REAL TIME RESOURCE ALLOCATION
IN A DISTRIBUTED SYSTEM

John Reif
Paul Spirakis

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**Abstract:** See reverse side.
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REAL TIME RESOURCE ALLOCATION
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by

John Reif and Paul Spirakis
Harvard University

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ABSTRACT

In this paper we consider a resource allocation problem which is local in the sense that the maximum number of users competing for a particular resource at any time instant is bounded and also at any time instant the maximum number of resources that a user is willing to get is bounded. The problem may be viewed as that of achieving matchings in dynamically changing hypergraphs, via a distributed algorithm. We show that this problem is related to the fundamental problem of handshake communication (which can be viewed as achieving matchings in a dynamically changing graph, via distributed algorithms) in that an efficient solution to each of them implies an efficient solution to the other. We provide real-time solutions to the resource allocation problem (that is, we give distributed algorithms with real time response). We make essential use of probabilistic techniques as first used by [Rabin, 80b], where processes are allowed to make independent probabilistic choices. On the other hand, no probability assumptions about the system behavior are made. One of our solutions assumes the existence of an underlying real-time handshake communication system, as described in [Reif, Spirakis, 81]. Our other solution is based on efficient synchronization by flag variables, which are written only by one process and read by at most one other process. The special case of equi-speed processes is first examined. Then we generalize to asynchronous processes. Applications are made to dining philosophers, scheduling and two-phase locking in databases.
I. INTRODUCTION

1.1 The Resource Granting System

In this paper we consider a resource allocation problem which is
local in the sense described in [Lynch, 1980]. The set of resources $\rho$
and the set of user processes $U$ may be infinite sets. However, there is
a limit to the number of user processes requesting a particular resource.
We assume that names of processes are integers. Resources are controlled
by a set of granting processes $R$. Each granting process $j \in R$ controls
a resource $\rho(j) \in \rho$. We assume user processes communicate only with those
granting processes for which they request resources. (It is easy to super-
impose each granting process into a requesting user process so that $U = R$.)

A system described as above, is called a Resource Granting System
(RGS). The goal of a RGS is to satisfy dynamically changing user requests
for resource allocation. This is done in a distributed way, by only a local
communication between granting and requesting processes. An implementation of the
RGS determines the synchronization algorithms that the processes run. It is symmetric
if these algorithms do not depend on the location of the processes in the net-
work. At each time $t \geq 0$ the requests by user processes $U$ are specified
by an adverse oracle $\mathcal{A}$, (which may be an "enemy" of the resource allocation
algorithm, by setting actions in the worst way to increase the response
time.) (Note that in practice no such oracle $\mathcal{A}$ may exist but instead each
user process is executing some program which requires from time to time some
resources to be allocated to that user process. The oracle $\mathcal{A}$ is used as an
artificial device for specifying worst case situations for the resource
allocation algorithm.) $\mathcal{A}$ also has the capability to select at time $t = 0$
the schedule of the speeds of all processes at all times $t \geq 0$. The oracle $\mathcal{A}$
is restricted, to allow users to keep asking for their resources until they are granted. We assume that there is a global time $t$ totally ordering events, but processes do not have access to it.

An RGS with priorities is defined to be a resource granting system in which the requesting processes communicate to each resource granting process a rational number on the interval $[0,1]$, indicating the priority of the request. A priority greater than 0 indicates a request for that resource. These priorities can change dynamically. The processes of $R$ use the priorities, that the processes of $U$ communicate to them, to grant their controlled resource with preference to users of higher priority. This must be done in a way avoiding user starvation.

1.2 Complexity of an RGS

A process step consists of either an assignment of a variable, a test, a logical or arithmetic operator or a no-op. A step is considered to be a finite time interval $\Delta$ in which a single instruction is executed instantaneously at the last moment of $\Delta$. Note that, by these semantics, there can never be read-write conflicts in the use of flag variables.

Let a process be tame during a time interval $\Delta$, if for any interval $\Delta' \in [0,\infty)$, which intersects $\Delta$ and is a single step of the process, then $|\Delta'| \in [r_{\min}, r_{\max}]$ where $r_{\min}$, $r_{\max}$ are fixed real constants and $0 < r_{\min} \leq r_{\max}$. Without loss of generality we assume that $r_{\max}/r_{\min}$ is an integer. We do not assume processes always tame, however, they are supposed to be tame in the complexity analysis of response time. We require that, at no time, any granting process $i \in R$ simultaneously grant the resource $R(i)$ to more than one requesting process. We also require that, as soon as a
process \( j \in U \) has got all its required resources, then it can keep them only for a time interval upper bounded by a fixed parameter (resulting in a bounded number of its steps \( \mu \), if the process is tame). Let \( \text{resources}(i) \) be the set of all possible resources that a process \( i \in U \) is going to ever request, and let \( \text{resources}_t(i) \) be the set of resources \( i \) is requesting at time instant \( t \). Let \( k_{i,t} \) be \( |\text{resources}_t(i)| \). For each time \( t \), the willingness digraph \( G_t \) is defined such that if \( i \in U \) and \( j \in R \) and if \( i \) requests or has been granted resource \( p(j) \) at \( t \), then the edge \( i \xrightarrow{t} j \) belongs to \( G_t \). Let also \( j \xrightarrow{t} i \) if the granting process \( j \) is willing to allocate (or has granted) its resource to user \( i \). Let \( v_t \) be the maximum valence of the nodes \( j \in R \) of \( G_t \).

