DGridSim: A REAL-TIME DATA GRID SIMULATOR WITH HIERARCHICAL JOB AND DATA SCHEDULING

Safai Tandoğan¹, Atakan Doğan² and Celal Murat Kandemir³

¹ C Tech, TUBITAK MAM TEKSEB, Kocaeli, Turkey
² Department of Electrical and Electronics Engineering, Anadolu University, Eskisehir, Turkey
³ Department of Electrical and Electronics Engineering, Eskisehir Osmangazi University, Eskisehir, Turkey

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Abstract: In this study, DGridSim, a process oriented and discrete-event driven all-in-one Data Grid simulator, is introduced, and some initial simulation results are reported to validate its operation. DGridSim has some distinguishing features which make it unique among the Data Grid simulators available in the literature. First of all, in order to provide the guaranteed service to real-time jobs running on the system, DGridSim incorporates the advance reservation of all system resources, including computing, network, and storage. Second, DGridSim supports a hierarchical real-time scheduling architecture for both jobs and jobs’ data. Third, it allows the simulation of all related components of a Data Grid system which may have an impact on the system’s real-time performance.

1 INTRODUCTION

In this study, a real-time Data Grid simulator, namely DGridSim is presented. The main objective in developing DGridSim is to provide an all-in-one platform to study real-time job distribution, data replication, and data dissemination algorithms for Data Grids.

In the literature, a variety of simulators has been proposed: Optorsim (Camaron, Millar, Nicholson, Schiaffino, Zini and Stockinger, 2004), GridSim (Buyya and Murshed, 2002, Sulistio, Cibej, Robic and Buyya, 2008), SimGrid (Casanova, 2001). A comparison study of a variety of simulators for the Grid systems, including GridSim, Optorsim, and SimGrid, was presented by Quetier and Cappello, 2005.

None of aforementioned simulators focuses on the evaluation of the real-time performance of the Grid systems. Furthermore, among these simulators, DGridSim has some remarkable features, some of which are as follows: (1) Hierarchical real-time job scheduling algorithms can be simulated, where they can run either online or offline mode. (2) Pull or push based data replication algorithms can be simulated. (3) A data dissemination algorithm can be evaluated together with a data replication algorithm and job scheduling algorithm, or vice versa. (4) The network traffic model is based on the flows due to the file transfers. All network resources are treated as the first class entities similar to the computing and storage resources. As a result, the network connectivity is transformed into a scheduled service. (5) Its design is modular, extensible and layered to provide for the maximum flexibility in simulating different Data Grid system scenarios.

2 DGridSim

DGridSim is written in C++ programming language using C++SIM20 discrete-event simulator library by Mesquite Software. The C++SIM20 library allows the development of process-oriented discrete-event simulation programs. A program using C++SIM20 library models a system as a collection of C++SIM20 processes which interact with each other by using the C++SIM20 structures. The C++SIM20 structures that are used in DGridSim are the following: process, facility, and event.

DGridSim is a simulator that has been designed to achieve modularity, extensibility, and layered architecture. Specifically, the layered architecture means that DGridSim is built in layers with respect
to the layered software architecture of real Grid systems where the higher layers make use of the functionalities provided by the lower layers. DGridSim is composed of four layers: Basic Grid Fabric, Communication, Main Grid Services, and Applications. These layers as well as how advance reservation mechanism is implemented for computing, networking, and storage resources were explained by Atanak, Tandogan and Doğan, 2010. Note that without the advance reservation, it is impossible to guarantee any form of quality of service.

2.1 System Architecture

The system architecture of DGridSim is shown in Figure 1. DGridSim is a layered software with a well-defined object hierarchy. The hierarchy between the objects is established using interfaces, Abstract Base Classes in C++. This provides a flexible and expandable environment for researchers.

![Figure 1: DGridSim system architecture.](image)

At the bottom level, there exists CSIM. A Base Layer containing low level simulation utilities, such as registering events, advancing simulation time, etc., has been developed using this CSIM development kit. A Grid Component Library (GCL) Layer defining interfaces and contracts for grid services and grid models has been designed on this Base Layer. The User Code Layer consists of the standard and specialised implementations of the interfaces defined in GCL Layer. A Graphical User Interface Layer is planned for the future. The User Code Layer is the layer for researchers to inject their own implementations of different grid algorithms. If predefined service contracts or grid models are not adequate for the researcher, new interfaces should be defined in GCL Layer. It is expected that no intervention should be necessary to the Base Layer.

