Applying new scheduling theory to static priority pre-emptive scheduling

by N. Audsley, A. Burns, M. Richardson, K. Tindell and A.J. Wellings

The paper presents exact schedulability analyses for real-time systems scheduled at runtime with a static priority pre-emptive dispatcher. The tasks to be scheduled are allowed to experience internal blocking (from other tasks which they share resources) and (with certain restrictions) to release jitter, such as waiting for a message to arrive. The analysis presented is more general than that previously published and subsumes, for example, techniques based on the Rate Monotonic approach. In addition to presenting the relevant theory, an existing avionics case study is described and analysed. The predictions that follow from this analysis are seen to be in close agreement with the behaviour exhibited during simulation studies.

1 Introduction

One proposed method of building a hard real-time system is from a number of periodic and sporadic tasks, and a common way of scheduling such tasks is by using a static priority pre-emptive scheduler; at runtime the highest priority runnable task is executed, pre-empting other lower priority tasks. This scheme was employed in the Rate Monotonic approach defined by Liu and Layland [1], where the (unique) static priority of a task is obtained from the period of that task; for any two tasks L and M, period(L) > period(M) ⇒ priority(L) < priority(M). Liu and Layland derived a schedulability analysis which determines if a given task set will always meet all deadlines under all possible release conditions. The schedulability test given is sufficient (i.e. all task sets passing the test are guaranteed to be schedulable), but not necessary (i.e. a task set failing to pass the test is not necessarily unschedulable). Sha et al. [2] provided an exact Rate Monotonic test that is both sufficient and necessary.

The original Rate Monotonic approach had several restrictions:

- all tasks are periodic,
- no task can block waiting for an external event,
- all tasks share a common release time (called the critical instant),
- all tasks have a deadline equal to their period.

The restriction that tasks cannot interact has been removed by the Priority Ceiling protocol [3] (and other similar protocols such as the Stack Resource Protocol [4] for data/resource sharing). A method which allows sporadic tasks to be accommodated using periodic servers has been proposed by Lehoczky et al. [5] (analysis is provided which can guarantee the worst-case response time of a single sporadic task). Rajkuman [6] used external blocking (i.e. when a task is blocked awaiting an external event, such as a delay expiry) with the Rate Monotonic approach to model the operation of a multi-processor Priority Ceiling protocol, and provided schedulability analysis to bound its effects. The restriction that tasks are assumed to share a critical instant has been relaxed by Audsley [7].

The final restriction that the deadline of a task must be equal to the period has not been relaxed for the Rate Monotonic approach. This is perhaps the most important restriction to lift; requiring the deadline of a task to be less than the period of that task is essential if jitter requirements are to be met (i.e. the result of a piece of computation must be produced within precise intervals). Furthermore, when building distributed systems, the deadline of a task often needs to be shortened to allow time for communication between tasks on different processors. In general, hard sporadic tasks have deadlines that are not related to their minimum inter-arrival time, and hence they cannot be modelled as simple periodic tasks with period equal to deadline.

For tasks with deadlines less than (or equal to) periods, Leung and Whitehead [8] showed that Deadline Monotonic priority assignment is optimal*. Task priorities are now assigned in inverse order to task deadlines; a task with a short deadline (measured relative to the release time of the task) should have a high priority. A task with a long period

* Optimal in the sense that, if a task set can be scheduled by any static priority algorithm, it can also be scheduled by the Deadline Monotonic algorithm.
but short deadline would have a low priority according to the Rate Monotonic priority assignment, but a high priority according to the Deadline Monotonic priority assignment. Consequently, the Rate Monotonic assignment is suboptimal for such task sets. If two or more tasks have the same deadline, they are assigned an arbitrary priority order (among themselves).

To apply the Deadline Monotonic approach, scheduling tests must be available that will allow deadlines to be guaranteed. Such analysis is provided by Joseph and Pandya [9], Lehoczyk [10], and Audsley et al. [11]. All provide sufficient and necessary schedulability tests, differing only in the complexity of their computations. The basic approach is expanded by Audsley et al. to permit sporadic task deadlines to be guaranteed without the use of the servers required by the Rate Monotonic approach [8]. It should be pointed out that Audsley et al. and Joseph and Pandya provide schedulability tests for a task set with any arbitrary priority ordering (i.e. they do not just apply to task sets with priorities ordered by the Deadline Monotonic scheme). They also have the useful property that they furnish estimations of the actual worst-case response times for each task. The actual schedulability test is then a trivial comparison of each task’s response time and deadline. The calculation of response time is particularly important when deadline requirements are assigned to the behaviour of a collection of tasks (in a distributed system, for example). No single task has a hard ‘deadline’, but each task’s response time contributes to some system-wide timing requirement that must be satisfied.

