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Improving the Performance of Condor for Dag-Based Tasks through Cache Memory in Grid Environments

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ABSTRACT: Grid application consists of one or more tasks. These tasks are generally divided into two types such as Independent task and Workflow tasks. A number of resource brokers that act as middleware such as Condor, Nimrod-G etc are used in grid based application. Since, workflow tasks are not easily executed with these architecture, we slightly modified the DAG-Condor architecture to suit workflow tasks. This paper focuses on cache management system in DAG-Condor resource broker. By adding cache to store intermediate results and these results also stored in central database which contains old results. The overhead of read and writes to central database is reduced. It improves the performance, throughput of the system.

KEY WORDS: Grid Computing, Condor, Caching, Nimrod-G, Cluster, DAG, Workflow.

I. INTRODUCTION

Grid computing [1] is the current hot topic in Information Technology field. Lot of research is going for efficient utilization of grid. Most of the applications require high processing speed, lot of CPU time and memory.

Scheduling is the central part of grid computing [1]. It can be performed as simply as taking the next available resource, but often this task involves prioritizing job queues, managing the load, finding the workarounds when encountering reserved resources and monitoring progress. Job schedulers are able to submit, control and monitor the workload of jobs submitted in a network of computers.

Grid broker mediates access to distributed resources by (a) discovering suitable data sources for a given analysis scenario, (b) suitable computational resources, (c) optimally mapping analysis jobs to resources, (d) deploying and monitoring job execution on selected resources, and (e) accessing data from local or remote data source during job execution. A resource broker in a data grid must have the capability to locate and retrieve the required data from multiple data sources and to redirect the output to storage where it can be retrieved by processes downstream. It must also have the ability to select the best data repositories from multiple sites based on availability of files and quality of data transfer.

Most of the applications that use a grid have Workflow tasks. To utilize the grid for workflow tasks we have many schedulers and resource brokers like Condor and Nimrod-G. But those are not efficient for workflow tasks for instance DAG-Condor scheduler frequently queries the database for the intermediate results and the number of queries and the size of database results in extra overhead. The performance of system depends on other factors such as the numbers of machines in the cluster and the workload. This creates problems like congestion, and degradation of throughput and performance. To overcome the above problem we proposed architecture by adding cache to store intermediate results and these results also stored in central database which contains old results. To increase the efficiency and performance we have proposed cache



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management system in DAG-Condor resource broker.

II. RELATED WORK

A Workflow is a collection of Steps and data that define the paths that can be taken to complete a task. Workflows may contain activities such as displaying content to users, collecting information from users or computer systems, performing calculations, and sending messages to external computer systems. To maintain the workflow jobs we have collection of resource brokers such as Nimrod-G, Condor-G, AppLeS, Gridbus Broker, Globus.

The Nimrod-G [2] is a Grid resource broker that allows managing and steering task farming applications on computational Grids. It uses an economic model for resource management and scheduling. Nimrod-G provides resource discovery, resource trading, scheduling, resource staging on Grid nodes, result gathering, and final presentation to the user. Nimrod-G usesGRid Architecture for Computational Economy (GRACE) services to dynamically trade with resource owner agents to select appropriate resources. It follows the hierarchical and computational market model in resource management. It uses the services of Grid middleware systems such as Globus and Legion for resource discovery and uses either a network directory or object model based data organization. Nimrod/G is a resource management system with a focus on computational economy and schedules tasks based on their deadlines and budgets. Nimrod/G also addresses issues of scheduling single jobs, and does not address the requirements of workflow applications.

The Condor is a Meta computing system that uses computer idle time to run jobs in a network. Given the computing resources of an institution, Condor seeks to maximize job throughput, without disturbing human interaction. The Condor environment follows a layered architecture and offers powerful and flexible resource management services for sequential and parallel applications. The Condor has been extended to support submission of jobs to resources Grid-enabled using Globus services. The matchmaker is responsible for initiating contact between compatible agents.

The Condor-G [6] is the job management part of Condor. Condor-G helps us to submit jobs into a queue, have a log detailing the life cycle of our jobs, manage all input and output files, along with everything else you expect from a job queuing system. Condor-G gets its name from how it talks to the resource management part. Condor-G uses the Globus Toolkit(tm) to start the job on the remote machine. Condor-G provides a "window to the Grid" for users to both access resources and manage jobs running on remote resources. Condor-G is used to look across the Grid and see instantly how the jobs are doing.

The database is resident at the server and transactions are initiated from client sites, with the server providing facilities for shared data access. Dynamic local Caching of query results at client sites can enhance the overall performance of such a system, especially when the operational data spaces of clients are mostly disjoint. In effect, such caching of locally pertinent and frequently used data constitutes a form of dynamic data replication, whereby each client dynamically defines its own data space of interest. This concept is utilized in this proposed system.

III. WORKFLOW TASK

A workflow (fig.1) is composed of connected multiple scientific tasks according to their dependencies. Workflow [11] structure indicates the temporal relationship between the tasks. In general, a workflow can be represented as a Directed Acyclic Graph (DAG)[9,10] or a non-DAG.