In the following we will assume at all times \( t \geq 0 \) that \( v_t \) and \( k_{i,t} \) are above bounded by constants \( v, k \) respectively and that \( v < k \). This does not imply anything about the cardinality of the sets \( \text{resources}(i) \forall i \in U \), which could be unbounded. We also assume, as in [Lynch, 1980], that each resource allocator \( j \in R \) has a set \( S_j \) available to it of size \( \leq v \), containing the names of those processes willing to get the resource. We assume this to be a primitive of our system (which could be implemented by a queued message system or by other means). Finally, we restrict any oracle \( \mathcal{A} \) so that as soon as it produces a request for a resource (i.e., it orders the appearance of the edge \( i \xrightarrow{t} j \) in \( G_t \) for \( i \in U, j \in R \)) it has to insist on that request until \( i \) gets all its requested resources at that time. (Note that the relation \( \xrightarrow{t} \) can be viewed as a time varying hypergraph with node set \( \Pi = U \cup R \) and hyperedge set \( E = \{\{i\} \cup \text{resources}_t(i) | i \in U\} \). An RGS implementation dynamically achieves matchings in this hypergraph. We allow probabilistic RGS implementations where processes can use independent random number generators.
Fix an RGS implementation which may be probabilistic. For each \( k \) (\( 0 < k < v \)) and oracle \( \mathcal{A} \) let the \( k \)-grant response be the random variable \( Y_{k, \mathcal{A}} \) giving the length of the minimum interval \( \Delta \) required for any process \( i \in U \) to have \( k \) resource requests simultaneously granted, given that \( i \) requested these resources during the entire interval \( \Delta \), with priority 1, and assuming that \( i \) and all the allocators of the resources requested by \( i \) within \( \Delta \) are tame on \( \Delta \). Let the mean \( k \)-grant response be

\[
\bar{\gamma}_k = \max \{ \text{mean}(Y_{k, \mathcal{A}}) \} \text{ over all oracles } \mathcal{A}. \]

For each \( \epsilon \) in \([0,1]\) let the \( \epsilon \)-error \( k \)-grant response be the minimum \( \gamma_k(\epsilon) \) such that for every oracle

\[
\text{Prob}(Y_{k, \mathcal{A}} < \gamma_k(\epsilon)) > 1 - \epsilon. \]

The RGS implementation is real time if for every \( k \in \{1, \ldots, v\} \) and for every \( \epsilon \in (0,1) \), \( \gamma_k(\epsilon) > 0 \) and independent of any global measure of the network (except \( v \)). The network here has as nodes the elements of \( \Pi \) and each edge \( \{u, r\} \) denotes that \( \rho(r) \in \text{resources}(u) \). Note that if the RGS implementation is real time, then \( \bar{\gamma}_k \) is constant, independent of any global measure of the graph \( \Pi \) of the network (except perhaps \( v \)).

### 1.3 Previous Work

[Lynch, 1980] considered the problem of allocation of resources in a distributed system. Her RGS implementation was a deterministic, non-symmetric one (processes were allowed to know the color of each resource in a coloration of the resource graph, i.e. the graph whose nodes are the resources and two resources \( i, j \) have an edge between them iff \( \text{resource}^{-1}(i) \cap \text{resource}^{-1}(j) \neq \emptyset \)), and the communication system adopted was a message system requiring buffered communication. [Lynch, 1980] achieved
k-grant response of the order $\chi(H) \cdot v^H \cdot T$ where $\chi(H)$ is the chromatic number of the resource graph and $T$ is the time required for process communication. Note that since resources(i) for $i \in U$ may be unbounded, in the worst case $\chi(H)$ can be as large as the number of processes $|\Pi|$.

1.4 Results of this Paper

We shall present in Section 3 a probabilistic implementation of an RGS which has mean k-grant response $\gamma_k = O(kv^k \cdot T \cdot \log v)$ and $\epsilon$-error k-grant response

$$\gamma_k(\epsilon) = O\left(kv^k \log \left(\frac{1}{\epsilon}\right) \cdot T\left(\frac{1}{2v}\right)\right)$$

where $T$ and $T(\epsilon)$ are the mean time and $\epsilon$-error response required for handshake communication between processes (see Appendix for definitions). In [Reif, Spirakis 1981], [Reif, Spirakis 1982] handshake communication implementations were given with $T = O(v^2)$ and $T(\epsilon) = O(v^2 \log (1/\epsilon))$. Note that these implementations achieve real time, thus the resulting RGS implementation is also real time with mean

$$\gamma_k = O(kv^{k+2} \log v) \quad \text{and} \quad \gamma_k(\epsilon) = O\left(kv^{k+2} \log \left(\frac{v}{\epsilon}\right) \log \left(\frac{1}{\epsilon}\right)\right).$$

However, any other handshake communication implementation would also do.

In Section 4 we present a basic way to implement a real time RGS in both cases of equispeed and asynchronous tame processes. The implementation is probabilistic. No underlying handshake communication system is assumed. Instead, the means of synchronization between processes in $U$ and $R$ are flag variables (which are written by just one process and allowed to be read by at most one other process). The response time has mean $\gamma_k = O(kv^{k+1})$ and $\gamma_k(\epsilon) = O(kv^{k+1} \log (1/\epsilon))$. 
1.5 Organization of Paper

This paper is organized as follows: Section 2 contains applications of RGS to dining philosophers, scheduling, two-phase locking in databases, and real-time handshake communications. Section 3 discusses a real time RGS assuming an underlying real time handshake communication system. Section 4 discusses a real time RGS implementation by use of flags. First an implementation is discussed in which processes have the same speeds but their actions may be relatively shifted in time. At the end of Section 4 we generalize our algorithms to the case where processes are asynchronous but tame. Section 5 presents a probabilistic analysis of our algorithms. The Appendix defines handshake communication systems and their performance measures.

2. APPLICATIONS

2.1 Hasty Dining Philosophers

As a simple example, we consider an interesting RGS system which we call "hasty dining philosophers". Let the requesting processes \( U \) be distinct integers, \( r_0, \ldots, r_{n-1} \), and the granting processes \( R \) be \( 0, \ldots, n-1 \) so that \( U \cap R = \emptyset \). The resources are forks \( \{ p(0), \ldots, p(n-1) \} \). Each philosopher \( r_j \in U \) has resources \( r_j \) consisting of the forks \( \{ p(r_j), p(r_{(j+1) \mod n}) \} \).

Thus, the resource graph \( H \) is a cycle of length \( n \). The "hasty dining philosophers" has a high level real time RGS implementation with mean 2-grant response \( \tilde{\gamma}_2 = O(1) \) and \( \varepsilon \)-error 2-grant response \( \gamma_2(\varepsilon) = O(\log(1/\varepsilon))^2 \). The low level RGS implementation gives \( \tilde{\gamma}_2 = O(1) \) and \( \gamma_2(\varepsilon) = O(\log(1/\varepsilon)) \).

