

Reliability and Performance Models for Grid Computing

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Abstract

Grid computing is a newly developed technology for complex systems with large-scale resource sharing, wide-area communication, and multi-institutional collaboration. It is hard to analyze and model the Grid reliability because of its largeness, complexity and stiffness. Therefore, this chapter introduces the Grid computing technology, presents different types of failures in grid system, models the grid reliability with star structure and tree structure, and finally studies optimization problems for grid task partitioning and allocation. The chapter then presents models for star-topology considering data dependence and tree-structure considering failure correlation. Evaluation tools and algorithms are developed, evolved from Universal generating function and Graph Theory. Then, the failure correlation and data dependence are considered in the model. Numerical examples are illustrated to show the modeling and analysis.

Keywords: Reliability, Performance, Grid computing, Modeling, Graph theory, Bayesian approach.

1. Introduction

Grid computing (Foster & Kesselman, 2003) is a newly developed technology for complex systems with large-scale resource sharing, wide-area communication, and multi-institutional collaboration etc, see e.g. Kumar (2000), Das *et al.* (2001), Foster *et al.* (2001, 2002) and Berman *et al.* (2003). Many experts believe that the grid technologies will offer a second chance to fulfill the promises of the Internet.

The real and specific problem that underlies the Grid concept is coordinated resource sharing and problem solving in dynamic, multi-institutional virtual organizations (Foster *et al.*, 2001). The sharing that we are concerned with is not primarily file exchange but rather direct access to computers, software, data, and other resources. This is required by a range of collaborative problem-solving and resource-brokering strategies emerging in industry, science, and engineering. This sharing is highly controlled by the resource management system (Livny & Raman, 1998), with resource providers and consumers defining what is shared, who is allowed to share, and the conditions under which the sharing occurs.

Recently, the Open Grid Service Architecture (Foster *et al.*, 2002) enables the integration of services and resources across distributed, heterogeneous, dynamic, virtual organizations. A grid service is desired to complete a set of programs under the circumstances of grid computing. The programs may require using remote resources that are distributed. However, the programs initially do not know the site information of those remote resources in such a large-scale computing environment, so the resource management system (the brain of the grid) plays an important role in managing the pool of shared resources, in matching the programs to their requested resources, and in controlling them to reach and use the resources through wide-area network.

The structure and functions of the resource management system (RMS) in the grid have been introduced in details by Livny & Raman (1998), Cao *et al.* (2002), Krauter *et al.* (2002) and Nabrzyski *et al.* (2003). Briefly stated, the programs in a grid service send their requests for resources to the RMS. The RMS adds these requests into the request queue (Livny & Raman, 1998). Then, the requests are waiting in the queue for the matching service of the RMS for a period of time (called waiting time), see e.g. Abramson *et al.* (2002). In the matching service, the RMS matches the requests to the shared resources in the grid (Ding *et*

al., 2002) and then builds the connection between the programs and their required resources. Thereafter, the programs can obtain access to the remote resources and exchange information with them through the channels. The grid security mechanism then operates to control the resource access through the Certification, Authorization and Authentication, which constitute various logical connections that causes dynamicity in the network topology.

Although the developmental tools and infrastructures for the grid have been widely studied (Foster & Kesselman, 2003), grid reliability analysis and evaluation are not easy because of its complexity, largeness and stiffness. The grid computing contains different types of failures that can make a service unreliable, such as blocking failures, time-out failures, matching failures, network failures, program failures and resource failures. This chapter thoroughly analyzes these failures.

Usually the grid performance measure is defined as the task execution time (service time). This index can be significantly improved by using the RMS that divides a task into a set of subtasks which can be executed in parallel by multiple online resources. Many complicated and time-consuming tasks that could not be implemented before are working well under the grid environment now.

It is observed in many grid projects that the service time experienced by the users is a random variable. Finding the distribution of this variable is important for evaluating the grid performance and improving the RMS functioning. The service time is affected by many factors. First, various available resources usually have different task processing speeds online. Thus, the task execution time can vary depending on which resource is assigned to execute the task/subtasks. Second, some resources can fail when running the subtasks, so the execution time is also affected by the resource reliability. Similarly, the communication links in grid service can be disconnected during the data transmission. Thus, the communication reliability influences the service time as well as data transmission speed through the communication channels. Moreover, the service requested by a user may be delayed due to the queue of earlier requests submitted from others. Finally, the data dependence imposes constraints on the sequence of the subtasks' execution, which has significant influence on the service time.

This chapter first introduces the grid computing system and service, and analyzes various

failures in grid system. Both reliability and performance are analyzed in accordance with the performability concept. Then the chapter presents models for star- and tree-topology grids respectively. The reliability and performance evaluation tools and algorithms are developed based on the universal generating function, graph theory, and Bayesian approach. Both failure correlation and data dependence are considered in the models.

2. Grid Service Reliability and Performance

2.1. Description of the grid computing

Today, the **Grid computing** systems are large and complex, such as the IP-Grid (Indiana-Purdue Grid) that is a statewide grid (<http://www.ip-grid.org/>). IP-Grid is also a part of the TeraGrid that is a nationwide grid in the USA (<http://www.teragrid.org/>). The largeness and complexity of the grid challenge the existing models and tools to analyze, evaluate, predict and optimize the **reliability** and **performance** of grid systems. The global grid system is generally depicted by the Fig. 1. Various organizations (Foster *et al.*, 2001), integrate/share their resources on the global grid. Any program running on the grid can use those resources if it can be successfully connected to them and is authorized to access them. The sites that contain the resources or run the programs are linked by the global network as shown in the left part of Fig. 1.

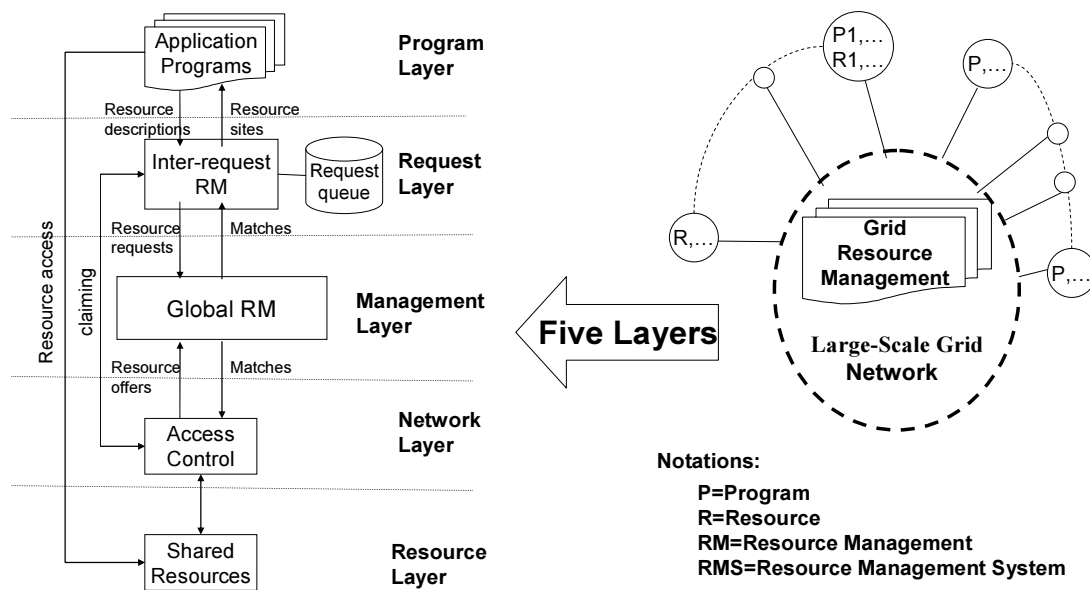


Fig. 1. Grid Computing System

The distribution of the service tasks/subtasks among the remote resources are controlled by the Resource Management System (RMS) that is the “brain” of the grid computing, see e.g. Livny & Raman (1998). The RMS has five layers in general, as shown in Fig. 1: program layer, request layer, management layer, network layer and resource layer.