2.2 Service Architecture

Data Grid components can be divided into two main groups, Resource Components and Service Components. Resource Components do have definite properties. For example, a Storage Element can have properties such as storage space, read/write bandwidth, and access latency, but do not have any business logic (a Storage Element does not contain any procedures). Since Resource Components are passive components, Resource Managers are required to perform necessary operations on them. Following the CSIM terminology, Resource Components are designed using facilities, and Service Components are designed using processes.

Service Components can be based on three different interfaces in DGridSim. Asynchronous Service Base is designed to operate on a queue. This queue can be a collection of grid jobs or data transfer requests. Queue handling mechanism can be triggered periodically or by a preconfigured event. Scheduling Services are developed by extending Asynchronous Service Base.

Synchronous Service Base provides its service by simple function calls. Each function call can have a cost which can be determined at run-time by extending this base interface. Information Services are usually developed on top of Synchronous Service Base.

The third base is the combination of Asynchronous and Synchronous Service Base and is called as Hybrid Base. Hybrid Base controls a queue and at the same time offers simple function calls. Reservation Services are based on Hybrid Base because they need to process the reservation requests which are placed in a queue and also they need to provide functions for committing or cancelling a reservation request.

3 JOB SCHEDULING

DGridSim supports the simulation of hierarchical Data Grid systems where a global scheduler (Grid Scheduler) for the whole system and a local scheduler (Site Scheduler) for each site are deployed to manage the available computing resources.

3.1 Grid Scheduler

In DGridSim, all Grid schedulers are implemented
as a part of Grid Scheduling Service (GSS). In order to support the job scheduling activities, Grid Scheduling Service interacts with some other global services as follows:

1. Grid Job Submission Service (GJSS): GJSS accepts the jobs submitted by the applications, and invokes the scheduler.
2. Grid Job Dispatch Manager (GJDM): Once a site has been determined for a job, GSS informs GJDM for the submission of the job to this site.
3. Grid Information Service (GIS): GIS provides GSS (and Data Management Service) with all sorts of static and dynamic information related to the system.
4. Replica Location Service (RLS): RLS is queried to find the physical location of all data items available in the system.

3.1.1 Random

The Random algorithm is implemented as an online algorithm in GSS as follows:

1. Once a new job is submitted to the system, GJSS places it to a queue and invokes GSS.
2. GSS handles each job in the queue one by one. For the current job, GSS randomly selects a site and informs GJDM.
3. GJDM forwards the job to the selected site.

3.1.2 Earliest Deadline First

The Earliest Deadline First algorithm is realized as an offline algorithm in GSS and it runs as follows:

1. Once a new job is submitted to the system, GJSS places it to a queue. Then, different from the online case, it invokes GSS periodically with a predefined period.
2. Upon invoking, GSS handles all jobs in the queue at the same time. GSS sorts the jobs in the increasing value of deadline, randomly selects a site in the sorted order for each job; informs GJDM.
3. GJDM forwards the job to the selected site.

3.1.3 Minimum Completion Time First

The Minimum Completion Time First (MCTF) algorithm is realized in GSS and it runs as follows:

1. When a new job is submitted to the system, GJSS first places it to a queue. Then, it invokes GSS immediately since MCTF is an online algorithm.
2. GSS handles each job in the queue one by one. For the current job, GSS sends an inquiry to GIS to fetch $p_i$ (total computing power in MIPS) and $l_i$ (average instantaneous computing load in seconds) for all sites.
3. GSS computes the expected job finish times $c_i$ for each site, where $c_i = l_i + t_j/p_i$ and $t_j$ is the number of instructions of the current job.
4. GSS selects the site with minimum job completion time and informs GJDM.
5. GJDM forwards the job to the selected site.