Intuitively, the RGS implementation requires each philosopher \( r_j \) to be at
any time granted both forks of resource(r_{j}) in expected constant time, but 
\textit{r}_{j} must be "hasty" and relinquish these resources within constant time inter-
val. Note that for each \( i \in \{0, \ldots, n-1\} \) the granting process \( i \) can be 
placed within process \( r_{i} \), thus resulting in essentially only \( n \) processes.

2.2 Scheduling

Suppose an acyclic digraph \( D \) is given with node out-degree \( \leq k \), and 
in-degree \( \leq v \). This graph can be separated into levels. We assume processes 
residing in the nodes of \( D \) can operate only after getting all their resources 
(which reside at the nodes which have no successors). Each process operates 
just once and in a constant time interval becomes a resource granting process, 
to serve the next higher level. All its successors are deleted. So, each 
node is initially a requesting process and after it gets all its resources 
once, it becomes a resource granting process. By assuming a real time RGS 
implementation the above system can be processed in time of the order of the 
depth of the digraph.

2.3 Two-Phase Locking in Databases

Two-phase locking is a concurrency control method in databases (see 
the survey paper of [Bernstein, Goodman 1980]) with the feature that as soon 
as a transaction releases a lock, it never obtains additional locks. The 
technique of two-phase locking produces serializable transaction executions. 
We propose in [Reif, Spirakis 1982B] a technique to implement two-phase 
locking in a distributed database, which is different from any known algo-
rithms. In fact, it can be viewed as intermediate between the techniques of 
static and dynamic locking. Our underlying assumption is that transactions
are allowed to act on the data only if they got all the locks requested. Let the processes of the user set $U$ be called transaction modules and the processes of the set $R$ be called data modules. If each transaction requests to lock at most $k$ data modules at a time and if at most $v$ transactions can compete for a lock at a time instant $t$, then a real time RGS will result in real time lock allocation per transaction. The transaction modules go through a "round" (of constant number of steps in duration) during which they communicate with all data modules they want to lock. They keep the locks they get only for constant number of steps hoping in the meanwhile to get all of them. If the round finishes and they did not succeed in getting all the locks, then they release whatever they got and try again in the next round. Given the previously stated response time $\gamma_k(\varepsilon)$ for a real-time RGS, one can now decompose the distributed system into a system of parallel servers with almost independent geometrically distributed service time, (one per transaction module). In [Reif, Spirakis, 1982B] we then analyze the transaction waiting time, throughput etc., by assuming a probability distribution in the transaction arrival rate.

2.4 Handshake Communication in Real Time Via a Real Time RGS

We can implement here handshake communication in the sense of the Appendix, assuming the existence of a real-time RGS. Assume that each of the processes $j \in R$ control just one resource called the channel and when they allocate this resource, we say they open the channel. The processes $i \in U$ are assumed to open their channel when they are granted their resource and to close their channel when the resource is removed.

Each of the processes $i \in U$ have $|\text{resource}_t(i)| < 1$, so as soon as any process $i \in U$ is granted one resource, process $i$ does not compete for any
other, until \( i \) releases that resource.

Given bounds \( v \) on the processes \( i \in U \) competing for the same \( p(j) \), \( j \in R \) and a bound \( k < v \) on the number of resources a user is willing to be granted at any time \( t \), a real-time RGS will imply a real-time handshake communication scheduler, with the same time performance (see the Appendix).

It is interesting to note (as we show in Section 3) that one can implement also a real time RGS by a real time distributed communication subsystem.

3. **HIGH LEVEL RGS IMPLEMENTATION ASSUMING A REAL-TIME HANDSHAKE COMMUNICATION SYSTEM**

3.1 **The Algorithm**

Our implementation of a probabilistic real time RGS is as follows:

The granting processes \( i \) are always willing to communicate only to the requesting processes in the set \( S_i \) which have rights to resources \( p(i) \) (as defined in the Introduction). The requesting processes are willing to communicate only to those granting processes whose resources they want (or have been allocated). By communication here we mean a handshake communication. We shall assume here the existence of a DCS with \( \epsilon \)-error response \( T(\epsilon) \), as in the Appendix.

We may view the actions of the requesting processes time-sliced in rounds, where each round is a minimal time interval in which \( i \) communicates at least once with all \( k \) of the resource controlling processes of the resources which \( i \) wishes to obtain.

The granting processes do forever the following grant algorithm which is a loop, a single execution of which is called a grant phase:
Grant Algorithm
for process $i \in \mathcal{R}$

Do forever

begin

[1]. Do a handshake communication with anyone of the requesting processes in $S_i$ and get their priorities (in $\delta$ steps).

[2]. Probabilistically select $j \in S_i$, in $\delta$ steps. (The values of the priorities determine the probability of each requesting process to be selected as determined in Section 3.2.)

[3]. On first handshake with the selected process $j$, say "yes" to $j$ allocating your resource to $j$. ($\delta$ steps)

[4]. For 2 $\delta$ steps, the granting process says "no" to any requesting process but $j$ in any handshake and says "yes" to $j$ on any communication.

[5]. On handshake with any other process than $j$, the granting process $i$ says "no". On first handshake communication with $j$, the granting process $i$ says "no" to $j$, indicating that resource $\rho(i)$ has been withdrawn from $j$ and ending the grant phase. ($\delta$ steps).

[6]. Wait for $[w]$ steps where $w$ is randomly chosen from $[0, 6\delta']$.

end

In the above algorithm we fix parameters $\delta = \lceil \tau(\varepsilon/2v)/\tau_{\min} \rceil$ and $\delta' = \lceil \delta(\tau_{\max}/\tau_{\min}) \rceil$. Hence $\delta$ steps of any tame process contain at least time $\tau(\varepsilon/2v)$ and the maximum time duration of $\delta$ steps is equal to the minimum time duration of $\delta'$ steps. Thus the maximum possible length of the random wait is, at least the time length of the rest of the stages of the grant phase. Note that stages [1], [3] and [5] may take more than $\delta$ steps (with very low likelihood). This is taken into account in the analysis of performance of the algorithm.
3.2 Probabilistic Selection

We give here a very simple implementation of probabilistic selection of one out of \( v \) processes (using their priorities) in \( O(v) \) steps as required in phase 2 of the algorithm in Section 3.1. This implementation can easily be improved to \( O(\log v) \) steps (see [Reif, Spirakis 1982]).