- 1) *Program layer*: The program layer represents the programs of the customer’s applications. The programs describe their required resources and constraint requirements (such as deadline, budget, function etc). These resource descriptions are translated to the resource requests and sent to the next request layer.
- 2) *Request layer*: The request layer provides the abstraction of “program requirements” as a queue of resource requests. The primary goals of this layer are to maintain this queue in a persistent and fault-tolerant manner and to interact with the next management layer by injecting resource requests for matching, claiming matched resources of the requests.
- 3) *Management layer*: The management layer may be thought of as the global resource allocation layer. It has the function of automatically detecting new resources, monitoring the resource pool, removing failed/unavailable resources, and most importantly matching the resource requests of a service to the registered/detected resources. If resource requests are matched with the registered resources in the grid, this layer sends the matched tags to the next network layer.

- 4) *Network layer*: The network layer dynamically builds connection between the programs and resources when receiving the matched tags and controls them to exchange information through communication channels in a secure way.
- 5) *Resource layer*: The resource layer represents the shared resources from different resource providers including the usage policies (such as service charge, reliability, serving time etc.)

2.2. Failure analysis of grid service

Even though all online nodes or resources are linked through the Internet with one another, not all resources or communication channels are actually used for a specific service. Therefore, according to this observation, we can make tractable models and analyses of grid computing via a virtual structure for a certain service. The grid service is defined as follows:

Grid service is a service offered under the grid computing environment, which can be requested by different users through the RMS, which includes a set of subtasks that are allocated to specific resources via the RMS for execution, and which returns the result to the user after the RMS integrates the outputs from different subtasks.

The above five layers coordinate together to achieve a grid service. At the “Program layer”, the subtasks (programs) composing the entire grid service task initially send their requests for remote resources to the RMS. The “Request layer” adds these requests in the request queue. Then, the “Management layer” tries to find the sites of the resources that match the requests. After all the requests of those programs in the grid service are matched, the “Network layer” builds the connections among those programs and the matched resources.

It is possible to identify various types of failures on respective layers:

- *Program layer*: Software failures can occur during the subtask (program) execution; see e.g. Xie (1991) and Pham (2000).
- *Request layer*: When the programs’ requests reach the request layer, two types of failures may occur: “blocking failure” and “time-out failure”. Usually, the request queue has a limitation on the maximal number of waiting requests (Livny & Raman, 1998). If the queue is full when a new request arrives, the request blocking failure occurs. The grid service usually has its due time set by customers or service

monitors. If the waiting time for the requests in the queue exceeds the due time, the time-out failure occurs, see e.g. Abramson *et al.* (2002).

- *Management layer*: At this layer, “matching failure” may occur if the requests fail to match with the correct resources, see e.g. Xie *et al.* (2004, pp. 185-186). Errors, such as incorrectly translating the requests, registering a wrong resource, ignoring resource disconnection, misunderstanding the users' requirements, can cause these matching failures.
- *Network layer*: When the subtasks (programs) are executed on remote resources, the communication channels may be disconnected either physically or logically, which causes the “network failure”, especially for those long time transmissions of large dataset, see e.g. Dai *et al.* (2002).
- *Resource layer*: The resources shared on the grid can be of software, hardware or firmware type. The corresponding software, hardware or combined faults can cause resource unavailability.

2.3. Grid Service Reliability and Performance

Most previous research on distributed computing studied performance and reliability separately. However, performance and reliability are closely related and affect each other, in particular under the grid computing environment. For example, while a task is fully parallelized into m subtasks executed by m resources, the performance is high but the reliability might be low because the failure of any resource prevents the entire task from completion. This causes the RMS to restart the task, which reversely increases its execution time (i.e. reduces performance). Therefore, it is worth to assign some subtasks to several resources to provide execution redundancy. However, excessive redundancy, even though improving the reliability, can decrease the performance by not fully parallelizing the task. Thus, the performance and reliability affect each other and should be considered together in the grid service modeling and analysis.

In order to study performance and reliability interactions, one also has to take into account the effect of service performance (execution time) upon the reliability of the grid elements. The conventional models, e.g. Kumar *et al.* (1986), Chen & Huang (1992), Chen *et al.* (1997), and Lin *et al.*, (2001), are based on the assumption that the operational probabilities of nodes or links are constant, which ignores the links' bandwidth,

communication time and resource processing time. Such models are not suitable for precisely modeling the grid service performance and reliability.

Another important issue that has much influence the performance and reliability is data dependence, that exists when some subtasks use the results from some other subtasks. The service performance and reliability is affected by data dependence because the subtasks cannot be executed totally in parallel. For instance, the resources that are idle in waiting for the input to run the assigned subtasks are usually hot-standby because cold-start is time consuming. As a result, these resources can fail in waiting mode.

The considerations presented above lead the following assumptions that lay in the base of grid service reliability and performance model.

Assumptions:

- 1) The service request reaches the RMS and is being served immediately. The RMS divides the entire service task into a set of subtasks. The data dependence may exist among the subtasks. The order is determined by precedence constraints and is controlled by the RMS.
- 2) Different grid resources are registered or automatically detected by the RMS. In a grid service, the structure of virtual network (consisting of the RMS and resources involved in performing the service) can form star topology with the RMS in the center or, tree topology with the RMS in the root node.
- 3) The resources are specialized. Each resource can process one or multiple subtask(s) when it is available.
- 4) Each resource has a given constant processing speed when it is available and has a given constant failure rate. Each communication channel has constant failure rate and a constant bandwidth (data transmission speed).
- 5) The failure rates of the communication channels or resources are the same when they are idle or loaded (hot standby model). The failures of different resources and communication links are independent.
- 6) If the failure of a resource or a communication channel occurs before the end of output data transmission from the resource to the RMS, the subtask fails.

- 7) Different resources start performing their tasks immediately after they get the input data from the RMS through communication channels. If same subtask is processed by several resources (providing execution redundancy), it is completed when the first result is returned to the RMS. The entire task is completed when all of the subtasks are completed and their results are returned to the RMS from the resources.
- 8) The data transmission speed in any multi-channel link does not depend on the number of different packages (corresponding to different subtasks) sent in parallel. The data transmission time of each package depends on the amount of data in the package. If the data package is transmitted through several communication links, the link with the lowest bandwidth limits the data transmission speed.
- 9) The RMS is fully reliable, which can be justified to consider a relatively short interval of running a specific service. The imperfect RMS can also be easily included as a module connected in series to the whole grid service system.

2.4. Grid Service time distribution and reliability/performance measures

The data dependence on task execution can be represented by $m \times m$ matrix H such that $h_{ki} = 1$ if subtask i needs for its execution output data from subtask k and $h_{ki} = 0$ otherwise (the subtasks can always be numbered such that $k < i$ for any $h_{ki} = 1$). Therefore, if $h_{ki} = 1$ execution of subtask i cannot begin before completion of subtask k . For any subtask i one can define a set ϕ_i of its immediate predecessors: $k \in \phi_i$ if $h_{ki} = 1$.

The data dependence can always be presented in such a manner that the last subtask m corresponds to final task processed by the RMS when it receives output data of all the subtasks completed by the grid resources.

The task execution time is defined as time from the beginning of input data transmission from the RMS to a resource to the end of output data transmission from the resource to the RMS.

The amount of data that should be transmitted between the RMS and resource j that executes subtask i is denoted by a_i . If data transmission between the RMS and the resource j is accomplished through links belonging to a set γ_j , the data transmission speed is

$$s_j = \min_{L_x \in \gamma_j} (b_x) \quad (1)$$

where b_x is the bandwidth of the link L_x . Therefore, the random time t_{ij} of subtask i execution by resource j can take two possible values

$$t_{ij} = \hat{t}_{ij} = \tau_j + \frac{a_i}{s_j} \quad (2)$$

if the resource j and the communication path γ_j do not fail until the subtask completion and $t_{ij} = \infty$ otherwise. Here, τ_j is the processing time of the j -th resource.

