



DEPARTMENT OF INFORMATICS

TECHNISCHE UNIVERSITÄT MÜNCHEN

Master's Thesis in Informatics

**Job Scheduling for Adaptive Applications
in Future HPC Systems**

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**Job Scheduling für Adaptive
Anwendungen auf Zukünftigen HPC
Systemen**

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Submission Date: May 15, 2015



I confirm that this master's thesis in informatics is my own work and I have documented all sources and material used.

Munich, May 15, 2015

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Acknowledgments

Abstract

Invasive Computing is a novel paradigm for the design and resource-aware programming of future parallel computing systems. It enables the programmer to write resource aware programs and the goal is to optimize the program for the available resources. Traditionally, parallel applications implemented using MPI are executed with a fixed number of MPI processes before submitting to a HPC(High Performance Computing) system. This results in a fixed allocation of resources for the job. Modern techniques in scientific computing such as AMR(Adaptive Mesh Refinement) result in applications exhibiting complex behaviors where their resource requirements change during execution. Invasive MPI which is a part of an ongoing research effort to provide MPI extensions for the development of Invasive MPI applications will result in jobs that are resource-aware for the HPC systems and can utilize such AMR techniques. Unfortunately, using only static allocations result in these applications being forced to execute using their maximum resource requirements that may lead to an inefficient resource utilization. In order to support such kind of parallel applications at HPC centers there is an urgent need to investigate and implement extensions to existing resource management systems or replace it with a new one. This thesis will extend the work done over the last few months during which an early prototype was implemented by developing a protocol for the integration of invasive resource management into existing standard batch systems. Specifically, This thesis will now investigate and implement a job scheduling algorithm in accordance with the new protocol developed earlier for supporting such an invasive resource management.

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1 Introduction

Over the last two decades, the landscape of Computer Architecture has changed radically from sequential to parallel . Due to the limiting factors of technology we have moved from single core processors to multi core processors having a network interconnecting them. Traditionally, the approach of designing algorithms has been sequential, but designing algorithms in parallel is gaining more importance now to better utilize the computing power available at our disposal. Another important trend that has changed the face of computing is an enormous increase in the capabilities of the networks that connect computers with regards to speed, reliability etc. These trends make it feasible to develop applications that use physically distributed resources as if they were part of the same computer. A typical application of this sort may utilize processors on multiple remote computers, access a selection of remote databases, perform rendering on one or more graphics computers, and provide real-time output and control on a workstation. Computing on networked computers ("Distributed Computing") is not just a subfield of parallel computing as the basic task of developing programs that can run on many computers at once is a parallel computing problem. In this respect, the previously distinct worlds of parallel and distributed computing are converging.

As technology advances, we have newer problems or applications that demand larger computing capabilities which push the limits of technology giving rise to newer advancements. The performance of a computer depends directly on the time required to perform a basic operation and the number of these basic operations that can be performed concurrently. A metric used to quantify the performance of a computer is FLOPS (floating point operations per second). The time to perform a basic operation is ultimately limited by the "clock cycle" of the processor, that is, the time required to perform the most primitive operation. The term *High Performance Computing (HPC)* refers to the practice of aggregating computing power (multiple nodes with processing units interconnected by a network in a certain topology) or the use of parallel processing for running advanced application programs efficiently, reliably and quickly. The term applies especially to systems that function above a *teraflop* or 10^{12} floating-point operations per second. The term HPC is occasionally used as a synonym for Supercomputer that works at more than a *petaflop* or 10^{15} floating-point operations per second. The most common users of HPC systems are scientific researchers, engineers, government

agencies including the military, and academic institutions. In general, HPC systems can refer to Clusters, Supercomputers, Grid Computing etc. and they are usually used for running complex applications.

A **Batch System** is used to manage the resources in a HPC System. It is a middleware that comprises of two major components namely the **Resource Manager** and **Scheduler**. The role of a Resource Manager is to act like a glue for a parallel computer to execute parallel jobs. It should make a parallel computer as easy to use as a Personal Computer (PC). A programming model such as **Message Passing Interface (MPI)** for programming on distributed memory systems would typically be used to manage communications within a parallel program by using the MPI library functions. A Resource Manager allocates resources within a HPC system, launches and otherwise manages Jobs. Some of the examples of widely used open source as well as commercial resource managers are **SLURM**, **TORQUE**, **OMEGA**, **IBM Platform LSF** etc. Together with a scheduler it is termed as a Batch System. The role of a job scheduler is to manage queue(s) of work when there is more work than resources. It supports complex scheduling algorithms which are optimized for network topology, energy efficiency, fair share scheduling, advanced reservations, preemption, gang scheduling (time-slicing jobs) etc. It also supports resource limits (by queue, user, group, etc.). Many batch systems provide both resource management and job scheduling within a single product (e.g. LSF) while others use distinct products(e.g. Torque Resource Manager and Moab Job Scheduler). Some other examples of Job Scheduling Systems are **LoadLeveler**, **OAR**, **Maui**, **SLURM** etc.

Existing Batch Systems usually support only static allocation of resources to an application before they start which means the resources once allocated are fixed for the lifetime of the application. The complexity of applications have been growing, However, especially when we consider advanced techniques in Scientific Computing like **Adaptive Mesh Refinement (AMR)** where applications exhibit complex behavior by changing their resource requirements during execution. The Batch Systems of today are not equipped to deal with such kind of complex applications in an intelligent manner apart from giving them the maximum number of resources before it starts that will result in a sheer wastage of resources leading to a poor resource utilization. In order to support such adaptive applications at HPC centers there is an urgent need to investigate and implement extensions to existing resource management systems or develop an entirely new system. These supporting infrastructures must be able to handle the new kind of applications and the legacy ones intelligently keeping in mind that they should now be able to achieve much higher system utilization, throughput, energy efficiency etc. compared to their predecessors due to the elasticity of the applications.

1.1 Invasive Computing

The throughput of HPC Systems depends not only on efficient job scheduling but also on the type of jobs forming the workload. As defined by Feitelson, and Rudolph, Jobs can be classified into four categories based on their flexibility:

- **Rigid Job:** Requires a fixed number of resources throughout its execution.
- **Moldable Job:** The resource requirement of the job can be molded or modified by the batch system before starting the job(e.g. to effectively fit alongside other rigid jobs). Once started its resource set cannot be changed anymore.
- **Evolving Job:** These kind of jobs request for resource expansion or shrinkage during their execution. Applications that use Multi-Scale Analysis or Adaptive Mesh Refinement (AMR) exhibit this kind of behavior typically due to unexpected increases in computations or having reached hardware limits (e.g. memory) on a node.
- **Malleable Job:** The expansion and shrinkage of resources are initiated by the batch system in contrast to the evolving jobs. The application adapts itself to the changing resource set.

The first two types fall into the category of what is called as the static allocation since the allocation of rigid and moldable jobs must be finalized before the job starts. Whereas, the last two types fall under the category of dynamic allocation since this property of expanding or shrinking evolving and malleable jobs (together termed adaptive jobs) happens at runtime. Adaptive Jobs hold a strong potential to obtain high system performance. Batch systems can substantially improve the system utilization, throughput and response times with efficient shrink/expand strategies for running jobs that are adaptive. Similarly, applications also profit when expanded with additional resources as this can increase application speedup and improve load balance across the job's resource set.

Invasive Computing is a novel paradigm for the design and resource-aware programming of future parallel computing systems. It enables the programmer to write efficient resource aware programs. This approach can be used to allocate, execute on and free resources during execution of the program. The result is an adaptive application which can expand and shrink in the number of its resources at runtime. HPC infrastructures like clusters, supercomputers execute a vast variety of jobs, majority of which are parallel applications. These centers use intelligent resource management systems that should not only perform tasks of job management, resource management and scheduling but also satisfy important metrics like higher system utilization, job throughput

and responsiveness. Traditionally, MPI applications are executed with a fixed number of MPI processes but with Invasive MPI they can evolve dynamically at runtime in the number of their MPI processes. This in turn supports advanced techniques like AMR where the working set size of applications change at runtime. Such kind of adaptive programming paradigms need to be complemented with intelligent resource management systems that can achieve much higher system utilization, energy efficiency, throughput etc. compared to their predecessors due to elasticity of the applications.

Under the collaborative research project funded by the **German Research Foundation (DFG)** in the **Transregional Collaborative Research Centre 89 (TRR89)**, research efforts are being made to investigate this Invasive Computing approach at different levels of abstraction right from the hardware up to the programming model and its applications. **Invasive MPI** is an effort towards invasive programming with MPI where the application programmer has MPI extensions available for specifying at certain safe points in the program, the possibility of a changing the resource set of the application during runtime.

1.2 Dynamic Resource Management

Two of the most widely used resource managers on HPC systems are **SLURM** and **TORQUE**. The two major components in general of any sophisticated resource manager are the batch scheduler and the process manager. The Process Manager is responsible for launching the jobs on the allocated resources and managing them throughout their lifetime. Examples of process manager are *Hydra*, *SLURM Daemon (slurmd)* etc. The process managers interact with the processes of a parallel application via the **Process Management Interface (PMI)**. In order to support Invasive Resource Management, The following components will be implemented: *iScheduler* (Batch Scheduler for Invasive Jobs) built as an extension into an existing batch system and *iDRScheduler* (Invasive Distributed Run Time Scheduler) similar to a controller daemon which will sit between the batch scheduler and the process manager. **SLURM** is the choice of an existing batch system on which this prototype will be implemented for demonstrating Invasive Computing and 1.1 shows a high level illustration of the architecture for such an Invasive Resource Management.