Suppose that each resource allocator \( i \) has just a random number generator drawing uniform numbers between 0 and 1. Let \( \pi_1, \pi_2, \ldots, \pi_v \) (for \( v' \leq v \)) be the processes requesting the resource of \( i \) at the current time and let \( P_{\pi_1}, P_{\pi_2}, \ldots, P_{\pi_v} \) be their priorities. Let \( h(x) = \sum_{j=1}^{x} P_{\pi_j} \) for any \( x \) in \( \{0, 1, 2, \ldots, v'\} \). To implement the selection process of stage [2] we do the following:

[2.1] draw a random number \( n \) in \( [0,1] \).

[2.2] find the process name \( \pi_x \) for which

\[
\frac{h(x-1)}{h(v')} < n < \frac{h(x)}{h(v')}
\]

Note that stage [2.1] takes one step and stage [2.2] takes \( O(v) \) steps since we have to evaluate and compare 2 partial sums each step. Since

\( h(x) = h(x-1) + P_{\pi_x} \) for any \( x \) in \( \{1, \ldots, v'\} \), the current partial sum can be evaluated from the previous partial sum by a single addition.

4. REAL TIME IMPLEMENTATION BY USE OF FLAG VARIABLES

4.1 The Algorithm for the Case of Equi-Speed Processes

For simplicity, we shall temporarily assume here a fixed time instant \( t_0 > 0 \) such that for time \( t > t_0 \) all processes are executing at the same
speed. (Section 4.2 drops the assumption of equi-speed processes.) However, for times $t < t_0$ in the past, the processes may be asynchronous and so, at $t > t_0$ the execution of their programs in time, though proceeding with equal speed for all processes, may be shifted (in an adverse way) relative to each other.

The communication between granting and requesting processes is done here by flag variables. To read one flag requires one of the process steps. In case of priorities, the flags are allowed to have rational values between 0 and 1, initially 0. The flags $P_{ij}$ indicate the priority of user $j$ with respect to resource $i$. In the simple case of equal priorities all flags need only be boolean. Each granting process $i$ has for each requesting process $j$ a special flag $F_{ij}$ whose value indicates if the resource $p(i)$ is allocated to $j$. If $j$ reads $F_{ij}$ and finds it 0, then it understands that it lost the resource. The granting processes execute forever the following loop, called a grant phase:

**Grant Algorithm**

of Granting Process $i \in R$

Do forever

begin

[1]. Read the priority flags of the requesting processes in the set $S_i$.

[2]. Probabilistically select each of the requesting processes $j \in S_i$ according to their priorities (see Section 3.2).

[3]. Set the flag $F_{ij}$ to 1 indicating that resource $p(i)$ has been allocated to process $j$.


[5] Set the warning flag $L_{ij}$ to 1 to indicate to $j$ that he will loose the resource after at most $2cv$ steps. Wait $2cv$ steps.

[6] Set $F_{ij}$ to 0 (indicating that $i$ removes the resource). Erase the warning by setting $L_{ij}$ to 0.
Wait for \( w \) steps where \( w \) is a random integer selected uniformly from \([0,5cv]\).

Stages [1], [2] are required to take exactly \( cv \) steps each, where \( c \) is a small constant whose exact value can be determined by counting the maximum number of steps of the granting process per requesting process in the stages [1], [2], [3].

The grant phase takes a random number of steps, uniform in the set \([5cv,5cv+1,...,10cv]\).

Each user process \( j \in U \) executes continuously the following loop, a single execution of which is called a round.

\[
\text{Do forever} \\
\text{begin} \\
[1]. \text{Set } P_{ij} \text{ to 1 for each resource } \rho(i) \text{ requested by user } j \text{ (cv steps)} \\
[2]. \text{Poll for cv steps to see which resources have been awarded to user } j. \text{ User } j \text{ considers the resource } \rho(i) \text{ awarded only if } \rho(i) \text{ has been both allocated (} F_{ij} = 1 \text{) and not yet warned (} L_{ij} = 0 \text{).} \\
[3]. \text{If all resources requested by } j \text{ are awarded use them for } \mu \text{ steps.} \\
\text{end} \\
\]

(See also Figures 1 and 2.)
Figure 1: A phase of a grant process.

Figure 2: A round of a requesting process.
4.2 The Flags Implementation of the RGS for the Case of Asynchronous, Tame Processes

Grant Algorithm

for \( i \in R \), for asynchronous case

It is the same as in the case of equi-speed processes. However, the stage lengths are modified as follows:

- Stage [1]: takes \( c'v \) steps
- Stage [2]: takes \( c'v \) steps
- Stage [3]: 1 step as in equi-speed case
- Stage [4]: takes \( c'v \) steps
- Stage [5]: takes \( 2c'v \) steps
- Stage [7]: choose a random integer \( w \in [0, 5c''v] \)

where

\[
c' = \frac{r_{\text{max}}}{r_{\text{min}}} \cdot c \quad \text{and} \quad c'' = \left( \frac{r_{\text{max}}}{r_{\text{min}}} \right)^2 \cdot c
\]

The round of \( j \in U \) in asynchronous case:

Same actions as in equispeed case. The stage lengths in steps are \( cv \), \( cv \) and

\[
u < (cv-1) \frac{r_{\text{min}}}{r_{\text{max}}}, \text{ for stages [1], [2], [3], respectively.}
\]
5. ANALYSIS OF THE REAL TIME RGS IMPLEMENTATION BY USE OF FLAG VARIABLES

5.1 The Case of Equispeed Processes

Note that in the equispeed case assumed here, the time is a constant multiple of the process steps. The power of the adversary is thus restricted to only a possibly malicious initial relative shift of the program counters of the various processes.

**Lemma 1.1.** For $\mu < cv - 1$, it is impossible for the user $j$ to conclude that it has got all resources and actually some of the resources to have been removed.