An acyclic digraph is a directed graph containing no directed cycles, also known as a **directed acyclic graph** or a "**DAG**." A workflow language is a particular XML notation representing the inter-task dependencies.



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In DAG-based workflow, workflow structure can be categorized into three types such as sequence, parallelism, and choice. Sequence is defined as an ordered series of tasks, with one task starting after a previous task has completed. Parallelism represents tasks which are performed concurrently, rather than serially. In choice structured workflows, a task is selected to execute at run-time when its associated conditions are true. In addition to all structures contained in a DAG-based, a non-DAG workflow also includes iteration structure, in which sections of workflow tasks in an iteration block are allowed to be repeated.

3.1 *Nature of scientific workflows:* In workflow, DAG (Directed Acyclic Graph) [11] refers to a set of programs with dependencies between them i.e the input of some programs may depend on the output of others. This places partial constraints on the order of execution of programs. Consider the "diamond" DAG shown in Figure 2. Since programs B and C depend on the output of program A, they can be run only after A completes, but B and C themselves can be run in any order or even in parallel.

A typical scientific workflow frequently consists of a number of such DAGs with identical structures. These DAGs may share some common (global) input data. Scientific workflows are characterized by the different types of I/O performed by jobs in their lifetime. Pipeline I/O refers to the data flow that occurs between parent and child programs within a particular DAG. The term batch I/O refers to input files that are common across all DAGs in a workflow. In Figure 2 files 'File1' and 'File2' are pipeline files. 'FileInput' is a batch input to the DAGs in the example workflow shown in Figure 3. Batch input can occur at any level in a DAG.



Fig.2 A Diamond DAG



Fig.3 Workflow from Diamond DAGs

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IV. JOBS EXECUTION IN CONDOR

Condor runs on a heterogeneous set of computers, the Condor pool. Each computer executes two daemons, the job scheduler daemon (Schedd) and the starter daemon (Startd), which launches new jobs.

In Condor, users create submit files for jobs to be executed. The submit file specifies details about the job such as the names of the executable, input and output files, and environment variables that need to be set at the time of job execution. In addition to that a user may also specify *requirements* and *rank* attributes. The *requirements* attribute places constraints on the machines that can run the job. *Rank* is used to specify an order of preference for machines that meet the job's requirements. A job is submitted by specifying a submit file to the *condor_ submit* tool. The submit file is converted into a "classad", which is a list of attribute-value pairs. Machines also advertise their resources using classads and can place constraints on the jobs they wish to run. In the *matchmaking* process each user is allotted some number of the machines in the pool based on a fair-share scheme [8]. Jobs that are selected for scheduling are first sorted in order of user priority. For each job, the list of available machines is scanned and the machine with the highest rank that satisfies the job's requirements is chosen for execution.(This machine is called the execute machine). The matchmaking process is done periodically (typically, every five minutes).

The *condor_submit* tool only accepts jobs consisting of a single program. Workflows consisting of DAGs are submitted using a different tool called *condor_submit_dag*. This tool allows users to specify parent-child dependencies between jobs. When a workflow is submitted, the Condor DAGMan daemon is spawned, and the top-level jobs in the workflow (jobs marked 'A' in Figure 3) are submitted using *condor_submit*. DAGMan continuously monitors the logs produced by Condor on the submit machine (the machine to which the user submits jobs) and uses *condo_submit* to submit a child job once its parents have completed execution. So the Condor matchmaker itself has no notion of workflows and cannot consider a job for scheduling until it is ready for execution.

Before a job begins, it's input and executable files are transferred from the submit machine to the execute machine. After a job has completed, the output files associated with that job are transferred back to the submit machine and deleted from the execute machine. When a child job is scheduled, files are again transferred from the submit machine to an execute machine (which could possibly be the same one that executed the parent job). This job scheduling and file transfer system is inefficient since the data dependencies within and across DAGs in a workflow are ignored. The main disadvantages are listed below;

- All jobs has to be submitted through submit machine only
- All the client machines send and receive intermediate results through central Database
- Congestion is high in condor
- Performance degrades as the number of tasks increases
- Data or internal results are not used in future
- It reduces throughput and increase the execution time.

V. PROPOSED CACHE MANAG-EMENT SYSTEM IN DAG-CONDOR

The Condor architecture has a database which is connected to submit machine, so all the jobs will be submitted through the submit machine only. To over come this problem we propose the new DAG-Condor Architecture (fig.5) in which the central database is connected to both submit and execute machine. Here we can submit any job from any part of the grid either through submit machine or execute machine.



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Fig.5 Stage 1 proposed DAG-Condor architecture

Previously in the DAG-Condor architecture, due to the presence of a central database the overhead of accessing the database increase as and when intermediate results are generated. It increases the read and writes overhead. This in turns reduces the performance of the system. Due to the increase of read and write operations it also increases the congestion, cost of operation and increases the loss of data. To reduce the loss of data retransmission of data will increase. To reduce the overhead of read and write operations we include the high storage cache which is connected to all the execute machine clusters in the proposed system. In the cluster, we use hierarchy cache management which stores all the internal results of machines connected in cluster. These results are stored in cache as well as in central Database.