The above figure illustrates the proposed invasive resource management architecture. In addition to a job queue for legacy static jobs, we now have an additional job queue for invasic jobs. The existing batch scheduler needs to be extended in order to schedule

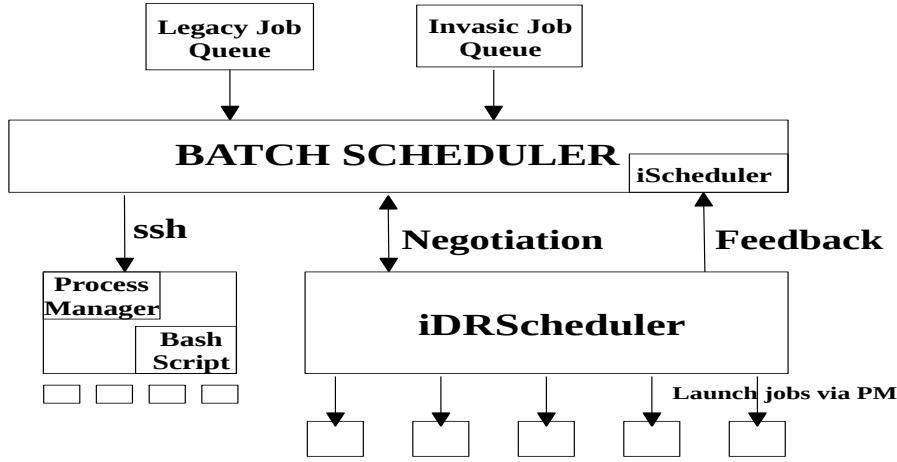


Figure 1.1: Invasive Resource Management Architecture

these new type of Invasive Jobs and a new component called *Invasive Scheduler (iScheduler)* is responsible for this. In contrast to modifying an existing system to support Invasive Resource Management, a new component called *Invasive Distributed Run Time Scheduler (iDRScheduler)* is proposed which sits between the Batch Scheduler and Process Manager. The objective of such a multi-level approach is to avoid modifying the existing system which will be a substantially large effort and rather have an independent component that caters specifically to Invasive Jobs. It is responsible for managing the resources present in the Invasive partition used specifically for running Invasive Jobs. With this approach, the existing Legacy Jobs can be served via the existing batch scheduler and the new Invasive Jobs can be served by via iScheduler and iDRScheduler. iDRScheduler talks to the iScheduler via a protocol called the Negotiation protocol to receive Invasive Jobs dispatched from iScheduler which it then launches for execution by performing some run time scheduling like pinning of jobs, expand/shrink etc.

The new components proposed in the architecture help achieve the objective of supporting dynamic resource management for invasive applications. iScheduler is responsible for scheduling Invasive Jobs. The scheduling decisions are communicated via negotiation protocol to iDRScheduler and these decisions are basically job(s) selected via a scheduling algorithm to be submitted for execution. The decisions will be made on the basis of Resource Offers sent by iDRScheduler which creates these resource offers based on the state of the partition. Upon receiving a resource offer, iScheduler will either accept it by selecting jobs from the queue that can be mapped to this offer or reject it. A Resource Offer can represent real or virtual resources because the iDRScheduler can also present a virtual view of resources in the hope of getting a mapping of jobs to offer that is

more suitable to satisfy its local metrics such as resource utilization, power stability, energy efficiency etc. It can either accept or reject the mapping received from iScheduler. Similar to iDRScheduler, the iScheduler makes its decisions to optimize for certain local metrics such as high job throughput, reduced job waiting times, deadlines, priorities etc. This highlights the mismatching policies/metrics for which both the iDRScheduler and iScheduler make their decisions on and hence both will be involved in some kind of a negotiation via the protocol to reach a common agreement. iDRScheduler is an independent entity introduced with the purpose of inter-operating with existing batch systems rather than replacing them with an entirely new one. It may be possible that in the future this component will not be a separate entity but will be built into the batch system itself.

Shrink/Expand Strategies: There are several strategies one may employ to tackle adaptive applications during runtime and take decisions that can lead to higher system utilisation, energy efficiency etc.:

- Let us consider that at any given point of time there are some Invasive Jobs running in the system. If the parallel runtime environment is able to anticipate that in the near future, there may be a large window of free resources available because some applications may shrink according to a prediction of their scalability behavior with the help of run time profiling and collected performance data then iDRScheduler can provide a resource offer to iScheduler. This offer can specify a virtual list of nodes more than what is available in order to get a mapping of jobs. These jobs can then be shrunk and fit into the existing space of resources with the knowledge that they may be able to expand later.
- Another scenario is where there is an anticipation of a smaller window of resources in the future because some of the currently running applications may expand. In such a case iDRScheduler will provide a resource offer to iScheduler with a virtual list of nodes smaller than what is currently available in order to get a mapping of jobs. These jobs can then be expanded and fit into the existing space of resources with the knowledge that they may have to shrink later.
- The third variation could be the case where the runtime anticipates the state of resources to remain the same as the current state for the near future since none of the running applications may expand or shrink. In such a scenario iDRScheduler will send a resource offer that is a list of nodes which is exactly the same as the available resources to iScheduler in order to get a mapping of jobs. These jobs can then be fit into the space of resources available as it is without expanding/shrinking them.

1.3 Master Thesis

The focus of this Master Thesis is to extend the early prototype developed as a part of the guided research in the last few months. It will now give concrete meanings to the negotiation protocol, defining the format of the invasic job records, its constraints, definition of resource offers, feedback reports and most important of all to investigate and develop an efficient job scheduling algorithm at the iScheduler level. Following this closely from the Guided Research would be to continue having an automated testing in place that will help in simulating a workload of jobs, testing the job scheduling algorithm for its correctness, evaluating and analysing the performance of such a prototype for various metrics. Given below is a tentative plan of the activities proposed monthwise starting from November 15 till May 15 to be carried out for the duration of 6 months in this Master Thesis.

1.3.1 Tentative Plan

- Literature survey for current/recent related work on batch job scheduling from research groups focusing on problems in the areas of batch scheduling, resource management, middleware etc. In addition to this, defining resource offers, invasic job records and job constraints that can come from a pre-computed performance model of an application during its runtime or could be user defined. Once these tasks have been completed, it will follow up with extending the early prototype by the implementation of these ideas and testing them for correctness using the automated testing feature.
- Integration of the fake iHypervisor with the real iHypervisor by making it as its plugin. Test this integration for all the earlier development using the automated testing feature for correctness. Review, analyse, fix any errors and repeat the process till the integration is stable. Follow this up by understanding the SLURM's existing scheduling algorithms including the recent Machine Learning approach for a possible reuse or enhancement. Define the format of feedback reports sent by iHypervisor and how they can be processed on the iScheduler side including its storage mechanism (possibly using SLURM's existing database functionalities). Implement these ideas in the prototype and test it thoroughly.
- Investigate and experiment with basic scheduling algorithms first to later proceed with complex ones. Design and implement a sophisticated scheduling algorithm for mapping jobs from Invasic job queue to the resource offers sent by iHypervisor. This algorithm must perform its decision making by also using the collected

history/feedback data about many statistical measures provided by iHypervisor periodically.

- Implementation of the Job Scheduling Algorithm continues for realizing all the necessary requirements as proposed earlier till it has been successfully implemented and later followed by a lot of functional testing.
- Test the full prototype along with the other component iHypervisor (Including its Run Time scheduling and Invasive Resource Management) for simple to complex workloads (emulating a queue of invasive jobs possibly including legacy static jobs). Testing also needs to be done with real Invasive applications developed by the *Chair Of Scientific Computing*. Find issues/errors, review, analyse, correct and repeat the process till it is stable and then collect all the results necessary.
- Coming up with the draft version of the Master Thesis Report followed by reviews, corrections. This process is repeated till the final version is decided. Also prepare the slides for the Master thesis to present them at a later stage.

The above timeline highlights a tentative plan for the activities to be taken up during the Master Thesis and the 6 items above correspond to these 6 months of the Thesis in a chronological order. The steps may overlap or shift depending on the progress but the same overall structure will be followed for the Thesis. It will also include in parallel small amounts of documentation in the report as and when necessary during the 6 month period and not necessarily everything at the end.

1.4 Motivation

1.5 Document Structure

This is end of the first section which gave an introduction to this Master Thesis and the kind of problem it deals with. The rest of this report is organized as follows:

- **Modeling:** This section will explain in brief on how the DNDP will be mathematically formulated using a bilevel linear program with all its constraints, decision variables and objective function. It will explain the approach to approximate the non-linear objective function of the inner problem of DNDP which is TAP and an introduction to metaheuristics that will be used to solve such kind of a combinatorial optimization problem.

- **Genetic Algorithm:** This section will dive into the details of the genetic algorithm that includes all the important aspects which fall under GA that needs to be tackled in order to come up with a correct implementation yielding good performance. It will explain in detail some of the many choices one has for implementation during every step of GA.
- **Implementation:** This section will illustrate with the help of a flow chart the high level view of the GA implementation followed by some pseudo codes to demonstrate in a simple language the implementation details of the inner workings of GA.
- **Experiments and Visualization:** This section will present all the results of the experiments conducted on different types(sizes) of datasets using the implemented GA in the form of tables. It will also mention in brief some of the observations and inferences that have been drawn by looking at these results.
- **Conclusion:** This section concludes the report on this project with a highlight of what was successfully achieved along with the possible scope of what can be done as a part of future research work. This is followed by a list of some useful references that played an important role in the understanding of many of the concepts towards the realization of this project.