Subtask i can be successfully completed by resource j if this resource and communication path γ_j do not fail before the end of subtask execution. Given constant failure rates of resource j and links, one can obtain the conditional probability of subtask success as

$$p_j(\hat{t}_{ij}) = e^{-(\lambda_j + \pi_j)\hat{t}_{ij}} \quad (3)$$

where π_j is the failure rate of the communication path between the RMS and the resource j ,

which can be calculated as $\pi_j = \sum_{x \in \gamma_j} \lambda_x$, λ_x is the failure rate of the link L_x . The

exponential distribution (3) is common in software or hardware components' reliability that had been justified in both theory and practice, see e.g. Xie *et al.* (2004).

These give the conditional distribution of the random subtask execution time t_{ij} :

$$\Pr(t_{ij} = \hat{t}_{ij}) = p_j(\hat{t}_{ij}) \text{ and } \Pr(t_{ij} = \infty) = 1 - p_j(\hat{t}_{ij}).$$

Assume that each subtask i is assigned by the RMS to resources composing set ω_i . The RMS can initiate execution of any subtask j (send the data to all the resources from ω_i) only

after the completion of every subtask $k \in \phi_i$. Therefore the random time of the start of subtask i execution T_i can be determined as

$$T_i = \max_{k \in \phi_i} (\tilde{T}_k) \quad (4)$$

where \tilde{T}_k is random completion time for subtask k . If $\phi_i = \emptyset$, i.e. subtask i does not need data produced by any other subtask, the subtask execution starts without delay: $T_i = 0$. If $\phi_i \neq \emptyset$, T_i can have different realizations \hat{T}_{il} ($1 \leq l \leq N_i$).

Having the time T_i when the execution of subtask i starts and the time t_{ij} of subtask i executed by resource j , one obtains the completion time for subtask i on resource j as

$$\tilde{t}_{ij} = T_i + t_{ij}. \quad (5)$$

In order to obtain the distribution of random time \tilde{t}_{ij} one has to take into account that probability of any realization of $\tilde{t}_{ij} = \hat{T}_{il} + \hat{t}_{ij}$ is equal to the product of probabilities of three events:

- execution of subtask i starts at time \hat{T}_{il} : $q_{il} = \Pr(T_i = \hat{T}_{il})$;
- resource j does not fail before start of execution of subtask i : $p_j(\hat{T}_{il})$;
- resource j does not fail during the execution of subtask i : $p_j(\hat{t}_{ij})$.

Therefore, the conditional distribution of the random time \tilde{t}_{ij} given execution of subtask i starts at time \hat{T}_{il} ($T_i = \hat{T}_{il}$) takes the form

$$\Pr(\tilde{t}_{ij} = \hat{T}_{il} + \hat{t}_{ij}) = p_j(\hat{T}_{il}) p_j(\hat{t}_{ij}) = p_j(\hat{T}_{il} + \hat{t}_{ij}) = e^{-(\lambda_j + \pi_j)(\hat{T}_{il} + \hat{t}_{ij})}, \quad (6)$$

$$\Pr(\tilde{t}_{ij} = \infty) = 1 - p_j(\hat{T}_{il} + \hat{t}_{ij}) = 1 - e^{-(\lambda_j + \pi_j)(\hat{T}_{il} + \hat{t}_{ij})}.$$

The random time of subtask i completion \tilde{T}_i is equal to the shortest time when one of the resources from ω_i completes the subtask execution:

$$\tilde{T}_i = \min_{j \in \omega_i} (\tilde{t}_{ij}). \quad (7)$$

According to the definition of the last subtask m , the time of its beginning corresponds to the service completion time, because the time of the task proceeds with RMS is neglected. Thus, the random service time Θ is equal to T_m . Having the distribution (pmf) of the random value $\Theta \equiv T_m$ in the form $q_{ml} = \Pr(T_m = \hat{T}_{ml})$ for $1 \leq l \leq N_m$, one can evaluate the reliability and performance indices of the grid service.

In order to estimate both the service reliability and its performance, different measures can be used depending on the application. In applications where the execution time of each task (service time) is of critical importance, the system reliability $R(\Theta^*)$ is defined (according to performability concept in Meyer (1980), Grassi *et al.* (1988) and Tai *et al.* (1993)) as a probability that the correct output is produced in time less than Θ^* . This index can be obtained as

$$R(\Theta^*) = \sum_{l=1}^{N_m} q_{ml} \cdot 1(\hat{T}_{ml} < \Theta^*). \quad (8)$$

When no limitations are imposed on the service time, the service reliability is defined as the probability that it produces correct outputs without respect to the service time, which can be referred to as $R(\infty)$. The conditional expected service time W is considered to be a measure of its performance, which determines the expected service time given that the service does not fail, i.e.

$$W = \sum_{l=1}^{N_m} \hat{T}_{ml} q_{ml} / R(\infty). \quad (9)$$

3. **Star Topology** Grid Architecture

A grid service is desired to execute a certain task under the control of the RMS. When the RMS receives a service request from a user, the task can be divided into a set of subtasks that

are executed in parallel. The RMS assigns those subtasks to available resources for execution. After the resources complete the assigned subtasks, they return the results back to the RMS and then the RMS integrates the received results into entire task output which is requested by the user.

The above grid service process can be approximated by a structure with star topology, as depicted by Fig. 2, where the RMS is directly connected with any resource through respective communication channels. The star topology is feasible when the resources are totally separated so that their communication channels are independent. Under this assumption the grid service reliability and performance can be derived by using the universal generating function technique.

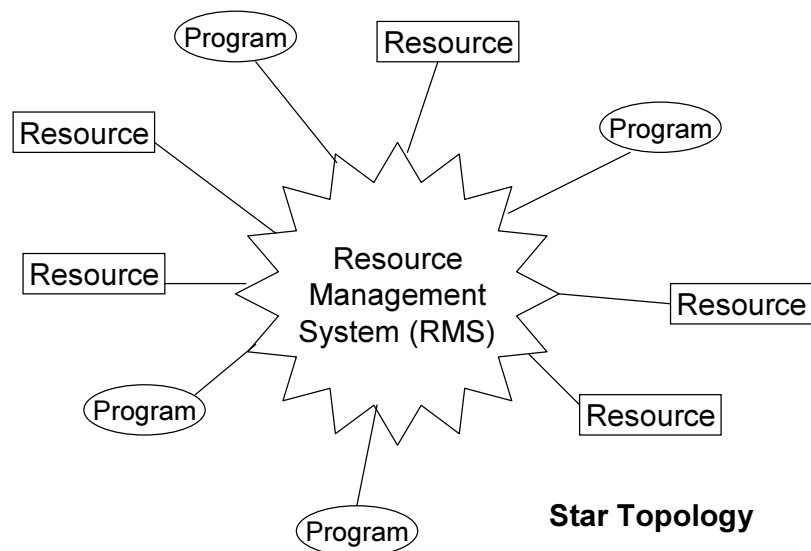


Fig. 2. Grid system with star architecture.

3.1. Universal Generating Function

The universal generating function (u-function) technique was introduced in (Ushakov, 1987) and proved to be very effective for the reliability evaluation of different types of multi-state systems.

The u-function representing the pmf of a discrete random variable Y is defined as a polynomial

$$u(z) = \sum_{k=1}^K \alpha_k z^{y_k}, \quad (10)$$

where the variable Y has K possible values and α_k is the probability that Y is equal to y_k .

To obtain the u-function representing the pmf of a function of two independent random variables $\varphi(Y_i, Y_j)$, composition operators are introduced. These operators determine the u-function for $\varphi(Y_i, Y_j)$ using simple algebraic operations on the individual u-functions of the variables. All of the composition operators take the form

$$U(z) = u_i(z) \underset{\varphi}{\otimes} u_j(z) = \sum_{k=1}^{K_i} \alpha_{ik} z^{y_{ik}} \underset{\varphi}{\otimes} \sum_{h=1}^{K_j} \alpha_{jh} z^{y_{jh}} = \sum_{k=1}^{K_i} \sum_{h=1}^{K_j} \alpha_{ik} \alpha_{jh} z^{\varphi(y_{ik}, y_{jh})} \quad (11)$$

The u-function $U(z)$ represents all of the possible mutually exclusive combinations of realizations of the variables by relating the probabilities of each combination to the value of function $\varphi(Y_i, Y_j)$ for this combination.