If any cluster require the results of workflow jobs which are executed in another cluster it will be accessed either from cache or central database based on the distance between the cache and central database. The distance between the systems will be calculated using Link-State algorithm [16] which is explained below.



Fig.6 Cache management in Proposed DAG-Condor Architecture

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Suppose we have four clusters A, B, C, D and central database CD in grid system. Figure 7 shows that the system is connected between the clusters as follows (A,B), (B,C), (B,D), (C,D), (A,CD),(D,CD) with distance 3,1,5,1,2,1 respectively. Let job x be executed at cluster D, the result is stored in cache and central database. An execute machine in cluster B requires the result of job x which is stored in central database and cache at cluster D. The question is that, from which memory it has to access the results .By using the Link–state Algorithm it finds that, access from cache memory at cluster D is between B, D is Shorter than B to CD



Fig.7 overall structure of system

V. EMPIRICAL RESULTS

We have simulated the architecture in java which gives good results .It gives good performance when compared to the previous architecture of Condor. Here the execution time is reduced compare to previous one, by reducing the read operation from the central database. It gives High performance and reduces the congestion due to cache management. We calculated the performance based on page faults. It reduces the number of page faults thus increases the performance of the system.

In the below, the graph results are based on the probability of read operation if the probability of read operations increases the page faults increases. This system is performs better when there are less read operations to the central database.

Number of Page	Prob. Of read
faults	operation
54	0.2
58	0.4
48	0.5
58	0.6
59	0.8
60	0.9
80	1



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Fig.8. Page faults Vs Read operations

VI. CONCLUSION AND FUTURE WORK

In this paper we introduce the cache management system in DAG-Condor architecture. This makes the easy transmission of data throughout the system. A major advantage is associative access to the contents of a cache, allowing effective reuse of cached information. Increased autonomy at client sites, less network traffic, and better scalability are a few other expected benefits.

Apart from the planned performance studies, many other important issues remain unexplored in this paper. We currently work on the architecture of cache management in DAG-Condor. The future implementation questions in cache are to derive suitable predicate-indexing techniques, optimization strategies, performance tuning, local index creation, and effective management of space by a client.

REFERENCES

1. I. Foster and C. Kesselman (editors), The Grid: Blueprint for a Future Computing Infrastructure, Morgan Kaufmann Publishers, USA, 1999.

2. D. Abramson, J. Giddy, and L. Kotler, High Performance Parametric Modeling with Nimrod/G: Killer Application for the Global

Proceedings of the International Parallel and Distributed Processing Symposium (IPDPS 2000), May 1-5, 2000, Cancun, Mexico, IEEE CS Press, USA, 2000.

3. Ncbi blast. <u>http://www.ncbi.nlm.nih.gov/BLAST/</u>.

4. Condor services for the Global Grid : Interoperability between Condor and OGSA

5. Litzkow, M., Livny, M., and Mutka, M., "Condor - A Hunter of Idle Workstations", Proc. 8th Intl Conf. on Distributed Computing Systems, 1988,

pp. 104-111.

6. Condor-G: A Computation Management Agent for Multi-Institutional Grids

James Frey, Todd Tannenbaum, Miron Livny Ian Foster, Steven Tuecke

7. <u>http://www.cs.wisc.edu/condor/manual/</u>

8. Condor fair share scheduling. http://www.cs.wisc.edu/condor/manual/v6.7/ 3 5User Priorities.html.

9. J. Blythe, S. Jain, E. Deelman, Y. Gil, K. Vahi, A. Mandal, and K. Kennedy. Resource Allocation Strategies for Workflows in Grids In IEEE International Symposium on Cluster Computing and the Grid (CCGrid 2005).

10. A. Mandal, K. Kennedy, C. Koelbel, G. Marin, J. Mellor-Crummey, B. Liu and L. Johnsson. Scheduling Strategies for Mapping Application Workflows onto the Grid. In IEEE International Symposium on High Performance Distributed Computing (HPDC 2005),2005.

11. Data Driven Workflow Planning in Cluster Management Systems HPDC'07, June 25–29, 2009, Monterey, California, USA.Copyright 2009

12. A predicate-based caching scheme for client-server database architectures Arthur M. Keller1 and Julie Basu2;3

13. Kamel N, King R (1992) intelligent database caching through the use of page-answers and page-traces. ACM Trans Database Syst 17:601-646

14. Wang Y, Rowe LA (1991) Cache consistency and concurrency control in a client-server DBMS architecture. Proceedings of the ACMSIGMOD International Conference on Management of Data, Denver, Colo, May

15. Run-time Adaptive Cache Hierarchy Management via Reference Analysis

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16. Link-State Routing with Hop-by-Hop Forwarding Can Achieve Optimal Traffic Engineering. Dahai Xu* AT&T Labs - Research ,Mung Chiang Dept. of EE, Princeton University, Jennifer Rexford Dept. of CS, Princeton University 17. Secure Link State Routing for Mobile Ad Hoc Networks

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