2 Related Work

2.1 Batch Scheduling

2.2 Runtime Scheduling

2.3 Adaptive Programming Paradigms

3 Invasive Computing

3.1 Job Classification

3.2 Resource Aware Programming

3.3 Traditional Resource Management

The role of a resource manager is to acts like a *glue* for a parallel computer to execute parallel jobs. It should make a parallel computer as easy to use as almost a PC. MPI would typically be used to manage communications within the parallel program. A resource manager allocates resources within a cluster, launches and otherwise manages jobs. Some of the examples of widely used open source as well as commercial resource managers are **SLURM**, **TORQUE**, **OMEGA**, **IBM Platform LSF** etc. Together with a scheduler it is termed as a *Batch System*. The Batch System serves as a middleware for managing supercomputing resources. The combination of *Scheduler+Resource Manager* makes it possible to run parallel jobs.

The role of a job scheduler is to manage queue(s) of work when there is more work than resources. It supports complex scheduling algorithms which are optimized for network topology, energy efficiency, fair share scheduling, advanced reservations, preemption, gang scheduling(time-slicing jobs) etc. It also supports resource limits(by queue, user, group, etc.). Many batch systems provide both resource management and job scheduling within a single product (e.g. LSF) while others use distinct products(e.g. Torque resource manager and Moab job scheduler). Some other examples of Job scheduling systems are **LoadLeveler**, **OAR**, **Maui**, **SLURM** etc.

3.3.1 Classification

Before the classification begins we define some terms that are used in the following.

- The term *scheduling* stands for the process of computing a schedule. This may be done by a queuing or planning based scheduler.

- A resource request contains two information fields: the number of requested resources and a duration for how long the resources are requested for.
- A job consists of a resource request plus additional information about the associated application. Examples are: information about the processing environment (e. g. MPI or PVM), file I/O and redirection of stdout and stderr streams, the path and executable of the application, or startup parameters for the application. We neglect the fact that some of these extra job data may indeed be needed by the scheduler, e. g. to check the number of available licenses.
- A reservation request is a resource request starting at a specified time for a given duration. Once the scheduler accepted such a request, it is a reservation.

We call a reservation Fix-Time request to emphasize that it can not be shifted on the time axis during scheduling. Accordingly we call a resource request Var-Time request, as the scheduler can move the request on the time axis to an earlier or later time according to the used scheduling policy. In the following we focus on resource management systems that use space-sharing.

The criterion for the differentiation of resource management systems is the planned time frame [42]. Queuing systems try to utilize currently free resources with waiting resource requests. Future resource planning for all waiting requests is not done. Hence, waiting resource requests have no proposed start time. Planning systems in contrast, plan for the present and future. Planned start times are assigned to all requests and a complete schedule about the future resource usage is computed and made available to the users. A comprehensive overview is given in Table 3.2 at the end of this section.

Queuing Systems Today almost all resource management systems fall into the category of queuing systems. Several queues with different limits on the number of requested resources and the duration exist for the submission of resource requests. Jobs within a queue are ordered according to a scheduling policy, e. g. FCFS (first come, first serve). Queues might be activated only for specific times (e. g. prime time, non prime time, or weekend). The task of a queuing system is to assign free resources to waiting requests. The highest prioritized request is always the queue head. If it is possible to start more than one queue head, further criteria like queue priority or best fit (e. g. leaving less resources idle) are used to select a request. There might also exist a high priority queue whose jobs are preferred at any time. If not enough resources are available to start any of the queue heads, the system waits until enough resources become available. These idle resources may be utilized with less prioritized requests by backfilling mechanisms.

As the queuing systems time frame is the present, no planning of the future is done. Hence, no information about future job starts are available. Consequently guarantees can not be given and resources can not be reserved in advance. However, if participating in grid environments this functionality is desirable. Using reservations eases the way of starting a multi-site application which synchronously runs on different sites. By reserving the appropriate resources at all sites, it is guaranteed that all requested resources are available at the requested start time. With queuing systems this still has to be done manually by the administrative staff. Usually high priority queues combined with dummy jobs for delaying other jobs are used.

Users do not necessarily have to specify run time estimates for their jobs, as a queuing system might let jobs run to completion. Obviously users would exploit this by starting very long running jobs which then block parts of the system for a long time. Hence, run time limits were added to the queues (Table 3.1). A longer run time than the limit of the queue is not allowed and the resource management system usually kills such jobs. If the associated application still needs more CPU time, the application has to be checkpointed and later restarted by the user.

Planning Systems Planning systems schedule for the present and future. With assigning start times to all requests a full schedule is generated. It is possible to query start and end times of requests from the system and a graphical representation of the schedule is also possible (Figure 3.1). Obviously duration estimates are mandatory for this planning. With this knowledge advanced reservations are easily made possible. There are no queues in planning systems. Every incoming request is planned immediately.

The re-planning process is the key element of a planning system. Each time a new request is submitted or a running request ends before it was estimated to end, a new schedule has to be computed and this function is invoked. At the beginning of a re-plan all non-running requests are deleted from the schedule and sorted according to the scheduling policy. Then all requests are re-inserted at the earliest possible start time in the schedule. After this step each request is assigned a planned start and end time. The non-running requests are usually stored in a list structure and different sorting criteria are applied. They define the scheduling policy of the system.

As planning systems work with a full schedule and assign start times to all requests, resource usage is guaranteed and advanced reservations are possible. A reservation usually comes with a given start time or if the end time is given the start time is

computed with the estimated run time. When the reservation request is submitted the scheduler checks with the current schedule, whether the reservation can be made or not. That is the amount of requested resources is available from the start time and throughout the complete estimated duration. If the reservation is accepted it is stored in an extra list for accepted reservations. During the re-planning process this list is processed before the list of variable requests. It does not have to be sorted as all reservations are accepted and therefore generate no conflicts in the schedule. Furthermore, additional types of job lists are thinkable which are then integrated in the re-planning process according to their priority (reservations should have the highest and variable requests the lowest priority).

Controlling the usage of the machine as it is done with activating different queues for e. g. prime and non prime time in a queuing system has to be done differently in a planning system. One way is to use time dependent constraints for the planning process, e. g. “during prime time no requests with more than 75 percent of the machines resources are placed”. Also project or user specific limits are possible so that the machine size is virtually decreased. Examples for such limitations are:

- Jobs requesting more than two thirds of the machine are not started during daytime, only at night and on weekends.
- Jobs that are estimated to run for more than two days are only started at weekends.
- It is not possible that three jobs are scheduled at the same time where each job uses one third of the machine.

If an already running request interferes with limits during its run time, it is not prematurely killed. It runs until the estimated end is reached. With the examples from above one could think of a job that requests all resources of a machine, is started on Sunday and runs until Tuesday. Such a job would then block the whole machine on Monday which contradicts the limits for Monday.

Planning systems also have drawbacks. The cost of scheduling is higher than in queuing systems. And as users can view the current schedule and know when their requests are planned, questions like “Why is my request not planned earlier? Look, it would fit in here.” are most likely to occur [50]. Besides the pure and easily measurable performance of the schedule (e. g. utilization or slowdown), other more social and psychologic criteria might also be considered. It might be beneficial to generate a less optimized schedule in favor of having a more understandable schedule. Furthermore, the usage of reservations should be observed, especially if made reservations are really used. Again, users tend to simply reserve resources without really needing and using

them [50]. This can be avoided by automatically releasing unused reservations after a specific idle time.

Table 3.2 shows a summary of the previously described differences between queuing and planning based resource management systems.

3.3.2 Job Scheduling

Scheduling Policies Typical resource management systems store requests in list-like structures. Therefore, a scheduling policy consists of two parts: inserting a new request in the data structure at its submission and taking requests out during the scheduling. Different sorting criteria are used for inserting new requests and some examples are (either in increasing or decreasing order):

- by arrival time: FCFS (first come first serve) uses an increasing order. FCFS is probably the most known and used scheduling policy as it simply appends new requests at the end of the data structure. This requires very little computational effort and the scheduling results are easy to understand: jobs that arrive later are started later. With this example the term fairness is described [79]. In contrast, sorting by decreasing arrival time is not commonly used, as 'first come last served' makes no sense in an on-line scenario with the potential risk of waiting forever (this is also called starvation). However, a stack works with decreasing order of arrival time.
- by duration: Both increasing and decreasing orders are used. Sorting by increasing order leads to SJF (shortest job first) respectively FFIH (first fit increasing height2). Accordingly LJF (longest job first) and FFDH (first fit decreasing height) sort by decreasing run time. In an on-line scenario this requires duration estimates, as the actual duration of jobs are not known at submission time. SJF and LJF are both not fair, as very long (SJF) and short (LJF) jobs potentially wait forever. LJF is commonly known for improving the utilization of a machine.
- by area: The jobs area is the product of the width (requested resources) and length (estimated duration). FFIA (first fit increasing area) is used in the SMART algorithm (Scheduling to Minimize Average Response Time) [91, 76].
- by given job weights: Jobs may come with weights which are used for sorting. Job weights consist of user or system given weights or a combination of both. For example: all jobs receive default weights of one and only very important jobs receive higher weights, i. e. they are scheduled prior to other jobs.

- by the Smith ratio: The Smith ratio of a job is defined by weight area and is used in the PSRS (Preemptive Smith Ratio Scheduling) algorithm [75].
- by many others: e. g. number of requested resources, current slowdown, ...

In the scheduling process jobs are taken out of the ordered data structure for either a direct start in queuing systems or for placing the job in a full schedule (planning system):

- front: The first job in the data structure is always processed. Most scheduling policies use this approach as only with this a sorting policy makes sense. In queuing systems jobs might have to wait until enough resources are available. Planning systems also process the front of the data structure while placing requests as soon as possible. FCFS, SJF, and LJF use this approach.
- first fit: The data structure is traversed from the beginning and always the first job is taken, that matches the search constraints, i. e. requests equal or less resources than currently free.
- best fit: All jobs are tested to see whether they can be scheduled. According to a quality criterion the best suited job is chosen. Commonly the job which leaves the least resources idle in order to increase the utilization is chosen. Of course this approach is more compute intensive as the complete data structure is traversed and tested. If more than one job is best suited an additional rule is required, e. g. always take the first, the longest/shortest job, or the job with the most weight.
- next fit: The SMART algorithm uses this approach in a special case (NFIW) [91, 76].