**Proof.** Since the polling time of $j$ lasts only $cv$ steps, by the time user $j$ concludes that the last resource is allocated to $j$, the first allocated resource can at most be in the middle of the warning period (and hence not removed yet). Since $\mu < cv - 1$, $j$ has then enough steps at its disposal to use the resources, before the first allocated resource is removed. \(\Box\)

**Lemma 1.2.** The probability that user $j$ will get a particular resource in its current round is $< 1/v$ for the worst case oracles.

**Proof.** Consider the subclass of oracles $\mathcal{W}$ which put maximum contention on the system. These oracles give a worst case of the response time, since contention cannot decrease the response time. However, in this case, the probability that user $j$ will get a particular resource in its current round is at best equal to the probability that $j$ is going to be selected by the resource allocator, so it is not more than $1/v$. \(\Box\)

Recall $t_0$ is the time instant at which the processes became synchronous. Let $t_1$ be any time instant after $t_0$. Let \(\text{resources}_{t_1}(j) = \{\rho_1, \ldots, \rho_k\}\).
where \( k' \leq k \), for a particular requesting process \( j \in R \) and let 
\( i_1, \ldots, i_{k'} \) be the resource allocators associated with those resources.

In the following we consider a time interval \( I \) starting at \( t_1 \) during
which the set resources \( t(j) \) for \( t \in I \) is equal to resources \( t(j) \). Let 
\( \Gamma_{t_1} \) be a complete description of the system's history up to time \( t_1 \)
(including the probabilistic choices of the processes up to that time). Let 
\( t_m \) (where \( 1 \leq m \leq k' \)) be the first time after \( t_1 \) at which the allocator 
\( i_m \) starts a random wait stage. Let \( t_M \) be the maximum of all \( t_m \)'s for 
m = 1, \ldots, k' and let \( t_S \) be the maximum of the time instances at the ends
of the first grant phase of processes \( i_1, \ldots, i_{k'} \) after time \( t_M \). The
time interval \((t_1, t_S]\) is called a session \( \Sigma \) of processes
\( \{j, i_1, i_2, \ldots, i_{k'}, \} \).

Let \( g_j(\Sigma) \) be the probability that the process \( j \) will get all its
resources in the session \( \Sigma \), given any history \( \Gamma_{t_1} \).

Note that a session is at most two grant phases of any resource
allocator, because processes are equispeed. Note also that after \( t_M \) the
effect of the history \( \Gamma_{t_1} \) is completely counteracted by the random waits
done by the processes.

DEFINITION. Let \( g_j(t_M, t_S]\) be the probability that process \( j \) will
get all its resources during the interval \((t_M, t_S]\), given any history \( \Gamma_{t_1} \).
Obviously \( g_j(t_M, t_S]\) \( \leq g_j(\Sigma) \). Note that there is at least one complete
round \( R \) of process \( j \) during a session \( \Sigma \), such that \( R \) starts after
\( t_M \). Let \( E_m \) be the event "the flag \( P_{i_m} \) of user \( j \) is seen by the
resource allocator \( i_m \) in the round \( R \) and during \((t_M, t_S]\) and \( j \) is
selected by \( i_m \)." Then \( g_j(t_M, t_S]\) \( \geq \text{Prob}(n_{m=1}^{k'} E_m) \).
Figure 3: A session $\Sigma$. 

a session $\Sigma$
LEMMA 1.3. The events $E_1, \ldots, E_k$ are independent.

Proof. Fix a time $t \in (t_M, t_S)$. By time $t$, all allocators have executed at least a random independent wait stage. The length of the wait stage is uniformly randomly chosen to take any integer value from 0 up to the length $5cv$ of all the other stages of a grant phase. Thus, at time $t$ any allocator $m$ is at each step of the non-wait part of its grant phase with equal probability. □

COROLLARY 1.3.

$$g_1(t_M, t_S) > \text{Prob}(E_1) \cdot \cdots \cdot \text{Prob}(E_k)$$

Proof. By the independence of events $E_m$, $m = 1, \ldots, k'$ as proven in Lemma 1.3. □

LEMMA 1.4. For any $m \in \{1, \ldots, k'\}$

$$\text{prob}(E_m) \geq \frac{1}{10v}$$

Proof. Let $E'_m$ be the event "the flag $P_{i_m,j}$ of user $j$ is seen by $i_m$ in the round $R$ and during $(t_M, t_S)$." Then $\text{Prob}(E_m \text{ given } E'_m) > 1/v$ since $i_m$ selects one out of at most $v$ processes with equiprobability and $j$ is one of them. Also, $\text{prob}(E'_m) = \text{prob}(\text{the start of } R \text{ falls into the first stage of a grant phase of } i_m \text{ during } (t_M, t_S))$ = ratio of length of first stage divided by total length of the grant phase of $i_m$ (by Lemma 1.3). But the total length of any grant phase is at least $5cv$ and at most $10cv$ and the length of first stage is exactly $c$ hence

$$\text{prob}(E'_m) > \frac{cv}{10cv} = \frac{1}{10}.$$
Also,

\[
\text{prob}(E_m) \geq \text{prob}(E'_m) \cdot \text{prob}(E_m \text{ given } E'_m) \\
\geq \frac{1}{10} \cdot \frac{1}{v}
\]

By Corollary 1.3 then

**COROLLARY 1.4.**

\[
g_i(t_M, t_S) \geq \frac{1}{(10v)^{k'}} \geq \frac{1}{(10v)^k} \quad \text{since } k' \leq k.
\]

**Hence**

\[
g_i(\Sigma) \geq \frac{1}{(10v)^k}
\]

**Also, note that**

\[
g_i(\Sigma) \leq \left( \frac{1}{v} \right)^k
\]

due to Lemma 1.2.