In the case of grid system, the u-function $u_{ij}(z)$ can define pmf of execution time for subtask i assigned to resource j . This u-function takes the form

$$u_{ij}(z) = p_j(\hat{t}_{ij}) z^{\hat{t}_{ij}} + (1 - p_j(\hat{t}_{ij})) z^{\infty} \quad (12)$$

where \hat{t}_{ij} and $p_j(\hat{t}_{ij})$ are determined according to Eqs. (2) and (3) respectively.

The pmf of the random start time T_i for subtask i can be represented by u-function $U_i(z)$ taking the form

$$U_i(z) = \sum_{l=1}^{L_i} q_{il} z^{\hat{T}_{il}}, \quad (13)$$

where $q_{il} = \Pr(T_i = \hat{T}_{il})$.

For any realization \hat{T}_{il} of T_i the conditional distribution of completion time \tilde{t}_{ij} for subtask i executed by resource j given $T_i = \hat{T}_{il}$ according to (6) can be represented by the u-function

$$\tilde{u}_{ij}(z, \hat{T}_{il}) = p_j(\hat{T}_{il} + \hat{t}_{ij})z^{\hat{T}_{il} + \hat{t}_{ij}} + (1 - p_j(\hat{T}_{il} + \hat{t}_{ij}))z^\infty. \quad (14)$$

The total completion time of subtask i assigned to a pair of resources j and d is equal to the minimum of completion times for these resources according to Eq. (7). To obtain the u-function representing the pmf of this time, given $T_i = \hat{T}_{il}$, composition operator with $\varphi(Y_j, Y_d) = \min(Y_j, Y_d)$ should be used:

$$\begin{aligned} \tilde{u}_i(z, \hat{T}_{il}) &= \tilde{u}_{ij}(z, \hat{T}_{il}) \otimes_{\min} \tilde{u}_{id}(z, \hat{T}_{il}) = [p_j(\hat{T}_{il} + \hat{t}_{ij})z^{\hat{T}_{il} + \hat{t}_{ij}} + (1 - p_j(\hat{T}_{il} + \hat{t}_{ij}))z^\infty] \\ &\quad \otimes_{\min} [p_d(\hat{T}_{il} + \hat{t}_{id})z^{\hat{T}_{il} + \hat{t}_{id}} + (1 - p_d(\hat{T}_{il} + \hat{t}_{id}))z^\infty] \\ &= p_j(\hat{T}_{il} + \hat{t}_{ij})p_d(\hat{T}_{il} + \hat{t}_{id})z^{\hat{T}_{il} + \min(\hat{t}_{ij}, \hat{t}_{id})} + p_d(\hat{T}_{il} + \hat{t}_{id})(1 - p_j(\hat{T}_{il} + \hat{t}_{ij}))z^{\hat{T}_{il} + \hat{t}_{id}} \\ &\quad + p_j(\hat{T}_{il} + \hat{t}_{ij})(1 - p_d(\hat{T}_{il} + \hat{t}_{id}))z^{\hat{T}_{il} + \hat{t}_{ij}} + (1 - p_j(\hat{T}_{il} + \hat{t}_{ij}))(1 - p_d(\hat{T}_{il} + \hat{t}_{id}))z^\infty. \end{aligned} \quad (15)$$

The u-function $\tilde{u}_i(z, \hat{T}_{il})$ representing the conditional pmf of completion time \tilde{T}_i for subtask i assigned to all of the resources from set $\omega_i = \{j_1, \dots, j_i\}$ can be obtained as

$$\tilde{u}_i(z, \hat{T}_{il}) = \tilde{u}_{ij_1}(z, \hat{T}_{il}) \otimes_{\min} \tilde{u}_{ij_2}(z, \hat{T}_{il}) \otimes_{\min} \dots \otimes_{\min} \tilde{u}_{ij_i}(z, \hat{T}_{il}). \quad (16)$$

$\tilde{u}_i(z, \hat{T}_{il})$ can be obtained recursively:

$$\begin{aligned} \tilde{u}_i(z, \hat{T}_{il}) &= \tilde{u}_{ij_1}(z, \hat{T}_{il}), \\ \tilde{u}_i(z, \hat{T}_{il}) &= \tilde{u}_i(z, \hat{T}_{il}) \otimes_{\min} \tilde{u}_{ie}(z, \hat{T}_{il}) \quad \text{for } e = j_2, \dots, j_i. \end{aligned} \quad (17)$$

Having the probabilities of the mutually exclusive realizations of start time T_i , $q_{il} = \Pr(T_i = \hat{T}_{il})$ and u-functions $\tilde{u}_i(z, \hat{T}_{il})$ representing corresponding conditional distributions of task i completion time, we can now obtain the u-function representing the unconditional pmf of completion time \tilde{T}_i as

$$\tilde{U}_i(z) = \sum_{l=1}^{N_i} q_{il} \tilde{u}_i(z, \hat{T}_{il}). \quad (18)$$

Having u-functions $\tilde{U}_k(z)$ representing pmf of the completion time \tilde{T}_k for any subtask $k \in \phi_i = \{k_1, \dots, k_i\}$, one can obtain the u-functions $U_i(z)$ representing pmf of subtask i start time T_i according to (4) as

$$U_i(z) = \tilde{U}_{k_1}(z) \otimes_{\max} \tilde{U}_{k_2}(z) \otimes_{\max} \dots \otimes_{\max} \tilde{U}_{k_i}(z) = \sum_{l=1}^{N_i} q_{il} z^{\hat{T}_{il}}. \quad (19)$$

$U_i(z)$ can be obtained recursively:

$$U_i(z) = z^0, \\ U_i(z) = U_i(z) \otimes_{\max} \tilde{U}_e(z) \quad \text{for } e = k_1, \dots, k_i. \quad (20)$$

It can be seen that if $\phi_i = \emptyset$ then $U_i(z) = z^0$.

The final u-function $U_m(z)$ represents the pmf of random task completion time T_m in the form

$$U_m(z) = \sum_{l=1}^{N_m} q_{ml} z^{\hat{T}_{ml}}. \quad (21)$$

Using the operators defined above one can obtain the service reliability and performance indices by implementing the following algorithm:

1. Determine \hat{t}_{ij} for each subtask i and resource $j \in \omega_i$ using Eq. (2);

Define for each subtask i ($1 \leq i \leq m$) $\tilde{U}_i(z) = U_i(z) = z^0$.

2. For all i :

If $\phi_i = \emptyset$ or if for any $k \in \phi_i$ $\tilde{U}_k(z) \neq z^0$ (u-functions representing the completion times of all of the predecessors of subtask i are obtained)

- 2.1. Obtain $U_i(z) = \sum_{l=1}^{N_i} q_{il} z^{\hat{T}_{il}}$ using recursive procedure (20);

- 2.2. For $l = 1, \dots, N_i$:

- 2.2.1. For each $j \in \omega_i$ obtain $\tilde{u}_{ij}(z, \hat{T}_{il})$ using Eq. (14);

- 2.2.2. Obtain $\tilde{u}_i(z)$ using recursive procedure (17);

2.3. Obtain $\tilde{U}_i(z)$ using Eq. (18).

3. If $U_m(z) = z^0$ return to step 2.

4. Obtain reliability and performance indices $R(\Theta^*)$ and W using equations (8) and (9).

3.2. Illustrative Example

This example presents analytical derivation of the indices $R(\Theta^*)$ and W for simple grid service that uses six resources. Assume that the RMS divides the service task into three subtasks. The first subtask is assigned to resources 1 and 2, the second subtask is assigned to resources 3 and 4, the third subtask is assigned to resources 5 and 6:

$$\omega_1 = \{1,2\}, \omega_2 = \{3,4\}, \omega_3 = \{5,6\}.$$

The failure rates of the resources and communication channels and subtask execution times are presented in Table 1.