In general, all combinations are possible but only a few are applicable in practice. Figure 3.2 shows example schedules for FCFS, SJF, LJF, and FFIA. Sorting requests in any order while using first or best fit is not necessary, as the best job is always chosen regardless of its position in the sorted structure. However, a sorting policy could be used to choose one job, if many jobs are equal. Furthermore, best fit comes with the risk of making schedules unfair and opaque for users.

If fairness in common sense has to be met, i. e. the starting order equals the arrival order, only the combination of sorting by increasing arrival time and always processing the front of the job structure can be used. All other combinations do not generate fair schedules. However, such a fair scheduler is not very efficient, as jobs usually have to wait until enough free resources are available. Therefore, basic scheduling policies are extended by backfilling, a method to avoid excessive idleness of resources. Backfilling

became standard in modern resource management systems today. If requests are scheduled out of their sorting order by first or best fit, some form of backfilling is carried out.

EASY Backfilling The default algorithms used by current job schedulers for parallel supercomputers are all rather simple and similar to each other [37], employing a straightforward version of variable partitioning. (Recall that this means space-slicing with static-partitioning, where users specify the number of processors required by their jobs upon submittal.) In essence, schedulers select jobs for execution in first-come first-served (FCFS) order, and run each job to completion, in batch mode. The problem with this simplistic approach is that it causes significant fragmentation, as jobs with arbitrary sizes/arrivals do not pack perfectly. Specifically, if the first queued job requires many processors, it may have to wait a long time until enough are freed. During this time, processors stand idle as they accumulate, despite the fact there may very well be enough of them to accommodate the requirements of other, smaller, waiting jobs.

To solve the problem, most schedulers therefore employ the following algorithm. Whenever the system status changes (job arrivals or terminations), the scheduler scans the queue of waiting jobs in order of arrival (FCFS) and starts the traversed jobs if enough processors are available. Upon reaching the first queued job that cannot be started immediately, the scheduler makes a reservation on its behalf for the earliest future-time at which enough free processors would accumulate to allow it to run. This time is also called the shadow time. The scheduler then continues to scan the queue for smaller jobs (require fewer processors) that have been waiting less, but can be started immediately without interfering with the reservation. In other words, a job is started out of FCFS order only if it terminates before the shadow time and therefore does not delay the first queued job, or if it uses extra processes that would not be needed by the first queued job. The action of selecting smaller jobs for execution before their time provided they do not violate the reservation constraint is called backfilling, and is illustrated in Fig. 1.1 (see detailed description in Section 2.3).

This approach was initially developed for the IBM SP1 supercomputer installed at the Argonne National Laboratory as part of EASY (Extensible Argonne Scheduling sYstem), which was the first backfilling scheduler [98].¹ The term “EASY” later became a synonym for FCFS with backfilling against a reservation associated with the first queued job. (Other backfill variants are described below.) While the basic concept is extremely simple, a comprehensive study involving 5 supercomputers over a period of 11 years has shown that consistent figures of 40–60 percent average utilization have gone up to around 70 percent, once backfilling was introduced [79]. Further, in terms of performance, backfilling was shown to be a close second to more sophisticated

algorithms that involve preemption (time slicing), migration, and dynamic partitioning [19, 170].

User Runtime Estimates The down side of backfilling is that it requires the scheduler to know in advance how long each job will run. This is needed for two reasons:

- to compute the shadow time for the longest-waiting job (e.g. in the example given in Fig. 1.1, we need to know the runtimes of job 1 and job 2 to determine when their processors will be freed in favor of job 3), and
- to know if smaller jobs positioned beyond the head of the wait-queue are short enough to be backfilled (we need to make sure backfilling job 4 will not delay job 3, namely, that job 4 will terminate before the shadow time of job 3).

Therefore, EASY required users to provide a runtime estimate for all submitted jobs [98], and the practice continues to this day. Importantly, jobs that exceed their estimates are killed, so as not to violate subsequent commitments (the reservation). This strict policy has the additional benefit that it supplies an inherent and clear motivation for users to provide high quality estimates , as short enough values increase the chances for backfilling, but too-short values will get jobs prematurely killed.

Popularity of EASY The burden placed on users to provide estimates has not been detrimental. Rather, the combination of simplicity, effectiveness, and FCFS semantics (often perceived as most fair [123]) has made EASY a very attractive and a very popular job scheduling strategy. Nowadays, virtually all major commercial and open-source production schedulers support EASY backfilling [37], including

- IBM’s LoadLeveler [60, 82],
- Cluster Resources’ commercial Moab [118] and open-source Maui [75] (which is probably the most popular scheduler used within the academia),
- Platforms’ LSF (Load Sharing Facility) [172, 24],
- Altair’s PBS (Portable Batch System) [68] in its two flavors: commercial PBS-Pro [33] and open-source OpenPBS [10], and
- Sun’s GridEngine [61, 106]

The default configuration of all these schedulers, except PBS, is either EASY or plain FCFS (with FCFS, however, the schedulers’ behavior becomes EASY if backfilling is nevertheless enabled). The CTO of Cluster Resources has estimated that 90-95 percent

of Maui/Moab installations do not change their default (EASY) settings [74]. Being the exception that implies the rule, the PBS variants use Shortest-Job First (SJF) as their basic default policy. However, even with PBS, when a job is “starved” (a situation defined by PBS to occur if the job is waiting for 24 hours or more) then the scheduling reverts to EASY until this job is started. As a testament for its immense popularity, a survey about the top 50 machines within the top-500 list revealed that, out of the 25 machines for which relevant information was available, 15 (= 60 percent) were operating with backfilling enabled [36].

Variations on Backfilling Despite the simplicity of the concept, backfilling has nevertheless been the focus of dozens of research papers attempting to evaluate and improve the basic idea.³ We do not list them all here, but rather, cite many of them (and more) when appropriate, within their respective contexts later on. The remainder of this section only briefly mentions some of the various tunable knobs of backfilling algorithms.

One tunable parameter is the number of reservations. As mentioned above, in EASY, only the first queued job receives a reservation. Thus, backfilling may cause delays in the execution of other waiting jobs which are not the first and therefore do not get a reservation [47]. The obvious alternative is to allocate reservation to all the jobs. This approach has been named “conservative backfilling” as opposed to the “aggressive” approach taken by EASY [108]. However, it has been shown that delaying other jobs is rarely a problem, and that conservative backfilling tends to achieve reduced performance in comparison to the aggressive alternative. The MAUI scheduler includes a parameter that allows system administrators to set up the number of reservations [75]. It has been suggested that allocating up to four reservations is a good compromise [15].

A second parameter is the looseness of reservations. For example, an intriguing suggestion is a “selective reservation” strategy depending on the extent different jobs have been delayed by previous backfilling decisions. If some job is delayed by too much, a reservation is made for this job [141]. This is somewhat similar to the “flexible backfilling” strategy, in which backfilling is allowed to violate the reservation(s) up to a certain slack [150, 166]. (Setting the slack in the latter strategy to be the threshold used for allocating selective reservations in the former strategy, is more or less equivalent.)

A third parameter is the order of queued jobs. EASY, as well as many other system designs, use FCFS order [98]. A general alternative is to prioritize jobs in a certain way, and select jobs for scheduling (including as candidates of backfilling) according to this priority order. For example, flexible backfilling combines three types of priorities: an administrative priority to favor certain users or projects, a user priority used

to differentiate between the jobs of the same user, and a scheduler priority used to guarantee that no job is starved [150]. The Maui scheduler has a priority function that includes even more components [75]. Another approach is to prioritize based on various job characteristics. In particular, a set of criteria related to the current queueing time and expected resource consumption of jobs has been proposed, which generalizes the well-known SJF algorithm for improved performance [174, 115] as well as combines it with fairness notions [19, 15]. The queuing order and the timing of reservations can also be determined by economic models [35] or various quality of service assurances [72].

A fourth parameter (related to the previous one) is the partitioning of reservations. The processors of a machine can be partitioned into several disjoint sets (free processors can dynamically move around between them based on current needs). Each set is associated with its own wait-queue and reservation. Lawson and Smirni divided the machine such that different sets serve different jobs classes, characterized by their estimated runtime (e.g. short, medium, and long) [90, 92]. A backfilling candidate is chosen in a round-robin fashion, each time from a different set, and must respect all reservations. By separating short from long jobs, this multiple queue policy reduces the likelihood that a short job is overly delayed in the queue behind a very long job, and therefore improves average performance metrics.

A fifth parameter is the adaptiveness of backfilling. An adaptive backfill scheduler continuously simulates the execution of recently submitted jobs under various scheduling disciplines, compares the hypothetical resulting performance, and periodically switches the scheduling algorithm to be the one that scored the highest. In the face of different workload conditions, this adaptiveness has the effect of both improving and stabilizing the observed performance results [144, 149].

A sixth parameter is the amount of lookahead into the queue. Most backfilling algorithms consider the queued jobs one at a time when trying to backfill them, which often leads to loss of resources to fragmentation. The alternative is to consider the whole queue at once, and try to find the set of jobs that together maximize the utilization while at the same time respecting the allocated reservation(s). This appears to be a NP-hard problem, but due to the fact machine sizes are relatively small, this can be done in polynomial time (in the complexity of the machine size) using dynamic programming, leading to optimal packing [132, 131].

