioned resources greatly decreases the system objectives, as it reduces the number of provisioned VMs in the cloudlets.

Related Work
The [1], proposes how cloud computing is integrated into the mobile environment (MCC). They showed the performance improvement with respect to battery life, storage, and bandwidth. Also, they have considered environmental factors like heterogeneity, scalability, and availability issues.

They also focussed on security issues related to reliability and privacy.

The Advantage of this paper is that it improved the performance metrics of cloud. But it doesn’t address security of mobile users and data in cloud.

The [5] proposes that a dynamic server migration and consolidation algorithm is introduced. The algorithm is shown to provide substantial improvement over static server consolidation in reducing the amount of required capacity and the rate of service level agreement violations.

The advantages proposed in this paper is that it consolidates the server capacity for better resource usage but the limitation is that it is difficult to virtualize large amounts of data servers.

The [6] proposes identification of various computing paradigms promising to deliver the vision of computing utilities; defines Cloud computing and provides the architecture for creating market-oriented Clouds by leveraging technologies such as VMs.

The advantages proposed in [6] is better load balancing strategies to improve performance. But some restrictions on available services may be faced, as it depends upon the cloud service provider.

The [7] proposes a CloneCloud i.e. a system that automatically transforms mobile applications to benefit from the cloud. The system is a flexible application partitioner and execution runtime that enables unmodified mobile applications running in an application-level virtual machine to seamlessly off-load part of their execution from mobile devices onto device clones operating in a computational cloud.

The advantage proposed by [7] is Off-loading a part of an application to a CloneCloud, reduces the task...
execution time. But more elasticity means more less control in case of public clouds. The [8] proposes a cloud - based solution that is based on the client running on the smartphone and the server side running in the cloud. Sensors embedded in the smartphone are used for the measurement of the different information about the monitored person. Basic algorithms evaluating current person’s health status run on the smartphone. Measured data are sent to the second part of the application running in the cloud for deeper analysis. The advantage proposed in [8] is cost - effectiveness from the consumer side. But it vastly demands for cloud security. The [9] states that ant colony optimisation is applied to the problems where a number of alternative pheromone representations are available, each of which interacts with this underlying bias in different ways. The [9] explores both the structural aspects of the problem that introduce this underlying bias and the ways two pheromone representations may either lead towards poorer or better solutions over time. Mobile computing continuously evolves through the sustained effort of many researchers. It seamlessly augments users’ cognitive abilities via compute-intensive capabilities such as speech recognition, natural language processing, etc. By thus empowering mobile users, we could transform many areas of human activity. The [10] discusses the technical obstacles to these transformations and proposes a new architecture for overcoming them. In this architecture, a mobile user exploits virtual machine (VM) technology to rapidly instantiate customized service software on a nearby cloudlet and then uses that service over a wireless LAN; the mobile device typically functions as a thin client with respect to the service. A cloudlet is a trusted, resource-rich computer or cluster of computers that’s well-connected to the Internet and available for use by nearby mobile devices. Our strategy of leveraging transiently customized proximate infrastructure as a mobile device moves with its user through the physical world is called cloudlet-based, resource-rich, mobile computing.

Existing system

In the existing system using ant colony algorithm its Mobile users may move from one Access Point (AP) to another, increasing their distances between current locations and the cloudlet, where the tasks are provisioned. So, it increases the task-execution time. Pheromones and Initial Pheromone values are calculated In the ACO algorithm, pheromones represent the desirability of choosing a solution. The pheromones represent the desirability of assigning a VM to a cloudlet. Each ant starts with an initial pheromone value for each VM to cloudlet pair. The initial solution is generated using a greedy First Fit(FF) VM migration approach.