Table 1. Parameters of grid system for analytical example

No of subtask i	No of resource j	$\lambda_j + \pi_j$ (sec ⁻¹)	\hat{t}_{ij} (sec)	$p_j(\hat{t}_{ij})$
1	1	0.0025	100	0.779
	2	0.00018	180	0.968
2	3	0.0003	250	-
	4	0.0008	300	-
3	5	0.0005	300	0.861
	6	0.0002	430	0.918

Subtasks 1 and 3 get the input data directly from the RMS, subtask 2 needs the output of subtask 1, the service task is completed when the RMS gets the outputs of both subtasks 2 and 3: $\phi_1 = \phi_3 = \emptyset$, $\phi_2 = \{1\}$, $\phi_4 = \{2,3\}$. These subtask precedence constraints can be represented by the directed graph in Fig. 3.

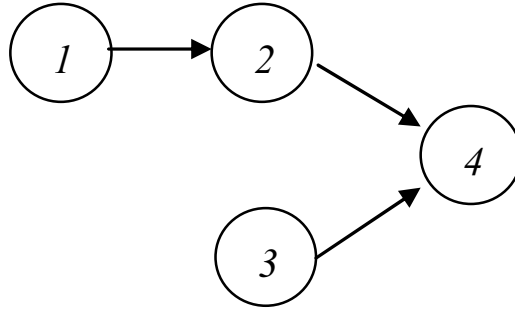


Fig. 3. Subtask execution precedence constraints for analytical example.

Since $\phi_1 = \phi_3 = \emptyset$, the only realization of start times T_1 and T_3 is 0 and therefore, $U_1(z) = U_2(z) = z^0$. According to step 2 of the algorithm we can obtain the u-functions representing pmf of completion times \tilde{t}_{11} , \tilde{t}_{12} , \tilde{t}_{35} and \tilde{t}_{36} . In order to determine the subtask execution time distributions for the individual resources, define the u-functions $u_{ij}(z)$ according to Table 1 and Eq. (9):

$$\tilde{u}_{11}(z,0) = \exp(-0.0025 \times 100)z^{100} + [1 - \exp(-0.0025 \times 100)]z^\infty = 0.779z^{100} + 0.221z^\infty.$$

In the similar way we obtain

$$\tilde{u}_{12}(z,0) = 0.968z^{180} + 0.032z^\infty;$$

$$\tilde{u}_{35}(z,0) = 0.861z^{300} + 0.139z^\infty; \quad \tilde{u}_{36}(z,0) = 0.918z^{430} + 0.082z^\infty.$$

The u-function representing the pmf of the completion time for subtask 1 executed by both resources 1 and 2 is

$$\begin{aligned} \tilde{U}_1(z) = \tilde{u}_1(z,0) &= \tilde{u}_{11}(z,0) \otimes_{\min} \tilde{u}_{12}(z,0) = (0.779z^{100} + 0.221z^\infty) \otimes_{\min} (0.968z^{180} + 0.032z^\infty) \\ &= 0.779z^{100} + 0.214z^{180} + 0.007z^\infty. \end{aligned}$$

The u-function representing the pmf of the completion time for subtask 3 executed by both resources 5 and 6 is

$$\begin{aligned} \tilde{U}_3(z) = \tilde{u}_3(z,0) &= \tilde{u}_{35}(z,0) \otimes_{\min} \tilde{u}_{36}(z,0) = (0.861z^{300} + 0.139z^\infty) \otimes_{\min} (0.918z^{430} + 0.082z^\infty) \\ &= 0.861z^{300} + 0.128z^{430} + 0.011z^\infty. \end{aligned}$$

Execution of subtask 2 begins immediately after completion of subtask 1. Therefore,

$$U_2(z) = \tilde{U}_1(z) = 0.779z^{100} + 0.214z^{180} + 0.007z^\infty$$

(T_2 has three realizations 100, 180 and ∞).

The u-functions representing the conditional pmf of the completion times for the subtask 2 executed by individual resources are obtained as follows.

$$\begin{aligned}\tilde{u}_{23}(z,100) &= e^{-0.0003 \times (100+250)} z^{100+250} + [1 - e^{-0.0003 \times (100+250)}] z^\infty = 0.9z^{350} + 0.1z^\infty; \\ \tilde{u}_{23}(z,180) &= e^{-0.0003 \times (180+250)} z^{180+250} + [1 - e^{-0.0003 \times (180+250)}] z^\infty = 0.879z^{430} + 0.121z^\infty; \\ \tilde{u}_{23}(z,\infty) &= z^\infty; \\ \tilde{u}_{24}(z,100) &= e^{-0.0008 \times (100+300)} z^{100+300} + [1 - e^{-0.0008 \times (100+300)}] z^\infty = 0.726z^{400} + 0.274z^\infty; \\ \tilde{u}_{24}(z,180) &= e^{-0.0008 \times (180+300)} z^{180+300} + [1 - e^{-0.0008 \times (180+300)}] z^\infty = 0.681z^{480} + 0.319z^\infty; \\ \tilde{u}_{24}(z,\infty) &= z^\infty.\end{aligned}$$

The u-functions representing the conditional pmf of subtask 2 completion time are:

$$\begin{aligned}\tilde{u}_2(z,100) &= \tilde{u}_{23}(z,100) \otimes_{\min} \tilde{u}_{24}(z,100) = (0.9z^{350} + 0.1z^\infty) \otimes_{\min} (0.726z^{400} + 0.274z^\infty) \\ &= 0.9z^{350} + 0.073z^{400} + 0.027z^\infty; \\ \tilde{u}_2(z,180) &= \tilde{u}_{23}(z,180) \otimes_{\min} \tilde{u}_{24}(z,180) = (0.879z^{430} + 0.121z^\infty) \otimes_{\min} (0.681z^{480} + 0.319z^\infty) \\ &= 0.879z^{430} + 0.082z^{480} + 0.039z^\infty; \\ \tilde{u}_2(z,\infty) &= \tilde{u}_{23}(z,\infty) \otimes_{\min} \tilde{u}_{24}(z,\infty) = z^\infty.\end{aligned}$$

According to Eq. (18) the unconditional pmf of subtask 2 completion time is represented by the following u-function

$$\begin{aligned}\tilde{U}_2(z) &= 0.779\tilde{u}_2(z,100) + 0.214\tilde{u}_2(z,180) + 0.007z^\infty \\ &= 0.779(0.9z^{350} + 0.073z^{400} + 0.027z^\infty) + 0.214(0.879z^{430} + 0.082z^{480} + 0.039z^\infty) + 0.007z^\infty \\ &= 0.701z^{350} + 0.056z^{400} + 0.188z^{430} + 0.018z^{480} + 0.037z^\infty\end{aligned}$$

The service task is completed when subtasks 2 and 3 return their outputs to the RMS (which corresponds to the beginning of subtask 4). Therefore, the u-function representing the pmf of the entire service time is obtained as

$$\begin{aligned}U_4(z) &= \tilde{U}_2(z) \otimes_{\max} \tilde{U}_3(z) \\ &= (0.701z^{350} + 0.056z^{400} + 0.188z^{430} + 0.018z^{480} + 0.037z^\infty) \otimes_{\max} (0.861z^{300} + 0.128z^{430} + \\ &\quad 0.011z^\infty) = 0.603z^{350} + 0.049z^{400} + 0.283z^{430} + 0.017z^{480} + 0.048z^\infty.\end{aligned}$$

The pmf of the service time is:

$$\Pr(T_4 = 350) = 0.603; \Pr(T_4 = 400) = 0.049;$$

$$\Pr(T_4 = 430) = 0.283; \Pr(T_4 = 480) = 0.017; \Pr(T_4 = \infty) = 0.048.$$

From the obtained pmf we can calculate the service reliability using Eq. (8):

$$R(\Theta^*) = 0.603 \text{ for } 350 < \Theta^* \leq 400; \quad R(\Theta^*) = 0.652 \text{ for } 400 < \Theta^* \leq 430;$$

$$R(\Theta^*) = 0.935 \text{ for } 430 < \Theta^* \leq 480; \quad R(\infty) = 0.952$$

and the conditional expected service time according to Eq. (9):

$$W = (0.603 \times 350 + 0.049 \times 400 + 0.283 \times 430 + 0.017 \times 480) / 0.952 = 378.69 \text{ sec.}$$

4. **Tree Topology** Grid Architecture

In the star grid, the RMS is connected with each resource by one direct communication channel (link). However, such approximation is not accurate enough even though it simplifies the analysis and computation. For example, several resources located in a same local area network (LAN) can use the same gateway to communicate outside the network. Therefore, all these resources are not connected with the RMS through independent links. The resources are connected to the gateway, which communicates with the RMS through one common communication channel. Another example is a server that contains several resources (has several processors that can run different applications simultaneously, or contains different databases). Such a server communicates with the RMS through the same links. These situations cannot be modeled using only the star topology grid architecture.