A seventh and final parameter is related to speculative backfilling, where the scheduler is allowed to exploit gaps in the schedule for backfilling, even if the backfilled job

interferes with the reservation. By doing so, the scheduler speculates that the backfilled job would terminate sooner than its estimate suggests, and in any case before the shadow time. Successful speculations obviously improve performance and utilization, and have no negative side effects. But unsuccessful speculations must somehow be dealt with. Unfortunately, all previously suggested solution resulted in a scheduling algorithm that lies outside the attractive variable partitioning domain: The simplest alternative is to kill the offending backfilled job and restart it later on [90]. A similar idea is to employ “short test runs”, during which jobs either manage to terminate, or are reinserted to the wait-queue with a tighter estimate deduced from the test [115, 15]. Unfortunately, both ideas assume jobs are restartable, which is often not the case. A possible workaround is to employ preemption (time sharing), such that instead of killing and restarting the job, it is suspended and kept in memory, only to be resumed later on from the point in which it was stopped [139]. However, if preemption comes into play, it may be preferable to instead combine backfilling with a full fledged gang scheduler altogether [169]. This combination has even been further extended by adding migration capabilities [85, 170]. A recent study suggested preemption is actually redundant if migration or dynamic partitioning are available. The idea is to reduce the backfilled job’s processor allocation by folding it over itself. This frees most of its processors, and limits the performance degradation to the offending job [162].

Finally, a new direction in job scheduling research is to try and minimize the electric power demand, which is rapidly becoming a problem in the context of supercomputing [129]. It has been suggested to integrate the concept of scheduling and power management within EASY [91]. The proposed scheduler continuously monitors load in the system and selectively puts certain nodes in “sleep mode” (makes them unavailable for execution), after estimating the effect of fewer nodes on the projected job slowdown. Using online simulation, the system adaptively selects the minimal number of processors that are required to meet certain negotiated service level agreements.

3.3.3 SLURM

The prime focus of this work will be on **SLURM(Simple Linux Utility For Resource Management)** which will be the choice of batch system upon which the support for Invasive Computing will be demonstrated. SLURM is a sophisticated open source batch system with about 500,000 lines of C code whose development started in the year 2002 at Lawrence Livermore National Laboratory as a simple resource manager for Linux Clusters and a few years ago spawned into an independent firm under the name SchedMD. SLURM has since its inception also evolved into a very capable job scheduler through the use of optional plugins. It is used on many of the world’s largest

supercomputers and is used by a large fraction of the world's TOP500 Supercomputer list. It supports many UNIX flavors like AIX, Linux, Solaris and is also fault tolerant, highly scalable, and portable.

Plugins are dynamically linked objects loaded at run time based upon configuration file and/or user options. 3.1 shows where these plugins fit inside SLURM. Approximately 80 plugins of different varieties are currently available. Some of them are listed below:

- *Accounting storage*: MySQL, PostgreSQL, textfile.
- *Network Topology*: 3D-Torus, tree.
- *MPI*: OpenMPI, MPICH1, MVAPICH, MPICH2, etc.

PLugins are typically loaded when the daemon or command starts and persist indefinitely. They provide a level of indirection to a configurable underlying function.

SLURM Kernel				
Authentication Plugin	MPI Plugin	Checkpoint Plugin	Topology Plugin	Accounting storage Plugin
Munge	mvapich	BLCR	Tree	MySQL

Figure 3.1: SLURM with optional Plugins

3.4 Support for Invasive Computing

3.4.1 Invasive MPI

3.4.2 Resource Management Extensions

Existing batch systems usually only support static allocation of resources to applications before job start. Hence we need some kind of an Invasive Resource Management to be integrated into these existing batch systems so that we can support malleable jobs allowing us to change the allocated resources dynamically at runtime. In order to achieve this we will follow the below approach:

- **Invasive Resource Manager (alias iHypervisor):** An independent component which will talk to the current batch systems via a communication mechanism(protocol) to obtain invasive job(s) submitted specifically to the invasic partition that can support invasive computing. The iHypervisor will then take these jobs and perform some kind of runtime scheduling for pinning these jobs to the resources in the partition and makes these decisions in order to optimize certain local metrics such as resource utilization, power stability, energy efficiency etc. The scheduling here is done at the granularity of cores and sockets. iHypervisor is the one that has the complete information of the resources in the invasic partition and also manages them. This component in an independent entity with the purpose of inter-operating with existing batch systems rather than replacing them with an entirely new one. It may be possible that in the future this component will not be a separate entity but will be built into the batch system itself.
- **Invasive Job Scheduler (alias iScheduler):** This component will be a new extension built into the existing batch systems for performing job scheduling. The scheduling decisions are communicated via the protocol used to speak to iHypervisor and these decisions are basically job(s) selected via a scheduling algorithm to be submitted to the iHypervisor for execution. The scheduling decisions will be made on the basis of available resources in the partition and it is the iHypervisor that communicates this to iScheduler in the form of resource offers (Real/Virtual). It can be a virtual resource offer because the iHypervisor can hide the real resources and present a rather fake view of them to iScheduler in the hope of getting a mapping of jobs to offer that is more suitable to satisfy its local metrics. Similar to iHypervisor, the iScheduler makes its decisions to optimize for certain local metrics such as high job throughput, reduced job waiting times, deadlines, priorities etc. This highlights the mismatching policies/metrics for which both the iHypervisor and iScheduler make their decisions on and hence both will be involved in some kind of a negotiation via the protocol to reach a common agreement.
- **Negotiation Protocol:** This protocol forms the core of the interaction between the iScheduler and iHypervisor. It allows for iHypervisor to make one or a set of resource offers to iScheduler which then needs to select jobs from its job queue to be mapped to these resource offers and finally sent back to the iHypervisor. The iHypervisor will then decide whether to accept/reject this mapping to satisfy its local metrics. If it accepts it will launch them based on some run time scheduling and if it rejects then it informs this to iScheduler in addition to sending it a new resource offer. The iScheduler can also reject the resource offers in which case it will be sent a new one. On accepting an offer, the iScheduler then repeats its tasks

to send back a mapping to iHypervisor and this interaction continues until both reach a common agreement. If the number of such attempts reach a threshold then iScheduler will just accept whatever offer it receives and iHypervisor will also accept whatever Map:Jobs→Offers it receives closing this transaction of negotiating. After this a new transaction will start.

- **SLURM:** SLURM is our choice of an existing batch system upon which this new implementation will be demonstrated as a proof of concept to support the new paradigm of resource-aware programming in the domain of invasive computing. In the near future, This can motivate further such supporting infrastructures with other batch systems using such Invasive Resource Management and Scheduling components.

This project implements a testing prototype for demonstrating how such an approach to support invasive computing with the above entities may work. It will involve implementing the communication infrastructure using SLURM API due to which iHypervisor and iScheduler will interact with each other using protocol messages filled with dummy values. It will also involve implementing the iScheduler as a new plugin(multithreaded) for SLURM and iHypervisor as a fake multithreaded Invasive Resource Manager daemon with bare minimum functionalities that in the near future will be an enhanced version of the daemon slurmd(built on top of it) found in SLURM. For the purposes of testing this prototype, the different scenarios that can be observed most of the time with the kind of negotiation protocol described earlier would be verified.

4 Dynamic Resource Management Architecture

4.1 System Design

This section illustrates and describes a high level design of the software implemented with the help of protocol sequence diagrams and state machine diagrams. It will help to understand at a high level as to how the system has been designed to support this new approach of invasive computing and how will many of its components in the software hierarchy interact with each other with new protocols or extensions of existing protocols to integrate such an invasive resource management into current batch systems.

The following page shows the software architecture of how Invasive Resource Management can be supported with a traditional resource manager and how exactly the new software components will fit in the existing software hierarchy. The 4.1 relates closely to how SLURM is organized since the intention of this work would be to demonstrate the support for Invasive Computing with the help of SLURM as a resource manager.

- The top layer is that of the core resource management component which has access to job queues. In this architecture, it will now have access to not only the queue for the legacy(static) jobs but also invasive job queue(jobs submitted to invasive partition that supports invasive computing).
- In a traditional setup the top layer will perform the task of job scheduling as well. This means that it will select a job(s) from the queue of jobs based on the current state of resources and many other factors to dispatch it to the traditional process manager below in the hierarchy. The process manager then takes the responsibility of launching these jobs on the allocated resources in the partition and managing them for their full lifetime. In case of parallel jobs, it will manage the job in a parallel environment along with facilitating the communication amongst the parallel tasks/processes with the help of a PMI(Process Manager Interface) server. The process manager may also spawn slave daemons on each of the nodes which are a part of the resource allocation for a single job to manage them more effectively.

- As discussed in the previous chapter, an independent Invasive resource management component by the name "iHypervisor" will be implemented which needs to communicate with a new scheduling component iScheduler and influence the scheduling decisions taken by it. The iHypervisor sits between the top layer and the process manager.
- A new job scheduler specifically for invasive jobs needs to be integrated into the existing batch system. This is due to the reason that the scheduler for invasive jobs will work in a different manner based on the approach described earlier in comparison to the legacy job scheduler for static jobs. In case of SLURM which has a modular design with several optional plugins, a new plugin by name "iScheduler" will be implemented for SLURM to handle job scheduling specifically for invasive jobs.
- Communication between iHypervisor and iScheduler will involve the negotiation protocol as explained in the previous chapter but will also include periodic feedbacks being sent by iHypervisor to iScheduler having some useful statistical measures about current state of resources, resource utilization, job throughput etc. that may help influence the decision making of iScheduler. This communication will also additionally support a means to service urgent jobs immediately.