Let \( u \) be the number of sessions required for user \( i \) to be granted all its \( k \) resources in some round, given that \( i \) starts requesting them at time \( t_1 \) and assuming any history \( \Gamma_{t_1} \) and oracle \( \mathcal{A} \). By the Corollary 1.4 and Baye's formula,

\[
\text{Prob}(u=m) \leq \left( 1 - \frac{1}{(10v)^k} \right)^{m-1} \frac{1}{(v)^k}
\]

If \( u(\varepsilon) \) is the least number such that \( \text{prob}(u > u(\varepsilon)) \leq \varepsilon \) then

\[
u(\varepsilon) = \frac{\log(\varepsilon/(10)^k)}{\log \left( 1 - \frac{1}{(10v)^k} \right)}
\]
It is easy to show:

PROPOSITION 1. For every $n \in \mathbb{N}$,

$$\frac{1}{\log \left(1 - \frac{1}{n}\right)} > - n$$

By Proposition 1, then

$$u(\epsilon) \leq k \cdot (10v)^k \log \frac{10}{\epsilon} = O(k^k \log \frac{1}{\epsilon})$$

Since each session takes $\leq 20cv$ steps, we have

$$\text{Prob}\{Y_{k+c} \leq 20cv \cdot u(\epsilon)\} \geq 1 - \epsilon$$

implying $\epsilon$-error response

$$\gamma_k(\epsilon) = O(k^{k+1} \log \frac{1}{\epsilon})$$

and mean

$$\bar{\gamma}_k = O(k^{k+1})$$

COROLLARY 1.5. In the case of equispeed processes, our flag implementation of RGS has real time response, with $\epsilon$-error response

$$\gamma_k(\epsilon) = O(k^{k+1} \log \frac{1}{\epsilon})$$

and mean $\bar{\gamma}_k = O(k^{k+1})$.

5.2 The Case of Asynchronous Tame Processes

Our analysis for the case of asynchronous tame processes parallels that of the equispeed case.

LEMMA 2.1. For $\mu < (cv-1)(r_{\min}/r_{\max})$, it is impossible for the user $j \in U$ to conclude that it has got all resources and actually some of the resources to have been removed.
Proof. The polling time of $i$ can at most be $cv\ r_{\text{max}}$, since it includes just $cv$ steps of $i$. Each resource allocator waits $2c'v$ steps after setting its warning flag to 1 and then it removes the resource. Since $c' = (r_{\text{max}}/r_{\text{min}})^c$, we have that $c'v$ steps of a resource allocator are at least $c'v\ r_{\text{min}}$ time which is at least $cv\ r_{\text{max}}$ time. Hence, by the time $j$ concludes that the last resource has been allocated to him, the allocator for the first resource granted can at most be in the middle of its warning period (in terms of steps). The maximum time corresponding to $\mu$ steps is $\mu\ r_{\text{max}} = (cv-1)r_{\text{min}}$ and the minimum time which corresponds to the remaining half steps of the warning period is at least $c'v\ r_{\text{min}} = cv\ r_{\text{max}} > (cv-1)r_{\text{min}}$. So, $j$ has enough time at his disposal to use the resources, before the first allocated resource is removed.

As in Lemma 1.2, we have:

**Lemma 2.2.** The probability that user $j$ will get a particular resource in its current round is $< 1/v$ for the worst case oracles.

Let $t_1, t_1', t_M, t_s, \Sigma, \{v_1, \ldots, v_k\}, j, I, q_i(\Sigma)$ and $q_i(t_M, t_s)$ just as defined in Section 5.1.

A crucial difference from 5.1 is that now a session is at most two grant phases for at least one allocator and not necessarily for all of them. Again, $g_i(t_M, t_s) \leq g_i(\Sigma)$.

Note also that the minimum time duration of a grant phase is $(5c'v + 5c''v)r_{\text{min}} = 5cv(r_{\text{max}}^2/r_{\text{min}})$ and the maximum time duration of a round of a requesting process is now $(2cv + \mu)r_{\text{max}} \leq (2cv + cv(r_{\text{min}}/r_{\text{max}}))r_{\text{max}} = cv(2r_{\text{max}} + r_{\text{min}})$. This implies that a grant phase of any allocator contains at least 2 rounds of any requesting process and hence there is at least one complete round $R$ of process $j$ during $\Sigma$ and after $t_M$. Again,
LEMMA 2.3. The events \( E_1, \ldots, E_{k'} \) are independent.

Proof. Fix a \( t \in (t_M, t_S) \). Since \( t > t_M \) by time \( t \) all allocators have each executed one (or more) random independent wait stages. The number of steps of a wait stage is a random integer chosen uniformly from 0 to \( 5c''v \), where \( c'' = r_{\max} / r_{\min} \). Hence the minimum time duration of a wait stage is at least \( 5c''v r_{\min} = 5c''(r_{\max} / r_{\min}) \cdot r_{\max} = 5c'r_{\max} \) the maximum time duration of the rest of the stages of the grant phase. So, in any case, the random wait can completely cover the length of the rest of the stages. So at time \( t > t_M \) any allocator \( m \) is at any step \( x_m \) with equal likelihood, independently of other allocators. \( \Box \)

As in Corollary 1.3, we have

COROLLARY 2.3.

\[
g_i(t_M, t_S) \geq \prod_{i=1}^{k'} \Pr(E_i)
\]

LEMMA 2.4. For any \( m \in \{1, \ldots, k'\} \)

\[
\Pr\left( E_m \right) \geq \frac{1}{5 \left( 1 + \frac{r_{\max}}{r_{\min}} \right)} \cdot \frac{r_{\min}}{r_{\max}} \cdot \frac{1}{v}
\]

Proof. Let \( E'_m \) be the event as defined in Lemma 1.4. Again, \( \Pr(E_m / E'_m) \geq 1/v \) and \( \Pr(E'_m) = \text{ratio of length of first stage divided by total length of the grant phase} \).
\[
\frac{c'v r_{\text{min}}}{(5c'v+5c''v)r_{\text{max}}} = \frac{1}{5 \left(1 + \frac{r_{\text{max}}}{r_{\text{min}}} \right)} \cdot \frac{r_{\text{min}}}{r_{\text{max}}}
\]

and since \( \text{Prob}(E_m) \geq \text{Prob}(E'_m) \cdot \text{Prob}(E_m/E'_m) \) the lemma follows.  \( \square \)

Hence, by Corollary 2.3.