In this section, we present a more reasonable virtual structure which has a tree topology. The root of the tree virtual structure is the RMS, and the leaves are resources, while the branches of the tree represent the communication channels linking the leaves and the root. Some channels are commonly used by multiple resources. An example of the tree topology is given in Fig. 3 in which four resources (R1, R2, R3, R4) are available for a service.

The tree structure models the common cause failures in shared communication channels. For example, in Fig. 3, the failure in channel L6 makes resources R1, R2, and R3 unavailable. This type of common cause failure was ignored by the conventional parallel computing models, and the above star-topology models. For small-area communication, such as a LAN

or a cluster, such assumption that ignores the common cause failures on communications is acceptable because the communication time is negligible compared to the processing time. However, for wide-area communication, such as the grid system, it is more likely to have failure on communication channels. Therefore, the communication time cannot be neglected. In many cases, the communication time may dominate the processing time due to the large amount of data transmitted. Therefore, the virtual tree structure is an adequate model representing the functioning of grid services.

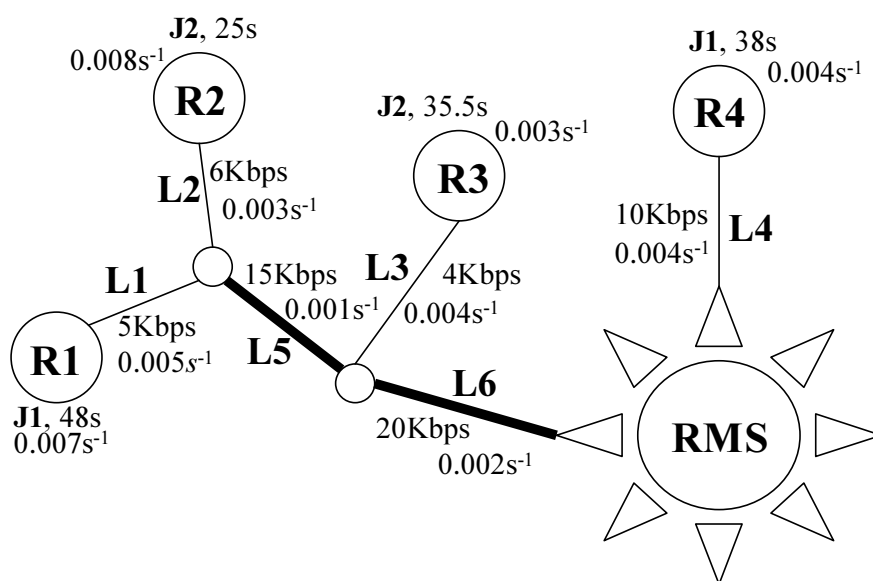


Fig. 3: A virtual tree structure of a grid service.

4.1. Algorithms for determining the pmf of the task execution time

With the tree-structure, the simple u-function technique is not applicable because it does not consider the failure correlations. Thus, new algorithms are required. This section presents a novel algorithm to evaluate the performance and reliability for the tree-structured grid service based on the graph theory and the Bayesian approach.

4.1.1. Minimal Task **Spanning Tree** (MTST)

The set of all nodes and links involved in performing a given task form a task spanning tree. This task spanning tree can be considered to be a combination of minimal task spanning trees (MTST), where each MTST represents a minimal possible combination of available elements

(resources and links) that guarantees the successful completion of the entire task. The failure of any element in a MTST leads to the entire task failure.

For solving the graph traversal problem, several classical algorithms have been suggested, such as Depth-First search, Breadth-First search, etc. These algorithms can find all MTST in an arbitrary graph (Dai *et al.*, 2002). However, MTST in graphs with a tree topology can be found in a much simpler way because each resource has a single path to the RMS, and the tree structure is acyclic.

After the subtasks have been assigned to corresponding resources, it is easy to find all combinations of resources such that each combination contains exactly m resources executing m different subtasks that compose the entire task. Each combination determines exactly one MTST consisting of links that belong to paths from the m resources to the RMS. The total number of MTST is equal to the total number of such combinations N , where

$$N = \prod_{j=1}^m |\omega_j| \quad (22)$$

(see Example 4.2.1).

Along with the procedures of searching all the MTST, one has to determine the corresponding running time and communication time for all the resources and links.

For any subtask j , and any resource k assigned to execute this subtask, one has the amount of input and output data, the bandwidths of links, belonging to the corresponding paths γ_k , and the resource processing time. With these data, one can obtain the time of subtask completion (see Example 4.2.2).

Some elements of the same MTST can belong to several paths if they are involved in data transmission to several resources. To track the element involvement in performing different subtasks and to record the corresponding times in which the element failure causes the failure of a subtask, we create the lists of two-field records for each subtask in each MTST. For any MTST S_i ($1 \leq i \leq N$), and any subtask j ($1 \leq j \leq m$), this list contains the names of the elements involved in performing the subtask j , and the corresponding time of subtask completion y_{ij} (see Example 4.2.3). Note that y_{ij} is the conditional time of subtask j completion given only MTST i is available.

Note that a MTST completes the entire task if all of its elements do not fail by the maximal time needed to complete subtasks in performing which they are involved. Therefore, when calculating the element reliability in a given MTST, one has to use the corresponding record with maximal time.

4.1.2. pmf of the task execution time

Having the MTST, and the times of their elements involvement in performing different subtasks, one can determine the pmf of the entire service time.

First, we can obtain the conditional time of the entire task completion given only MTST S_i is available as

$$Y_{\{i\}} = \max_{1 \leq j \leq m} (y_{ij}) \text{ for any } 1 \leq i \leq N: \quad (23)$$

For a set ψ of available MTST, the task completion time is equal to the minimal task completion times among the MTST.

$$Y_{\psi} = \min_{i \in \psi} (Y_{\{i\}}) = \min_{i \in \psi} \left[\max_{1 \leq j \leq m} (y_{ij}) \right]. \quad (24)$$

Now, we can sort the MTST in an increasing order of their conditional task completion times $Y_{\{i\}}$, and divide them into different groups containing MTST with identical conditional completion time. Suppose there are K such groups denoted by G_1, G_2, \dots, G_K where $1 \leq K \leq N$, and any group G_i contains MTST with identical conditional task completion times θ_i ($0 \leq \theta_1 < \theta_2 < \dots < \theta_K$). Then, it can be seen that the probability $Q_i = \Pr(\Theta = \theta_i)$ can be obtained as

$$Q_i = \Pr(E_i, \bar{E}_{i-1}, \bar{E}_{i-2}, \dots, \bar{E}_1) \quad (25)$$

where E_i is the event when at least one of MTST from the group G_i is available, and \bar{E}_i is the event when none of MTST from the group G_i is available.

Suppose the MTST in a group G_i are arbitrarily ordered, and F_{ij} ($j=1,2,\dots, N_i$) represents an event when the j -th MTST in the group is available. Then, the event E_i can be

expressed by

$$E_i = \bigcup_{j=1}^{N_i} F_{ij}, \quad (26)$$

and (25) takes the form

$$\Pr(E_i, \bar{E}_{i-1}, \bar{E}_{i-2}, \dots, \bar{E}_1) = \Pr\left(\bigcup_{j=1}^{N_i} F_{ij}, \bar{E}_{i-1}, \bar{E}_{i-2}, \dots, \bar{E}_1\right). \quad (27)$$

Using the Bayesian theorem on conditional probability, we obtain from (27) that

$$Q_i = \sum_{j=1}^{N_i} \Pr(F_{ij}) \cdot \Pr\left(\bar{F}_{i(j-1)}, \bar{F}_{i(j-2)}, \dots, \bar{F}_{i1}, \bar{E}_1, \bar{E}_2, \dots, \bar{E}_{i-1} \mid F_{ij}\right). \quad (28)$$

The probability $\Pr(F_{ij})$ can be calculated as a product of the reliabilities of all the elements belonging to the j -th MTST from group G_i .