3.2 Site Scheduler

DGridSim realizes all site schedulers as a part of Site Scheduling Service (SSS). In order for Site Scheduling Service to carry out the job scheduling activities, it cooperates with other local services as follows:

1. Site Job Submission Service (SJSS): SJSS accepts the jobs submitted by GJSS, and invokes the site scheduler.
2. Site Job Invoke Manager (SJIM): Once a scheduling decision has been made for a job, SSS informs SJIM about this job as well as the related information, such as the computing element chosen, job start time and duration of execution. SJIM simulates the execution of the job by creating a CSIM process which uses the CSIM facility for the chosen computing element for the duration of execution starting from the start time.
3. Local Reservation Service (LRS): LRS holds all the advance reservations made for the computing/storage/networking resources in the site.
4. Local Data Manager (LDM): LDM is the facade of Local Data Management Service, which is explained in Section 4. Local Data Management Service has the responsibility of making data available for the jobs running in the site. LDM can provide the requested data from the local storage element, or from a remote one with the help of global Data Manager.

3.2.1 Real-time Min-max

DGridSim currently supports an online site scheduling algorithm, namely real-time Min-max. The operation of SSS with real-time Min-max is as follows:

1. Once a new job is received by SJSS, SJSS places it to a queue and invokes SSS.
2. SSS handles each job in the queue one by one. For the current job, SSS sends an inquiry to LRS asking for a free time-slot on each computing element, where the slot start time is the current time and the slot finish time is the task deadline.

3. Local Reservation Service returns the time slot with the latest start time for each computing machine. Thus, this is the max operation. Among the time slots returned by LRS, SSS selects the computing element with the earliest start time, which is the min operation.

4. Once a deadline satisfying time-slot is found, SSS submits a data transfer request for the job’s data to Local Data Manager, where the data request deadline is set to the start time of the time-slot.

5. If LDMS guarantees that the requested data will be copied into the site before the data request deadline, SSS considers the job to be satisfied, and invokes SJSS. Otherwise, SSS drops the job from the queue.

6. Site Job Invoke Manager enables running the job in the time-slot determined by the scheduler on the chosen computing element.

4 DATA SCHEDULING

DGridSim realizes data scheduling in hierarchical fashion similar to the job scheduling. That is, there is a global Data Management Service (DMS) for the whole system, and a Local Data Management Service (LDMS) for each site.

4.1 Data Management Service

In the DGridSim, Data Management Service (DMS) coordinates the transfer of data from one site to another. In doing so, DMS cooperates with some other services.

1. Data Manager (DM): DM is the service point of DMS for all sites in the Grid. That is, all Local Data Management Services use DM to submit a site-to-site data transfer request to DMS.

2. Reservation Service (RS): When DMS finds a feasible path for a data transfer request by means of a data dissemination algorithm, a specific bandwidth value must be reserved on all links of the path. Thus, DMS sends a reservation request to RS so that RS can make these bandwidth reservations on the respective links.

3. File Transfer Service (FTS): After DMS is informed by RS that all reservations are successful, DMS starts FTS in order to copy a data item from one site to another.

4.1.1 Minimum Delay Feasible Path First

Data Management Service runs a data dissemination algorithm, namely Minimum Delay Feasible Path First (MD/FPF), to find a feasible path for each data request. A path is considered to be feasible if and only if it has the sufficient bandwidth to deliver the data to its destination at its deadline. DMS realizes MD/FPF as follows.

1. Once a site-to-site data transfer request is received by DM, DM puts it to a queue and invokes DMS. Thus, DMS runs in online mode.

2. DMS handles each request in the queue one by one. For the current data request, DMS sends an inquiry to Replica Location Service to find out all source sites where the requested data item is located.

3. DMS fetches the information about available link bandwidths and network topology from GIS.

4. DMS computes the required minimum bandwidth value: \(bw = \frac{\text{file size}}{\text{request deadline} - \text{current time}}\). Based on the undirected graph model of the system due to the network topology, DMS deletes all links with an available bandwidth value less than \(bw\) from the graph.

5. In the modified graph, DMS runs the Dijkstra’s shortest path algorithm to find out a minimum delay feasible path.

6. After finding a path, DMS submits a reservation request to Reservation Service which includes all the links on the path, reservation start (current time) and finish (request deadline) times, and bandwidth value \(bw\).

7. If all reservations are succeeded, DMS calls for File Transfer Service to start the site-to-site data transfer over the reserved path. Otherwise, this request is deemed to be unsatisfiable. In either case, Data Manager informs LDMS accordingly.