**Corollary 2.4.**

\[
\left(\frac{1}{\lambda v}\right)^k \leq g_i(\xi) \leq \left(\frac{1}{v}\right)^k
\]

for

\[
\lambda = 5 \left(1 + \frac{r_{\text{max}}}{r_{\text{min}}} \right) \frac{r_{\text{max}}}{r_{\text{min}}}
\]

Let \( u \) be defined as in case of equispeed processes.

By following same steps as in that case, we get

\[
u(\epsilon) \leq k(\lambda v)^k \log \frac{\lambda}{\epsilon} = O\left(kv^k \log \left(\frac{1}{\epsilon}\right)\right)
\]

A session is at most two grant phases of at least one allocator. Therefore the time length of a session is bounded by \( (5c'v+5c''v)r_{\text{max}} \), so we have

\[
\text{Prob}\{\gamma_k \leq 5v(c' + c'')r_{\text{max}} \cdot u(\epsilon)\} \geq 1 - \epsilon
\]

**Corollary 2.5.** Our flag implementation of RGS in the general case of tame asynchronous processes, has real time response with

\[
\gamma_k(\epsilon) = O\left(kv^{k+1} \log \left(\frac{1}{\epsilon}\right)\right)
\]

and

\[
\tilde{\gamma}_k = O(kv^{k+1})
\]
5.3 Analysis of our RGS which Uses a Handshake Communication System.

By the stated properties of DCS,

**Proposition 3.1.**

\[
\text{Prob} \left( \text{a round length is at most } t \left( \frac{L}{2v} \right) \right) \geq \left( 1 - \frac{L}{2v} \right)^v
\]

**Lemma 3.2.** The probability that a particular requesting process \( i \) will get a "yes" answer in a handshake in stage 3 is \( \leq 1/v \) for the worst case oracles.

**Proof.** Consider again the class of oracles \( \mathcal{O} \) putting maximum contention in the system.

Let \( t_1, t_M, t_S, t_1, \Sigma, g_1(\Sigma) \) and \( g_1(t_M, t_S) \) be just as defined in Section 5.1. Again, \( g_1(t_M, t_S) \leq g_1(\Sigma) \).

Let \( E_m \) be now the event "a requesting process \( j \) gets a "yes" answer in a handshake with allocator \( i_m \) in the interval \( (t_M, t_S) \) and in the first round \( R \) of \( j \) after \( t_M \)." Note that \( g_1(t_M, t_S) \geq \text{Prob}(\cap_{m=1}^{k'} E_m) \).

We show again that the events \( E_m \) are statistically independent, as in the Proof of 1.3. Again we observe that the length of a random wait has a nonzero probability of completely covering the length of the rest of the stages in a grant phase. The length of a random wait is randomly chosen from \( [0, 6\delta'] \) so its maximum duration is at least \( 6\delta' r_{\min} = 6\delta' r_{\min} = 6\delta' r_{\max} = \text{the maximum time duration of the rest of the grant phase stages}. \) So, Lemma 1.3 holds again, hence

**Corollary 3.3.**

\[
g_1(t_M, t_S) \leq \prod_{m=1}^{k'} \text{Prob}(E_m)
\]
LEMMA 3.3. For any \( m \in \{1, \ldots, k'\} \)

\[
\text{Prob}(E_m) \geq \frac{1}{6(1 + \frac{r_{\text{max}}}{r_{\text{min}}})} \frac{r_{\text{min}}}{r_{\text{max}}} \frac{1}{v}.
\]

Proof. Let \( E'_m \) be the event "the handshake of processes \( j \) and \( i_m \) will take place at the first stage of the grant phase of \( i_m \), in the first round \( R \) of \( j \), in the interval \( [t_M', t_S'] \)." Then \( \text{Prob}(E_m | E'_m) \geq 1/v \) due to probabilistic selection of requesting processes. Also \( \text{prob}(E'_m) = \text{prob}(E) \) (the start of \( R \) falls into the first stage of a grant phase of \( i_m \) during time interval \( (t_M', t_S') = \text{ratio of length of the first stage divided by total length of the grant phase}, \) because by \( t_M \) process \( i_m \) has executed at least one random wait (after \( t_1 \)) and hence the start of \( R \) will be any time instant of a grant phase of \( i_m \) with equal probability, independent of the history \( i_{t_1} \). Hence,

\[
\text{prob}(E'_m) \geq \frac{\text{min length of stage } 1}{\text{max length of grant phase}} \geq \frac{\delta r_{\text{min}}}{(6\delta + 6\delta') r_{\text{max}}} \geq \frac{1}{6(1 + \frac{r_{\text{max}}}{r_{\text{min}}})} \frac{r_{\text{min}}}{r_{\text{max}}} \frac{1}{v}.
\]

So, \( \text{prob}(E_m) \geq \text{Prob}(E'_m) \cdot \text{prob}(E | E'_m) \)

\[
\geq \frac{1}{6(1 + \frac{r_{\text{max}}}{r_{\text{min}}})} \frac{r_{\text{min}}}{r_{\text{max}}} \frac{1}{v}.
\]

Due to Corollary 3.3, Lemma 3.3 and Lemma 3.2 we get

COROLLARY 3.4.

\[
\left(\frac{1}{\lambda v}\right)^k \leq g_1(\Sigma) \leq \left(\frac{1}{v}\right)^k.
\]
Let $u$ be the number of sessions after $t_1$ needed for process $i$ to get all $k'$ resources in one round, given any history $\Gamma_{t_1}$.

The previous corollary implies
\[
\text{prob}(u=m) \leq \left(1 - \frac{1}{(\lambda'v)^k}\right)^{m-1} \left(\frac{1}{v}\right)^k.
\]

For $\epsilon \in (0,1)$, let $u(\epsilon)$ be the minimum value such that $\text{prob}(u > u(\epsilon)) < \epsilon$.