The probability $\Pr\left(\bar{F}_{i(j-1)}, \bar{F}_{i(j-2)}, \dots, \bar{F}_{i1}, \bar{E}_1, \bar{E}_2, \dots, \bar{E}_{i-1} \mid F_{ij}\right)$ can be computed by the following two-step algorithm (see Example 4.2.4).

Step 1: Identify failures of all the critical elements' in a period of time (defined by the start and end time), during which they lead to the failures of any MTST from groups G_m for $m=1,2,\dots,i-1$ (events \bar{E}_m), and any MTST S_k from group G_i for $k=1,2,\dots,j-1$ (events \bar{F}_{ik}), but do not affect the MTST S_j from group G_i .

Step 2: Generate all the possible combinations of the identified critical elements that lead to the event $\bar{F}_{i(j-1)}, \bar{F}_{i(j-2)}, \dots, \bar{F}_{i1}, \bar{E}_1, \bar{E}_2, \dots, \bar{E}_{i-1} \mid F_{ij}$ using a binary search, and compute the probabilities of those combinations. The sum of the probabilities obtained is equal to $\Pr\left(\bar{F}_{i(j-1)}, \bar{F}_{i(j-2)}, \dots, \bar{F}_{i1}, \bar{E}_1, \bar{E}_2, \dots, \bar{E}_{i-1} \mid F_{ij}\right)$. When calculating the failure probabilities of MTSTs' elements, the maximal time from the corresponding records in a list for the given MTST should be used. The algorithm for obtaining the probabilities $\Pr\{\bar{E}_1, \bar{E}_2, \dots, \bar{E}_{i-1} \mid E_i\}$ can be found in Dai *et al.* (2002).

Having the conditional task completion times $Y_{\{i\}}$ for different MTST, and the corresponding probabilities Q_i , one obtains the task completion time distribution (Θ_i, Q_i) , $1 \leq i \leq K$, and can easily calculate the indices (8) & (9) (see Example 4.2.5).

4.2. Illustrative Example

Consider the virtual grid presented in Fig. 3, and assume that the service task is divided into two subtasks J1 assigned to resources R1 & R4, and J2 assigned to resources R2 & R3. J1, and J2 require 50Kbits, and 30Kbits of input data, respectively, to be sent from the RMS to the corresponding resource; and 100Kbits, and 60Kbits of output data respectively to be sent from the resource back to the RMS.

The subtask processing times for resources, bandwidth of links, and failure rates are presented in Fig. 3 next to the corresponding elements.

4.2.1. The service MTST

The entire graph constitutes the task spanning tree. There exist four possible combinations of two resources executing both subtasks: {R1, R2}, {R1, R3}, {R4, R2}, {R4, R3}. The four MTST corresponding to these combinations are: S_1 : {R1, R2, L1, L2, L5, L6}; S_2 : {R1, R3, L1, L3, L5, L6}; S_3 : {R2, R4, L2, L5, L4, L6}; S_4 : {R3, R4, L3, L4, L6}.

4.2.2. Parameters of MTSTs' paths

Having the MTST, one can obtain the data transmission speed for each path between the resource, and the RMS (as minimal bandwidth of links belonging to the path); and calculate the data transmission times, and the times of subtasks' completion. These parameters are presented in Table 2. For example, resource R1 (belonging to two MTST S_1 & S_2) processes subtask J1 in 48 seconds. To complete the subtask, it should receive 50Kbits, and return to the RMS 100Kbits of data. The speed of data transmission between the RMS and R1 is limited by the bandwidth of link L1, and is equal to 5 Kbps. Therefore, the data transmission time is $150/5=30$ seconds, and the total time of task completion by R1 is $30+48=78$ seconds.

Table 2: Parameters of the MTSTs' paths

Elements, subtasks	R1, J1	R2, J2	R3, J2	R4, J1
Data transmission speed (Kbps)	5	6	4	10
Data transmission time (s)	30	15	22.5	15
Processing time (s)	48	25	35.5	38
Time to subtask completion (s)	78	40	58	53

4.2.3. List of MTST elements

Now one can obtain the lists of two-field records for components of the MTST.

S_1 : path for J1:(R1,78); (L1,78); (L5,78); (L6,78); path for J2: (R2,40); (L2,40); (L5,40); (L6,40).

S_2 : path for J1: (R1,78), (L1,78), (L5,78), (L6,78); path for J2: (R3,58), (L3,58), (L6,58).

S_3 : path for J1: (R4,53), (L4,53); path for J2: (R2,40), (L2,40), (L5,40), (L6,40).

S_4 : path for J1: (R4,53), (L4,53); path for J2: (R3,58), (L3,58), (L6,58).

4.2.4. pmf of task completion time

The conditional times of the entire task completion by different MTST are

$$Y_1=78; \quad Y_2=78; \quad Y_3=53; \quad Y_4=58.$$

Therefore, the MTST compose three groups:

$$G_1 = \{S_3\} \text{ with } \Theta_1 = 53; G_2 = \{S_4\} \text{ with } \Theta_2 = 58; \text{ and } G_3 = \{S_1, S_2\} \text{ with } \Theta_3 = 78.$$

According to (25), we have for group G_1 : $Q_1 = \Pr(E_1) = \Pr(S_3)$. The probability that the MTST S_3 completes the entire task is equal to the product of the probabilities that R4, and L4 do not fail by 53 seconds; and R2, L2, L5, and L6 do not fail by 40 seconds.

$$\begin{aligned} \Pr(\Theta=53) &= Q_1 = \exp(-0.004 \times 53) \exp(-0.004 \times 53) \exp(-0.008 \times 40) \\ &\quad \times \exp(-0.003 \times 40) \exp(-0.001 \times 40) \exp(-0.002 \times 40) = 0.3738. \end{aligned}$$

Now we can calculate Q_2 as

$$Q_2 = \Pr(E_2, \bar{E}_1) = \Pr(F_{21}) \Pr(\bar{E}_1 | F_{21}) = \Pr(F_{21}) \Pr(\bar{F}_{11} | F_{21}) = \Pr(S_4) \Pr(\bar{S}_3 | S_4)$$

because G_2 , and G_1 have only one MTST each. The probability that the MTST S_4 completes the entire task $\Pr(S_4)$ is equal to the product of probabilities that R3, L3, and L6 do not fail by 58 seconds; and R4, and L4 do not fail by 53 seconds.

$$\begin{aligned} \Pr(S_4) &= \exp(-0.004 \times 53) \exp(-0.003 \times 58) \exp(-0.004 \times 53) \exp(-0.004 \times 58) \exp(-0.002 \times 58) \\ &= 0.3883 \end{aligned}$$

To obtain $\Pr(\bar{S}_3 | S_4)$, one first should identify the critical elements according to the algorithm presented in the Dai *et al.* (2002). These elements are R2, L2, and L5. Any failure occurring in one of these elements by 40 seconds causes failure of S_3 , but does not affect S_4 . The probability that at least one failure occurs in the set of critical elements is

$$\Pr(\bar{S}_3|S_4)=1-\exp(-0.008\times 40)\exp(-0.003\times 40)\exp(-0.001\times 40)=0.3812.$$

Then,

$$\Pr(\Theta=58)=\Pr(E_2, \bar{E}_1)=\Pr(S_4)\Pr(\bar{S}_3|S_4)=0.3883\times 0.3812=0.1480.$$