Communication Phases

- **Protocol Initialization:** This phase basically establishes the initial environment between the communicating parties (iScheduler and iHypervisor) for proper communication later on. Successful initialization of this phase prepares both the parties to start negotiating based on the negotiation protocol described in the following points. During this protocol initialization various parameters such as protocol version, maximum attempts for negotiation, timer intervals and several others could be exchanged to set up the internal data structures and configuration tables for both the communicating parties. This protocol is a bi-directional communication.
- **Protocol Finalization:** This phase signals the end of communication between iHypervisor and iScheduler using negotiation protocol. It leads to a safe termination of this communication followed by the release of any internal data structures allocated earlier along with configuration parameters. This results in consistent behaviour of both the communicating parties which can then proceed to safely terminate and exit. This protocol is a bi-directional communication.
- **Negotiation:** This is the most important phase in this whole approach to support invasive computing as discussed in the previous chapter. It is the phase during

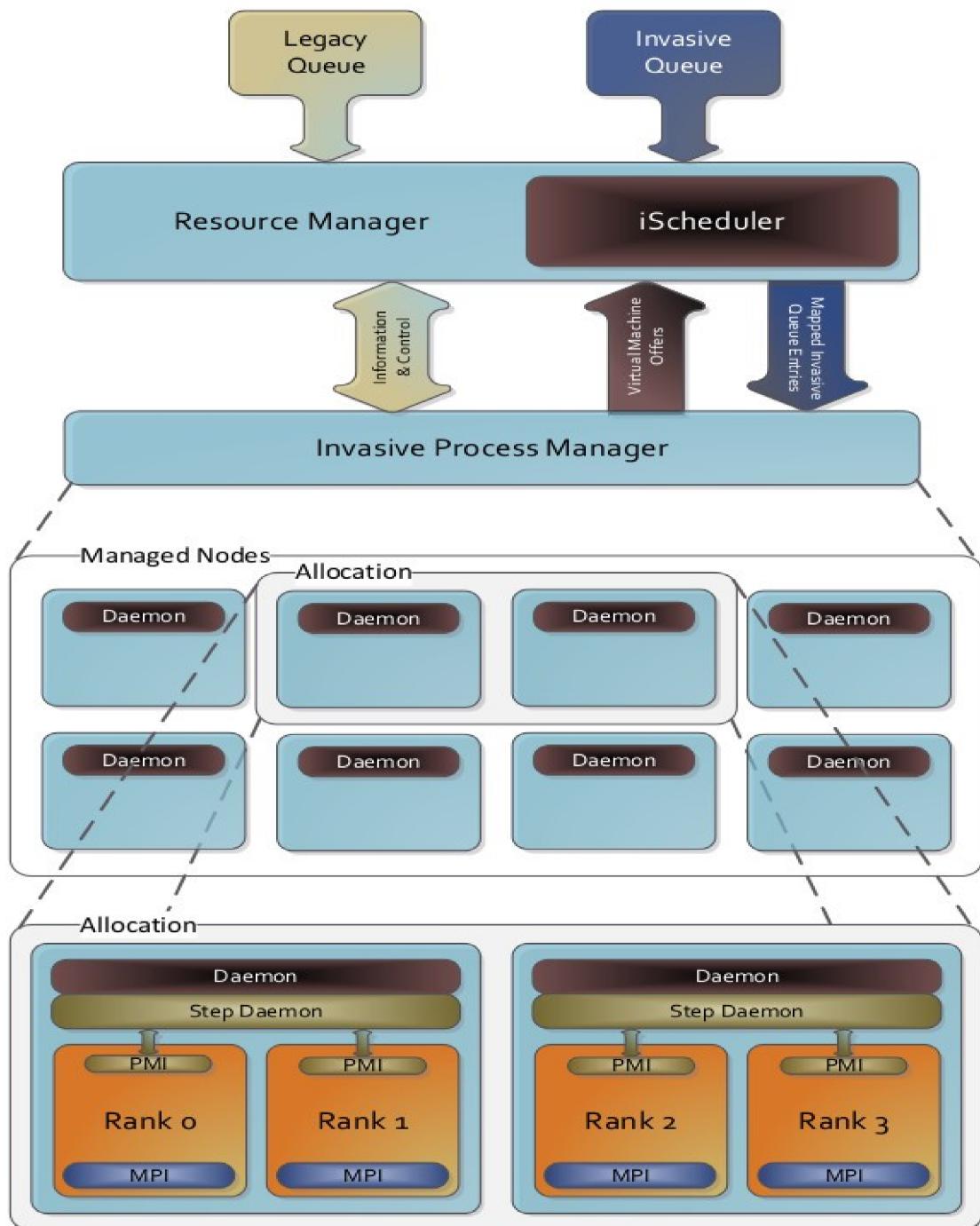


Figure 4.1: Invasive Resource Management Architecture

<EVENT> <=> <PACKET>	DIRECTION OF COMMUNICATION
<REQUEST_RESOURCE_OFFER>	iScheduler → iHypervisor
<RESOURCE_OFFER>	iHypervisor → iScheduler
<RESPONSE_RESOURCE_OFFER>	iScheduler → iHypervisor
<NEGOTIATION_START>	iScheduler → iHypervisor
<RESPONSE_NEGOTIATION_START>	iHypervisor → iScheduler
<NEGOTIATION_END>	iScheduler → iHypervisor iHypervisor → iScheduler
<RESPONSE_NEGOTIATION_END>	iScheduler → iHypervisor iHypervisor → iScheduler
<STATUS_REPORT>	iHypervisor → iScheduler
<URGENT_JOB>	iScheduler → iHypervisor
<RESPONSE_URGENT_JOB>	iHypervisor → iScheduler

Figure 4.2: Message Types

which both iHypervisor and iScheduler are negotiating with each other till they reach an agreement. If they do not then they continue till a certain limit to the number of negotiating attempts are reached after which both of them just agree in their final attempt closing the current negotiation. After this a new transaction of negotiation begins.

- **Feedback:** This concerns the periodic feedback sent by the iHypervisor to the iScheduler containing useful information such as the job states, latest snapshot of the resources in the invasic partition and many other statistical measures not limited to system utilization, job throughput, waiting times of jobs to help and influence the iScheduler in its decision making for scheduling jobs during its future transactions of negotiation. This protocol is a uni-directional communication.
- **Urgent Jobs:** This protocol concerns the support for urgent jobs. At any given point of time a cluster or supercomputing center may want to support very high priority jobs immediately without any further delay. By introducing support for invasive computing, it makes it all the more feasible to help run these urgent jobs immediately by either shrinking the resources of other jobs or suspending/Killing them.

Separation of Concerns: In this thesis, The idea of separating the batch and runtime

scheduling components of a Job Scheduler is explored.

4.1.1 Batch Scheduler

Today almost all resource management systems fall into the category of queuing systems. Several queues with different limits on the number of requested resources and the duration exist for the submission of resource requests. Jobs within a queue are ordered according to a scheduling policy, e. g. FCFS (first come, first serve). Queues might be activated only for specific times (e. g. prime time, non prime time, or weekend). The task of a queuing system is to assign free resources to waiting requests. The highest prioritized request is always the queue head. If it is possible to start more than one queue head, further criteria like queue priority or best fit (e. g. leaving less resources idle) are used to select a request. There might also exist a high priority queue whose jobs are preferred at any time. If not enough resources are available to start any of the queue heads, the system waits until enough resources become available. These idle resources may be utilized with less prioritized requests by backfilling mechanisms.

4.1.2 Distributed Run Time Scheduler

4.1.3 MPI Process Manager

4.2 Negotiation Protocol

4.2.1 State Machine Diagrams

This section focuses on iScheduler and a thread iRM_AGENT that it spawns which is the one responsible for all the communication with the iHypervisor including spawning other agent threads for handling feedbacks and urgent jobs.

- Above diagram and the ones in the following pages illustrate state machine diagrams for few of the communication phases described earlier starting first with a general diagram of how the multithreaded component iRM agent inside iScheduler starts up and shuts down.