A Heuristic Value is calculated to build a solution each ant uses the local heuristic value to select a cloudlet for a VM. This heuristic value defines the favourability of choosing a cloudlet for a VM to construct the solution. As PRIMIO tries to jointly optimize the objectives of total task-execution time and resource over-provisioning in cloudlets. So, the local heuristic value has to be defined in such a way that the system can optimize the task execution time and resource over-provisioning to reflect the system objectives. Local heuristic is defined as, \( \eta_u,v = \alpha \times \eta_E u,v + (1 - \alpha) \eta_R u,v \) (19) where, \( \alpha \) is the system parameter defines the relative weight between the objectives of total task execution time and resource over-provisioning of cloudlets. It can be tuned according to the system environment. \( \eta_E i, j \) and \( \eta_R i, j \) is the objectives of total task execution time and the resource over-provisioning respectively, if VM migrates from cloudlet i to cloudlet j. Local Pheromone Update: When an ant chooses a VM pair to construct a solution, it immediately updates the local pheromone value in relation to the initial pheromone value. Local pheromones are updated by each ant using the following relation, \( \tau_{u,v}(t + 1) = (1 - \gamma_l) \times \tau_{u,v}(t) + \gamma_l \tau_0 \) (24) where, \( \gamma_l \) is the system parameters, denoting the relative importance of current pheromone value at time t, \( \tau_{u,v}(t) \). Global Pheromone Update: Global pheromone values are updated for each pair of VM and cloudlet only when all ants constructed their local optimal solutions and updated the global optimal solution. The global pheromone values are updated using the following relation, \( \tau_{u,v}(t + 1) = (1 - \gamma_g) \times \tau_{u,v}(t) + \gamma_g \Delta \tau_{u,v} \) (25) where the \( \gamma_g \) is the global pheromone system parameter, this parameters can be tuned according to the system objectives. \( \Delta \tau_{u,v} \) is the global pheromone value for the updated global solution, which is defined as \( \Delta \tau_{u,v} = \tau_{u,v}, \) if \( (u,v) \in \text{PGS} \), 0 otherwise (26) where, \( \text{PGS} \) is the global solution set of the selected VM u for migration and the target VM v where the migration will be occurred.

Proposed system

We propose a VM migration technique for a heterogeneous MCC system following the user’s mobility pattern. That is, when a user moves from one cloudlet to another cloudlet, the resource or VM must be migrated to the cloudlet that is nearest to the user. We use particle swarm Optimization (PSO) to identify the optimal target cloudlet. We develop a particle swarm Optimization (PSO) based VM migration model, in which VM are migrated to
candidate cloud servers so as to maximize the total utility of the MCC system.
The Cornfield Vector: Heppner’s bird simulations had a feature which introduced a dynamic force into the simulation. His birds flocked around a “roost,” a position on the pixel screen that attracted them until they finally landed there. This eliminated the need for a variable like craziness, as the simulation took on a life of its own. While the idea of a roost was intriguing, it led to another question which seemed even more stimulating. Heppner’s birds knew where their roost was, but in real life birds land on any tree or telephone wire that meets their immediate needs. Even more importantly, bird flocks land where there is food. How do they find food? Anyone who has ever put out a bird feeder knows that within hours a great number of birds will likely find it, even though they had no previous knowledge of its location, appearance, etc. It seems possible that something about the flock dynamic enables members of the flock to capitalize on one another’s knowledge, as in Wilson’s quote above. The second variation of the simulation defined a “comfield vector,” a two-dimensional vector of XY coordinates on the pixel plane. Each agent was programmed to evaluate its present position in terms of the equation: 

\[ \text{Eval} = Jw + \text{so that at the (100,100) position the value was zero. Each agent “remembered” the best value and the XY position which had resulted in that value. The value was called pbest} \] and the positions pbestx\] and pbesty\] (brackets indicate that these are arrays, with number of elements = number of agents). As each agent moved through the pixel space evaluating positions, its X and Y velocities were adjusted in a simple manner. If it was to the right of its pbestx, then its X velocity (call it vx) was adjusted negatively by a random amount weighted by a parameter of the system: vx\]=vx\] - rand(\)^p\-increment. If it was to the left of pbestx, rand(\)^p\-increment was added to vx\]. Similarly, Y velocities vy\] were adjusted up and down, depending on whether the agent was above or below pbesty. Secondly, each agent “knew” the globally best position that one member of the flock had found, and its value. This was accomplished by simply assigning the array index of the agent with the best value to a variable called gbest, so that pbestx[gbest] was the group’s best X position, and pbesty[gbest] its best Y position, and this information was available to all flock members. Again, each member’s vx\] and vy\] were adjusted as follows, where g-increment is a system parameter. iifpresentx\] > pbestx[gbest] then vx\] = vx\] - rand\() *g\-increment if presentx\] < pbestx[gbest] then vx\] = vx\] + rand\() *g\-increment ifpresenty\] > pbesty[gbest] then vy\] = vy\] - rand\() *g\-increment ifpresenty\] < pbesty[gbest] then vy\] = vy\] + rand\() *g\-increment In the simulation, a circle