Now one can calculate Q_3 for the last group $G_3 = \{S_1, S_2\}$ corresponding to $\Theta_3 = 78$ as

$$\begin{aligned} Q_3 &= \Pr(E_3, \bar{E}_2, \bar{E}_1) = \Pr(F_{31})\Pr(\bar{E}_1, \bar{E}_2|F_{31}) + \Pr(F_{32})\Pr(\bar{F}_{31}, \bar{E}_1, \bar{E}_2|F_{32}) \\ &= \Pr(S_1)\Pr(\bar{S}_3, \bar{S}_4|S_1) + \Pr(S_2)\Pr(\bar{S}_1, \bar{S}_3, \bar{S}_4|S_2) \end{aligned}$$

The probability that the MTST S_1 completes the entire task is equal to the product of the probabilities that R1, L1, L5, and L6 do not fail by 78 seconds; and R2, and L2 do not fail by 40 seconds.

$$\begin{aligned} \Pr(S_1) &= \exp(-0.007\times 78)\exp(-0.008\times 40)\exp(-0.005\times 78)\exp(-0.003\times 40) \\ &\quad \times \exp(-0.001\times 78)\exp(-0.002\times 78) = 0.1999. \end{aligned}$$

The probability that the MTST S_2 completes the entire task is equal to the product of the probabilities that R1, L1, L5, and L6 do not fail by 78 seconds; and R3, and L3 do not fail by 58 seconds.

$$\begin{aligned} \Pr(S_2) &= \exp(-0.007\times 78)\exp(-0.003\times 58)\exp(-0.005\times 78)\exp(-0.004\times 58) \\ &\quad \times \exp(-0.001\times 78)\exp(-0.002\times 78) = 0.2068. \end{aligned}$$

To obtain $\Pr(\bar{S}_3, \bar{S}_4|S_1)$, one first should identify the critical elements. Any failure of either R4 or L4 in the time interval from 0 to 53 seconds causes failures of both S_3 , and S_4 ; but does not affect S_1 . Therefore,

$$\Pr(\bar{S}_3, \bar{S}_4|S_1)=1-\exp(-0.004\times 53)\exp(-0.004\times 53)=0.3456.$$

The critical elements for calculating $\Pr(\bar{S}_1, \bar{S}_3, \bar{S}_4|S_2)$ are R2, and L2 in the interval from 0 to 40 seconds; and R4, and L4 in the interval from 0 to 53 seconds. The failure of both elements in any one of the following four combinations causes failures of S_3 , S_4 , and S_1 , but does not affect S_2 :

1. R2 during the first 40 seconds, and R4 during the first 53 seconds;
2. R2 during the first 40 seconds, and L4 during the first 53 seconds;
3. L2 during the first 40 seconds, and R4 during the first 53 seconds; and

4. L2 during the first 40 seconds, and L4 during the first 53 seconds.

Therefore,

$$\Pr(\bar{S}_1, \bar{S}_3, \bar{S}_4 | S_2) = 1 - \prod_{i=1}^4 \left[1 - \prod_{j=1}^2 [1 - \exp(-\lambda_{ij} \cdot t_{ij})] \right] = 0.1230,$$

where λ_{ij} is the failure rate of the j -th critical element in the i -th combination ($j=1,2$), ($i=1,2,3,4$); and t_{ij} is the duration of the time interval for the corresponding critical element.

Having the values of $\Pr(S_1)$, $\Pr(S_2)$, $\Pr(\bar{S}_3, \bar{S}_4 | S_1)$, and $\Pr(\bar{S}_1, \bar{S}_3, \bar{S}_4 | S_2)$, one can calculate

$$\Pr(\Theta=78) = Q_3 = 0.1999 \times 0.3456 + 0.2068 \times 0.1230 = 0.0945.$$

After obtaining Q_1 , Q_2 , and Q_3 , one can evaluate the total task failure probability as

$$\Pr(\Theta=\infty) = 1 - Q_1 - Q_2 - Q_3 = 1 - 0.3738 - 0.1480 - 0.0945 = 0.3837,$$

and obtain the pmf of service time presented in Table 3.

Table 3: pmf of service time.

θ_i	Q_i	$\theta_i Q_i$
53	0.3738	19.8114
58	0.1480	8.584
78	0.0945	7.371
∞	0.3837	∞

4.2.5. Calculating the reliability indices.

From Table 3, we obtain the probability that the service does not fail as

$$R(\infty) = Q_1 + Q_2 + Q_3 = 0.6164,$$

the probability that the service time is not greater than a pre-specified value of $\theta^*=60$ seconds as

$$R(\theta^*) = \sum_{i=1}^3 Q_i \cdot 1(\theta_i < \theta^*) = 0.3738 + 0.1480 = 0.5218,$$

and the expected service execution time given that the system does not fail as

$$W = \sum_{i=1}^3 \theta_i Q_i / R(\infty) = 35.7664 / 0.6164 = 58.025 \text{ seconds.}$$

4.3. Parameterization and Monitoring

In order to obtain the reliability and performance indices of the grid service one has to know such model parameters as the failure rates of the virtual links and the virtual nodes, and bandwidth of the links. It is easy to estimate those parameters by implementing the monitoring technology.

A monitoring system (called *Alertmon Network Monitor*, <http://www.abilene.iu.edu/noc.html>) is being applied in the IP-grid (Indiana Purdue Grid) project (www.ip-grid.org), to detect the component failures, to record service behavior, to monitor the network traffics and to control the system configurations.

With this monitoring system, one can easily obtain the parameters required by the grid service reliability model by adding the following functions in the monitoring system:

- 1) Monitoring the failures of the components (virtual links and nodes) in the grid service, and recording the total execution time of those components. The failure rates of the components can be simply estimated by the number of failures over the total execution time.
- 2) Monitoring the real time network traffic of the involved channels (virtual links) in order to obtain the bandwidth of the links.

To realize the above monitoring functions, network sensors are required. We presented a type of sensors attaching to the components, acting as neurons attaching to the skins. It means the components themselves or adjacent components play the roles of sensors at the same time when they are working. Only a little computational resource in the components is used for accumulating failures/time and for dividing operations, and only a little memory is required for saving the data (accumulated number of failures, accumulated time and current bandwidth). The virtual nodes that have memory and computational function can play the sensing role themselves; if some links have no CPU or memory then the adjacent processors or routers can perform this data collecting operations. Using such self-sensing technique avoids overloading of the monitoring center even in the grid system containing numerous components. Again, it does not affect the service performance considerably since only small part of computation and storage resources is used for the monitoring. In addition, such self-sensing technique can also be applied in monitoring other measures.

When evaluating the grid service reliability, the RMS automatically loads the required parameters from corresponding sensors and calculates the service reliability and performance according to the approaches presented in the previous sections. This strategy can also be used for implementing the Autonomic Computing concept.

5. Conclusions

Grid computing is a newly developed technology for complex systems with large-scale resource sharing, wide-area communication, and multi-institutional collaboration. Although the developmental tools and techniques for the grid have been widely studied, grid reliability analysis and modeling are not easy because of their complexity of combining various failures.

This chapter introduced the grid computing technology and analyzed the grid service reliability and performance under the context of performability. The chapter then presented models for star-topology grid with data dependence and tree-structure grid with failure correlation. Evaluation tools and algorithms were presented based on the universal generating function, graph theory, and Bayesian approach. Numerical examples are presented to illustrate the grid modeling and reliability/performance evaluation procedures and approaches.

Future research can extend the models for grid computing to other large-scale distributed computing systems. After analyzing the details and specificity of corresponding systems, the approaches and models can be adapted to real conditions. The models are also applicable to wireless network that is more failure prone.

Hierarchical models can also be analyzed in which output of lower level models can be considered as the input of the higher level models. Each level can make use of the proposed models and evaluation tools.

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Key Terms and Their Definitions

Reliability: the probability for the service to be successfully completed given a execution time.

Performance: the inverse of the execution time.

Grid computing: Grid computing is a newly developed technology for complex systems with large-scale resource sharing, wide-area communication, and multi-institutional collaboration.

Modeling: a representation, generally in mathematical presentations, to show the construction or appearance of a computing system.

Graph theory: Use graph algorithms to analyze given a network graph.

Bayesian analysis: Use Bayes method to get the posterior distribution from a prior distribution.

Universal generating function: Also called as u-function that is a technique to express and evaluate models in a polynomial format.