Then
\[
u(\epsilon) = \frac{\log(\epsilon/(\lambda'v)^k)}{\log\left(1 - \frac{1}{(\lambda'v)^k}\right)}
\]
and again (as in Section 5.2)
\[
u(\epsilon) \leq k(\lambda'v)^k \log\left(\lambda'\epsilon\right).
\]

Note that a session $\Sigma$ contains at most two grant phases of at least one resource allocator. So, the time length of a session $\Sigma$ is at most
\[
2(6\delta + 6\delta')r_{\max} = 12 \left(\tau\left(\frac{\epsilon}{2v}\right) + \tau\left(\frac{\epsilon}{2v}\right) \frac{r_{\max}}{r_{\min}}\right) r_{\max}
\]
\[
= 12 \left(1 + \frac{r_{\max}}{r_{\min}}\right) r_{\max} \cdot \tau\left(\frac{\epsilon}{2v}\right).
\]

Note that at least $\tau\left(\frac{\epsilon}{2v}\right) = \delta$ steps are in $(t_M,t_S)$. Hence the probability that at least one complete round $R$ is contained in $(t_M,t_S)$ is at least $\left(1 - \frac{\epsilon}{2v}\right)^v$. Now
\[ \text{Prob}\{Y_{k,n} \leq 12 \left(1 + \frac{r_{\text{max}}}{r_{\text{min}}}\right) r_{\text{max}} \cdot \tau \left(\frac{\epsilon}{2v}\right) \cdot u\left(\frac{\epsilon}{2}\right)\} \]

\[ \geq \text{Prob}\{\text{each session contains at least one complete round } R \text{ in } (t_M, t_S) \text{ and } u \leq u(\epsilon/2)\} \]

\[ \geq \left(1 - \frac{\epsilon}{2v}\right)^v \left(1 - \frac{\epsilon}{2}\right) \]

because the "length of a round" distribution is determined by the underlying DCS implementation independently of the number of rounds needed for a process to get all its \( k \) resources. So

\[ \text{Prob}\{Y_{k,n} \leq 12 \left(1 + \frac{r_{\text{max}}}{r_{\text{min}}}\right) r_{\text{max}} \cdot \tau \left(\frac{\epsilon}{2v}\right) \cdot u\left(\frac{\epsilon}{2}\right)\} \]

\[ \geq \left(1 - \frac{\epsilon}{2}\right)^2 \geq 1 - \epsilon \]

implying

\[ \gamma_k(\epsilon) = O\left(\tau\left(\frac{\epsilon}{2v}\right) u\left(\frac{\epsilon}{2}\right)\right) = O\left(kv^k \log\left(\frac{1}{\epsilon}\right) \tau\left(\frac{\epsilon}{2v}\right)\right) \]
which has mean
\[ \gamma_k = O(kv^k \bar{\tau} \log v) \]

Using \( \tau(\epsilon) = O(v^2 \log(1/\epsilon)) \) and \( \bar{\tau} = O(v^2) \) as given in [Reif, Spirakis, 1981] we get
\[ \gamma_k(\epsilon) = O(v^{k+2} \log \left( \frac{1}{\epsilon} \right) \log \left( \frac{v}{\epsilon} \right)) \]
and
\[ \gamma_k = O(kv^{k+2} \log v) \]

Note that our response for the flag implementation is slightly better than those based on a handshake communication system, since there is no uncertainty about flag communication in each round. However, when processes are not tame, the correctness of the implementation by flags may be violated, while the correctness of the implementation by an underlying DCS will be preserved (because of the handshake communication which accompanies allocation or deallocation of a resource) given that the correctness of the underlying DCS is not violated when processes are not tame.
REFERENCES


APPENDIX

Distributed (Handshake) Communication Systems (DCS)

Suppose that each process has a special resource called channel which can be in one of two states open, closed. A handshake of two processes \( i, j \) in time \( t \) is a combination of processes states at time \( t \) so that both channels of \( i \) and \( j \) are open at the same time.

Successful direct communication is a handshake of at least 1 step overlap of both processes so that the handshake relation is a matching. At any instant \( t \) no process is allowed to be handshaking with more than one other process. During the one step overlap, a message can be transmitted from one process to the other. The problem is usually to synchronize processes (via a distributed scheduler) so that they can handshake at their will, given that the means of synchronization is some low level construct (a message system, buffered communication, shared variables or flags) which does not guarantee the handshake property if used in an unsophisticated way. A distributed scheduler is called real time if it has the property that if two processes \( i, j \) are willing to handshake mutually for at least a constant time interval, then they will actually achieve successful direct communication during that time interval with arbitrarily small probability of error.

Formally, let \( T(\epsilon) \) be the smallest real number such that if two processes \( i, j \) are mutually willing to handshake for at least \( T(\epsilon) \) time, then they will actually succeed in 1 step overlap of open channels during that time, with probability \( \geq 1-\epsilon \). \( T(\epsilon) \) is called the \( \epsilon \)-error.
response of the handshake algorithm. The mean response $\bar{\tau}$ of a handshake algorithm is the maximum (over all adverse speed schedules of tame processes and overall adverse communication requests subject to restrictions stated in the Introduction) of the mean time needed for two processes to handshake, from the time instant they start to be mutually willing. A real time probabilistic scheduler has $\tau(\varepsilon)$ depending only on $v$ and not on any other global measure of the communications graph. ($v$ is a fixed upper bound on the out-valence of the dynamic communication willingness digraph at any time instant $t$). We also require $\tau(\varepsilon)$ to increase at most linearly with $1/\varepsilon$. Note that such a scheduler has $\bar{\tau}$ also depending only on $v$.

The handshake problem has been given some attention in literature [Schwarz, 79], [Francez, Rodeh 80], [Francez, 81], [Reif, Spirakis 81], and [Reif, Spirakis 82A].

For Section 3 we require a Distributed Communication System (DCS) as defined above with a distributed real time probabilistic scheduler. We also require the DCS to have the following property:

If a process $i$ is willing to communicate with $k \leq v$ processes for at least time $\geq \tau(\varepsilon)$ and if they are also willing to (handshake) communicate with $i$ during that interval, then the probability that $i$ will be able to communicate with all of them (in some order) within $\tau(\varepsilon)$, is $\geq (1-\varepsilon)^v$. Such a real time DCS was implemented in [Reif, Spirakis 81] with $\varepsilon$-error response

$$\tau(\varepsilon) = O(v^2 \log(1/\varepsilon))$$

and mean

$$\bar{\tau} = O(v^2)$$.
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