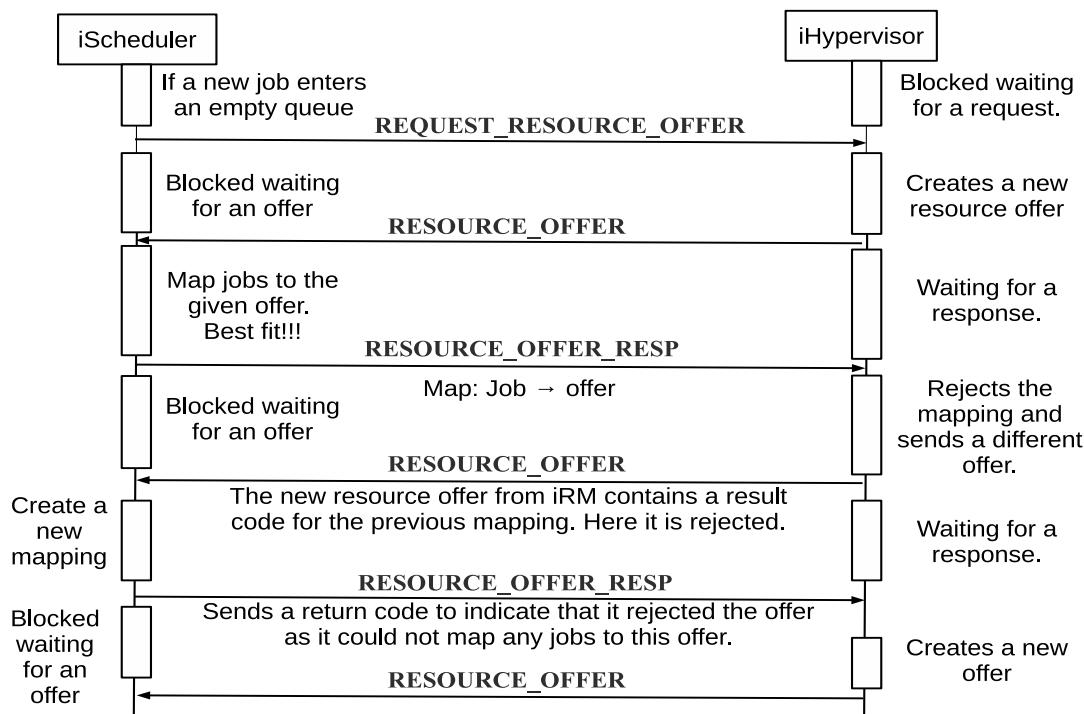


Figure 4.3: Scenario 1

- Above diagram illustrates a scenario where both iScheduler and iHypervisor are negotiating with each other. The scenario is continued in the next page. 4.5 illustrates another scenario where negotiations may stop when job queue becomes empty and iHypervisor then will wait for a request from iScheduler for a resource offer that will happen when new jobs arrive.
 - iScheduler makes scheduling decisions at a coarser level of granularity which is nodes whereas iHypervisor does at the granularity of cores and sockets. Both will negotiate with each other till they reach an agreement.
 - It is an event based scheduling which means iScheduler makes a scheduling decision only when it is triggered by receiving a resource offer from iHypervisor. It is only at the start when there are no jobs in the queue and during the operations when the queue may become empty that the iScheduler will have to explicitly send a request message to iHypervisor for a resource offer otherwise at all other times scheduling is event based.

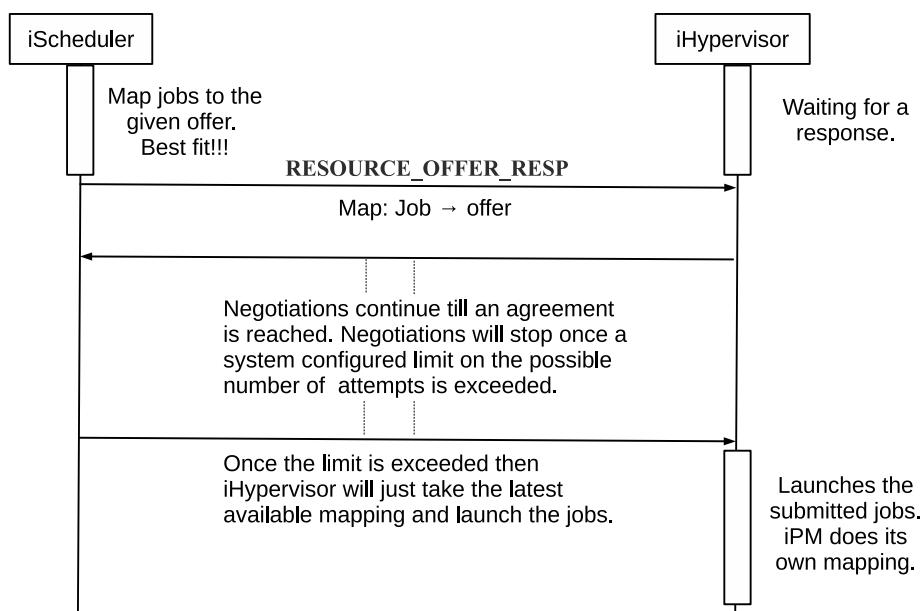


Figure 4.4: Scenario 1 contd.