4.2 Local Data Management Service

Local Data Management Service (LDMS) is decoupled from Site Scheduling Service and it coordinates the site’s data services as explained
below.

1. While a job is being scheduled by SSS, SSS places a data request to Local Data Manager, which is the LDMS’s service point to SSS. LDM puts this request in a queue and invokes LDMS.

2. For the current data request, LDMS contacts with Local Replica Location Service and learns if the requested data item(s) can be locally provided.

3. For the data items which are already available in the site, LDMS submits a reservation request to Local Reservation Service. This request includes the all the links from a local storage element to the chosen computing element, reservation start (current time) and finish (request deadline) times, and bandwidth value (bw).

4. For the unfound data items, LDMS sends a data request to Data Manager so that these items can be copied from some remote site(s) into this site.

5. If all local reservations are successfully made, and/or LDMS is informed by Data Manager that the requested data item(s) will be available at the task start time, Local Data Manager notifies SSS with either positive/negative acknowledgement accordingly.

5 EXPERIMENTS

Using DGridSim, the three Grid scheduling algorithms were evaluated. With the start of the simulation, a Data Grid system was created. The system was assumed to have the following properties. It has ten sites each of which includes thirty-two heterogeneous computing elements and a single storage element. The computing elements have MIPS rating of \( U \sim [7500, 12500] \), where \( U \sim \) means uniformly distributed, and the storage elements have storage capacity of \( U \sim [175000, 225000] \) Mbytes. Furthermore, every site is equipped with a gateway router to which all computing and storage elements are connected. The links between computing and storage elements and their gateway have bandwidth of \( U \sim [750, 1250] \) Mbytes/sec and \( U \sim [1750, 2250] \) Mbytes/sec, respectively. Ten gateway routers are interconnected by a randomly generated network topology composed of ten routers and twenty-five links whose bandwidths with an average bandwidth of \( U \sim [750, 1250] \) Mbit/sec.

After the creation of a Data Grid system, jobs were produced. The jobs were characterized with job size, deadline, and the number of data items. In the base set of simulation studies, job sizes are \( U \sim [3750000, 7500000] \) MI (Million Instruction), deadlines are \( U \sim [750, 1250] \) seconds, and the number of data items needed in order for jobs to start their execution is just one. During the simulations, jobs were submitted to the system with a rate of one job per five seconds.

Initially, all two-hundred different data items were assumed to be stored in a single (Tier-0) site without any computing elements. Thus, all data are distributed from this Tier-0 site to all other sites with computing capability. Moreover, jobs are randomly associated with the data items whose sizes are \( U \sim [750, 1250] \) Mbytes.

Using DGridSim, a base set of results was first established for the following parameter values: Number of Jobs=1000, Mean Job Size=5000000 MI, Mean Job Deadline=100 sec, Mean Number of Data Items=1 and Mean Link Bandwidth=125 Mbytes/sec. Later, these parameters are varied and the effects are observed and reported in Tables 1-3. Each data in Tables 1-3 denotes the average satisfiability (the ratio of number of jobs finished before their deadlines to total number of jobs) in three simulation runs.

Table 1 shows the effect of changing the number of jobs submitted to the system from 1000 to 3000 on the algorithms. According to Table 1, all three algorithms maintain a relatively stable performance on average. Furthermore, Random and EDF have shown very similar performance, and they are slightly better than MCTF.

Table 1: Performance of the three grid scheduling algorithms when the number of jobs is increased.

<table>
<thead>
<tr>
<th>Number of Jobs</th>
<th>1000</th>
<th>1500</th>
<th>2000</th>
<th>2500</th>
<th>3000</th>
</tr>
</thead>
<tbody>
<tr>
<td>Random</td>
<td>0.94</td>
<td>0.96</td>
<td>0.96</td>
<td>0.97</td>
<td>0.98</td>
</tr>
<tr>
<td>EDF</td>
<td>0.94</td>
<td>0.96</td>
<td>0.96</td>
<td>0.97</td>
<td>0.98</td>
</tr>
<tr>
<td>MCTF</td>
<td>0.92</td>
<td>0.90</td>
<td>0.89</td>
<td>0.92</td>
<td>0.85</td>
</tr>
</tbody>
</table>

Table 2 shows the impact of increasing the mean job size from 5000000 (5 M) MI to 10000000 (10 M) MI on the algorithms. According to Table 2, increasing job sizes significantly degrades the performance of MCTF. On the other hand, it seems that both Random and EDF keep its performance at a top level.