marked the (100,100) position on the pixel field, and agents were represented as colored points. Thus an observer could watch the flocking agents circle around until they found the simulated cornfield. The results were surprising. With p-increment and g-increment set relatively high, the flock seemed to be sucked violently into the cornfield. In a very few iterations the entire flock, usually 15 to 30 individuals, was seen to be clustered within the tiny circle surrounding the goal. With p-increment and g-increment set low, the flock swirled around the goal, realistically approaching it, swinging out rhythmically with subgroups synchronized, and finally “landing” on the target.

Module description

1) Identify VMs for migration:

As user might move, the distance with respect to cloud differs. This increases latency and cause performance issues. Hence, we need to identify the best VMs that suits the user transaction. We will use Particle Swarm Optimization to identify best VMs.
Fitness value of VM will be computed with PSO algorithm. A fitness value is an objective function that summarizes a single figure of merit to achieve optimal design. The fitness value is used to compute the local best position and the global best position. Here a node, in this case, a virtual machine is a part of the belief space we consider. The virtual machine’s state depends upon the fitness value.

3) **Cloudlet migration for higher performance**

Once the best VMs are identified, we will validate the cloudlet and migrate transaction to the respective cloudlet using Live Virtual Migration technique. This reduces downtime for migrating overloaded VMs.
Fig. 7. VM Migration - Architecture

**Description of the system architecture**

In the proposed architecture, a computation migration system is depicted, which has two important functionalities. First, the computation migration between resource-constrained mobile devices and cloudlets; Second, virtual machine migration between two cloudlets to minimize the task execution time as well as resource over-provisioning. The application offloading manager helps a mobile device to decide which part of the mobile application should be offloaded. The client-request handler in a cloudlet manages incoming task execution requests from mobile devices and dispatches to the virtual machine scheduler. The VM scheduler uses the information from hypervisor to initiate the VM scheduling. Moreover, after a specific time epoch, the VM migration manager interacts with the VM scheduler to select tasks and to migrate so as to reduce the task-execution time. In this work, our main concern is to develop an algorithm for the VM migration manager for selecting a set of tasks and cloudlets, remapping tasks to minimize execution time. In the VM migration manager, there are three main components- VM Selection, Joint VM Migration Engine, and VM Migration Initiator, which contribute to making optimal migration decisions. Moreover, the user mobility manager takes local mobility management information in conjunction with global mobility prediction manager to decide on a set of probable cloudlets for a user. Finally, the VM migration manager initiates migration process for a set of tasks requiring faster execution.

**PSO Algorithm**

For each particle Initialize particle; END

Do For each particle

Calculate fitness value;

If the fitness value is better than the best fitness value (pBest) in history set current value as the new pBest; 

End

Choose the particle with the best fitness value of all the particles as the gBest; 

For each particle 

(a) Calculate particle velocity according equation 

(b) Update particle position according equation 

End

While maximum iterations or minimum error criteria is not attained

**Conclusion**

We propose a VM migration technique for a heterogeneous MCC system following the user’s mobility pattern. That is, when a user moves from one cloudlet to another cloudlet, the resource or VM must be migrated to the cloudlet that is nearest to the user.

We use Particle Swarm Optimization (PSO) to identify the optimal target cloudlet. We develop an Particle Swarm Optimization (PSO) based VM migration model, in which VM are migrated to candidate cloud servers so as to maximize the total utility of the MCC system.

**References**


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9. Acknowledgements

This work was supported in part by a grant from the National Science Foundation.

Contribution of others who might have given suggestions or review comments.

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