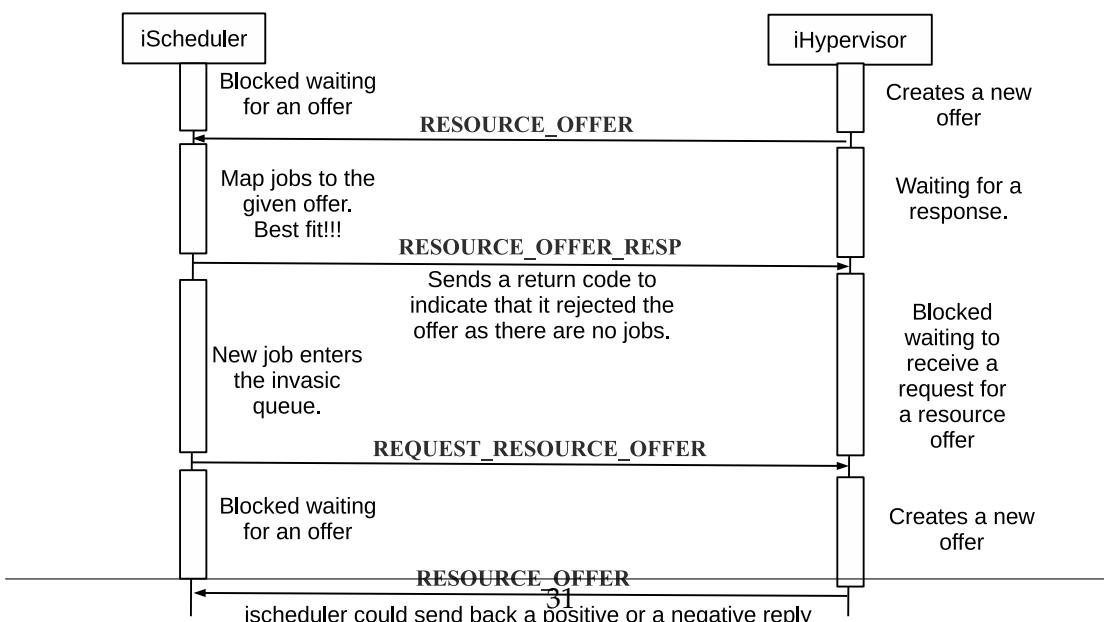


Figure 4.5: Scenario 2

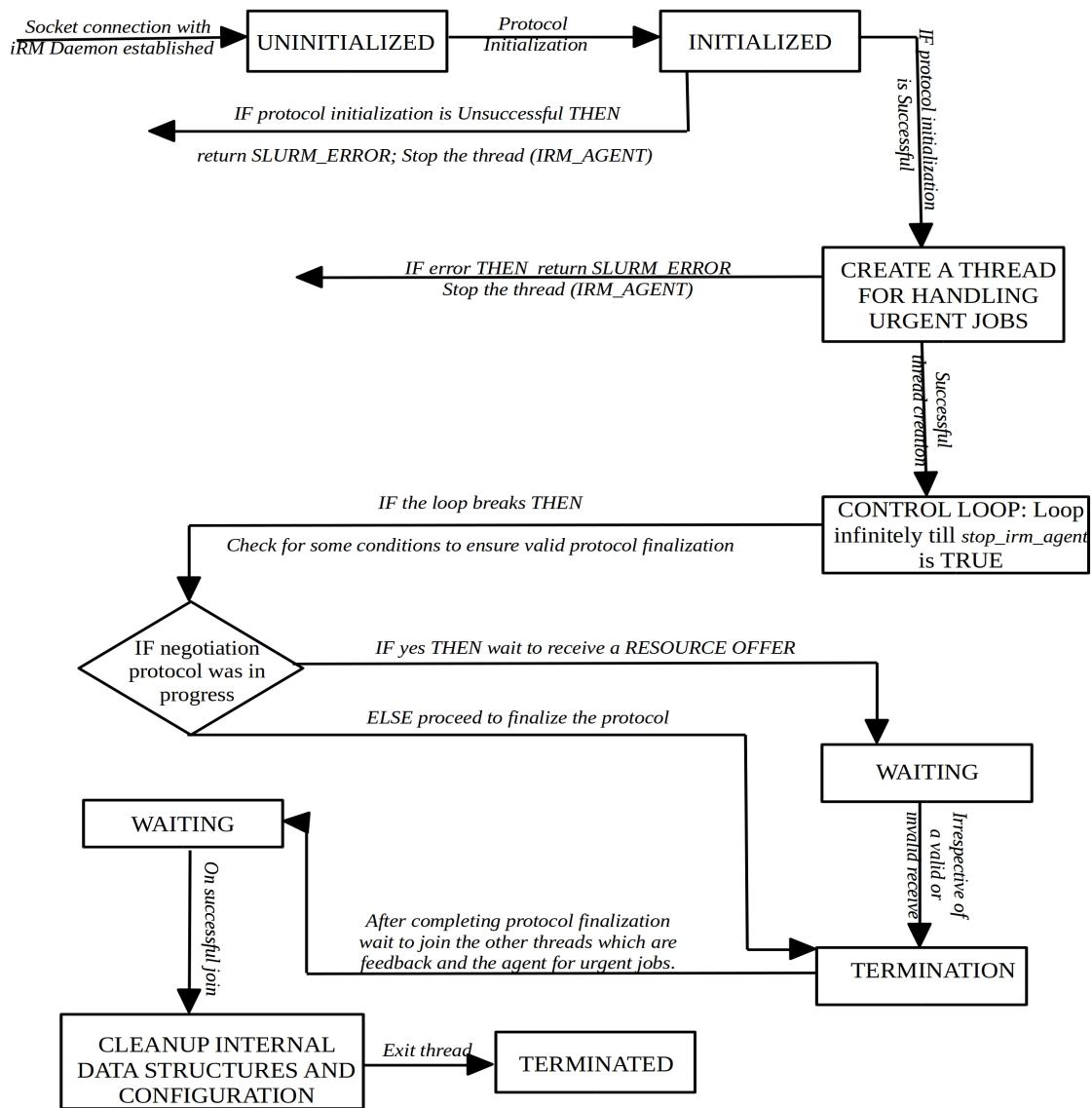


Figure 4.6: iRM Agent

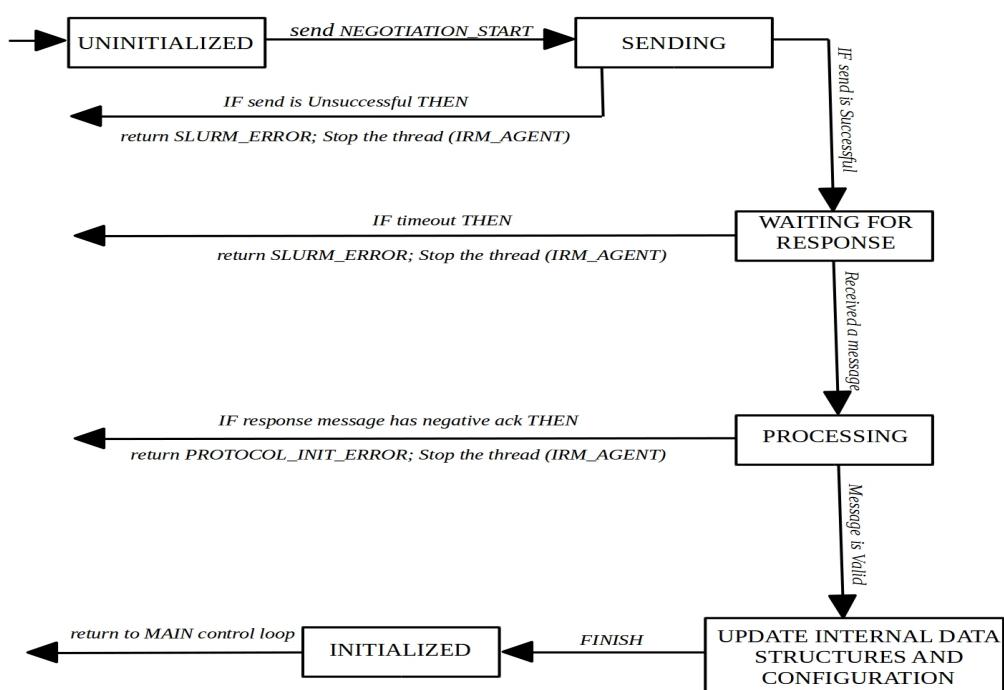


Figure 4.7: Protocol Initialization

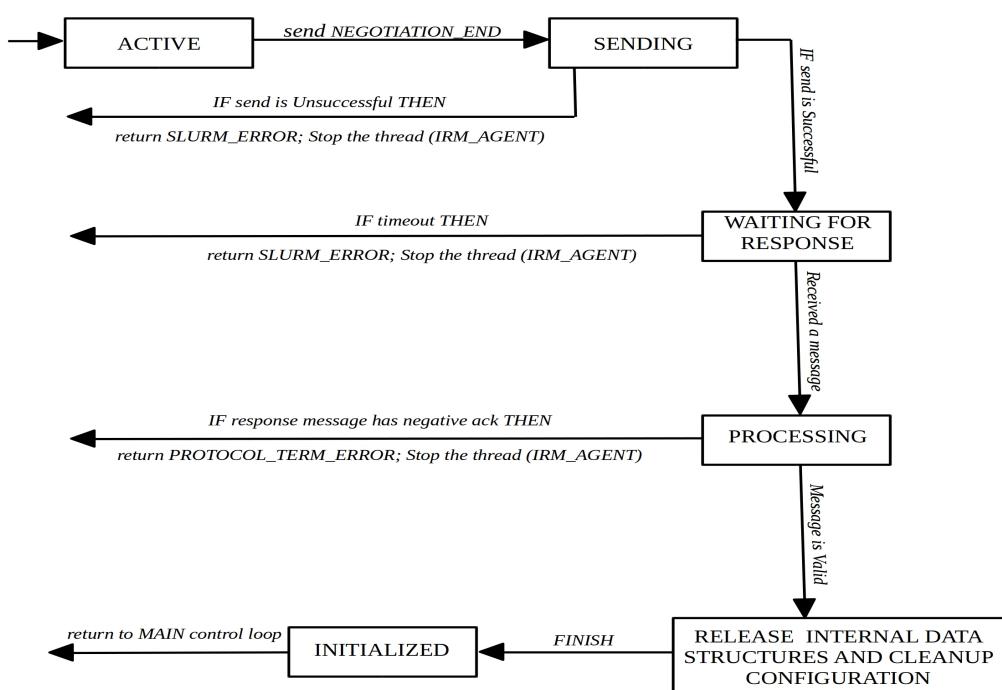
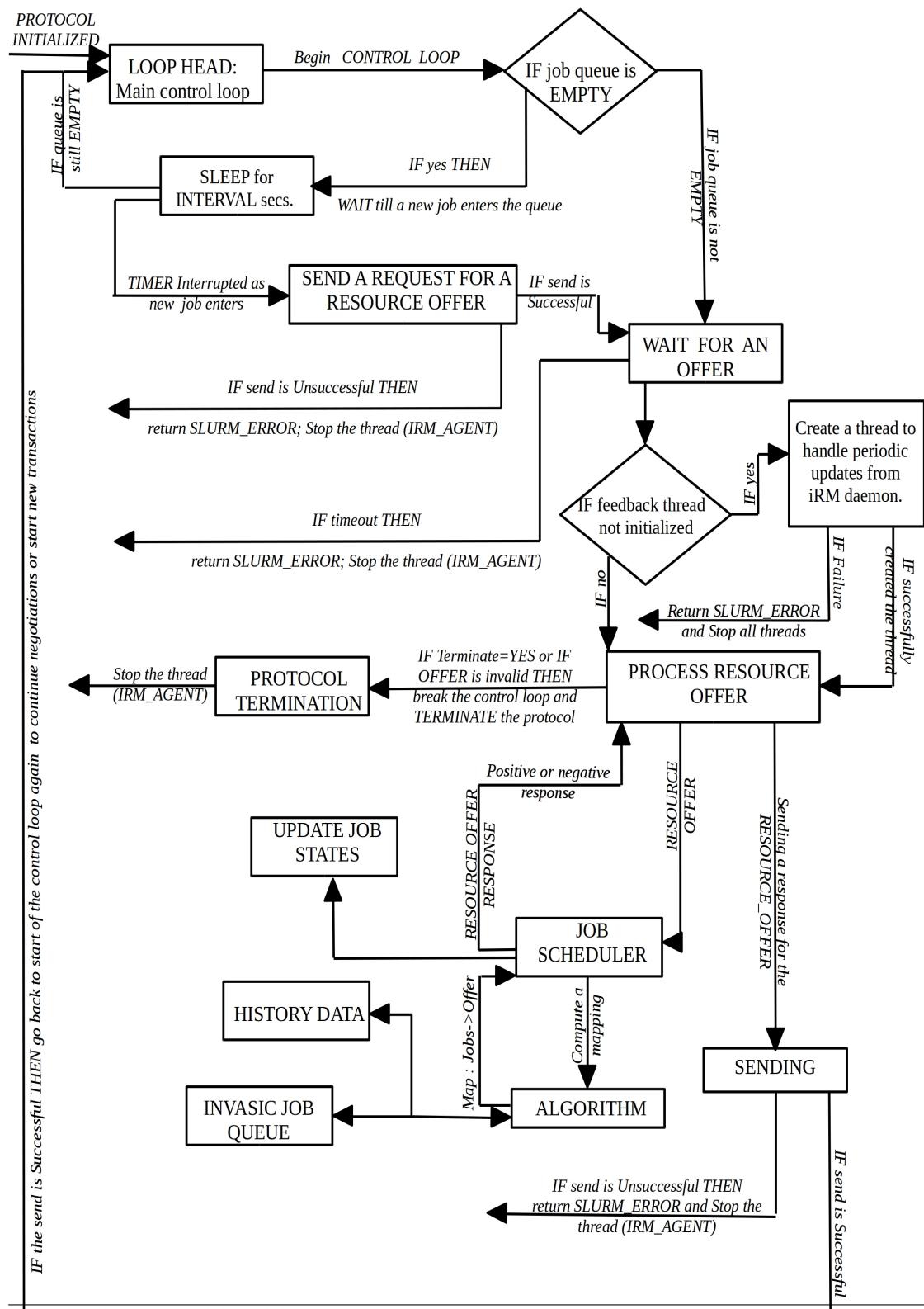


Figure 4.8: Protocol Termination



4.3 Invasive Jobs

5 iScheduler Design

5.1 Job Description

5.2 Resource Offers

5.3 Feedback Reports

5.4 Negotiation Protocol

5.5 Job Scheduling Algorithm

5.5.1 Problem Formulation

5.5.2 Pseudo Code

6 Implementation

6.1 Plugins for SLURM

6.2 Data Structures

6.3 State Machine Diagrams

6.3.1 iScheduler

6.3.2 iHypervisor

6.4 Important APIs

7 Evaluation

7.1 Method of Evaluation

7.1.1 Emulation of Workload

7.1.2 Real Invasive Applications

7.2 Setup

7.3 Experiments and Results

7.4 Performance and Graphs

8 Conclusion and Future Work

8.1 Future Work

[Pra+14] [Pra+15] [Ioa+11] [DL96] [Ure+12] [Ger+12] [Cer+10] [Mag+08] [UCA04] [KP01] [Bal+10] [YJG03] [Gup+14] [Lif95] [Sko+96] [Hun04] [Cao+10] [Zho+13] [Luc11] [TLD13] [Zho+15] [Tan+10] [Tan+05] [FW98] [FW12] [Sch]

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