Table 2: Performance of the three grid scheduling algorithms when the mean job size is increased.

<table>
<thead>
<tr>
<th>Mean Job Size</th>
<th>5000000</th>
<th>10000000</th>
</tr>
</thead>
<tbody>
<tr>
<td>Random</td>
<td>0.94</td>
<td>0.92</td>
</tr>
<tr>
<td>EDF</td>
<td>0.94</td>
<td>0.92</td>
</tr>
<tr>
<td>MCTF</td>
<td>0.92</td>
<td>0.85</td>
</tr>
</tbody>
</table>
Table 2: Performance of the three grid scheduling algorithms when the mean job size is increased.

<table>
<thead>
<tr>
<th>Mean Job Size (MI)</th>
<th>5 M</th>
<th>6 M</th>
<th>7 M</th>
<th>8 M</th>
<th>10 M</th>
</tr>
</thead>
<tbody>
<tr>
<td>Random</td>
<td>0.94</td>
<td>0.95</td>
<td>0.97</td>
<td>0.96</td>
<td>0.97</td>
</tr>
<tr>
<td>EDF</td>
<td>0.94</td>
<td>0.95</td>
<td>0.97</td>
<td>0.96</td>
<td>0.97</td>
</tr>
<tr>
<td>MCTF</td>
<td>0.92</td>
<td>0.90</td>
<td>0.67</td>
<td>0.50</td>
<td>0.33</td>
</tr>
</tbody>
</table>

Table 3 shows the impact of increasing the mean job deadline from 600 sec to 1000 sec on the algorithms. According to Table 3, increasing deadline clearly helps MCTF to improve its performance. On the other hand, neither Random nor EDF has experienced a significant variation in their performance.

Table 3: Performance of the three grid scheduling algorithms when the mean deadline is increased.

<table>
<thead>
<tr>
<th>Mean Job Deadline (sec)</th>
<th>600</th>
<th>700</th>
<th>800</th>
<th>900</th>
<th>1000</th>
</tr>
</thead>
<tbody>
<tr>
<td>Random</td>
<td>0.96</td>
<td>0.97</td>
<td>0.98</td>
<td>0.98</td>
<td>0.97</td>
</tr>
<tr>
<td>EDF</td>
<td>0.96</td>
<td>0.97</td>
<td>0.97</td>
<td>0.98</td>
<td>0.97</td>
</tr>
<tr>
<td>MCTF</td>
<td>0.85</td>
<td>0.69</td>
<td>0.89</td>
<td>0.90</td>
<td>0.92</td>
</tr>
</tbody>
</table>

Because of the space constraints, the other simulation results are not reported in detail here. Yet, the following trends were observed in those results. When the mean number of data items is increased from 1 to 5, all three algorithms have shown deteriorating performances. This is a somewhat expected result. That is, jobs need more than one data item to be copied to the sites where they are scheduled. While keeping the network resources at a fixed capacity, meeting the deadlines of increasing number of data transfer requests under stringent timing constraints is difficult to maintain.

When the mean link bandwidth is increased from 25 Mbytes/sec to 200 Mbytes/sec, all three algorithms have shown increasing performances. While keeping the deadlines close to some fixed value, increasing the mean link bandwidths enables the data transfer times will get shorter. Thus, jobs will get higher chance of meeting their deadlines, as observed in the table.

6 CONCLUSIONS

This study presented a novel all-in-one real-time Data Grid simulator supporting advance reservation to test the real time performance of the job scheduling and data dissemination algorithms in a Data Grid system. The design of DGridSim provides a flexible environment for researchers working on real-time characteristics of Data Grid systems. It should be emphasized that the presented results prove the correct operation of DGridSim. The development of DGridSim continues in several avenues, including the support for centralized and distributed Grid models, new heuristics for job and data scheduling and data replication.

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REFERENCES