In this paper, we are concerned with providing schedulability analysis to predict the worst-case response times for a set of periodic and sporadic tasks under any given priority assignment, and scheduled by a static priority preemptive scheduler.

2 Computational model

In this paper, unless explicitly mentioned, we consider the scheduling of tasks on a single processor. The techniques can also be used in a distributed environment with static task allocation [12].

A task \( i \) is assumed to consist of an infinite number of invocation requests, each separated by a minimum time \( T_i \). Each invocation is a request to perform a bounded amount of computation \( C_i \), and to lock and unlock semaphores from a bounded set \( S(i) \) according to the Priority Ceiling Protocol [3]. Some tasks have a deadline requirement such that all computation for an invocation must take place before a certain time measured relative to the invocation request. Deadlines, where required, are assumed to be constant and known \textit{a priori}. The deadline requirement of a task \( i \) is denoted \( D_i \).

The notion of a task at time \( t \) is subsequently recognized by the runtime dispatcher, and the task is placed in a notational queue of runnable tasks. The task is then said to be released. The variability in time between a task’s arrival and its release is known as release jitter. In Section 3, this is assumed to be zero. This is the assumption normally applied in scheduling theory (e.g. in the Rate Monotonic approach).

A task has a static base priority assigned to it \textit{a priori} (using the Deadline Monotonic priority assignment algorithm, for example). It may also inherit a higher dynamic priority due to the operation of the Priority Ceiling Protocol. The dispatcher chooses to run the highest dynamic priority runnable task, pre-empting lower priority tasks when necessary.

3 Finding worst-case response times

Before proceeding further, we introduce the following simple notation. All internal blocking is assumed to be the result of semaphore use (other synchronisation primitives could also be used and analysed [13]).

\[ C_i = \text{the worst-case computation time required by task } i \text{ on each release.} \]

\[ T_i = \text{the lower bound on the time between successive arrivals of } i. \]

\[ D_i = \text{the deadline requirement (if defined) of task } i, \text{ measured relative to a given release of } i. \]

\[ B_i = \text{the worst-case blocking time task } i \text{ can experience due to the operation of the priority ceiling protocol (or equivalent concurrency control protocol).} \]

\[ J_i = \text{the worst-case interference task } i \text{ can experience.} \]

Interference on \( i \) is defined as the time higher priority tasks can pre-empt and execute, and hence prevent \( i \) from executing.

Software Engineering Journal  September 1993  285
\( R_i \) = the worst-case response time for a task \( i \) measured from the time the task is released. For a schedulable task, \( R_i \leq D_i \) (if there is no deadline requirement for \( i \), the analysis below must have \( R_i \leq T_j \)).

\( hp(i) \) = the set of tasks of higher base priority than the base priority of \( i \) (these tasks could pre-empt \( i \)).

We now consider the problem of computing the worst-case response time for a task \( i \), denoted \( R_i \). Initially, tasks are assumed to be released when they arrive. This time can be viewed as a computational "window" (or busy period); the release of \( i \) marks the start of the window, and the completion of \( i \) marks the end of the window. The maximum width of the window is \( R_i \). In this window of duration \( R_i \), task \( i \) must (at worst) complete an amount of computation equal to \( C_i \) and be delayed when locking semaphores by \( B_i \) at most. Additionally, task \( i \) could be preempted by \( I_i \) at most. Therefore, we can say that:

\[
R_i = C_i + B_i + I_i \tag{1}
\]

If a task \( i \) has a deadline, then we must have \( R_i \leq D_i \).

The worst-case computation time \( C_i \) is constant and is somehow known a priori [14, 15]. The worst-case blocking time \( B_i \) is found according to the analysis given by Sha et al. [3], and is equal to the longest critical section of any lower priority task accessing a semaphore with ceiling of equal or higher priority than task \( i \).

In the rest of this Section, we present the analysis to find \( I_i \). Note that similar analysis, cast in a different form, was first produced by Joseph and Pandya, [9] and later by Audsley et al. [11]. The method described below has the advantage of being easily extended to cover situations such as nonzero release jitter time. Note also that the analysis is not based on the notion of processor utilization. Although process sets with deadlines equal to periods can be assessed according to their utilization, such techniques are not general-purpose. For example, two tasks with deadlines equal to their computation time will never be schedulable, regardless of processor utilization.

To find a formula for the interference, consider the sequence of computations illustrated in Fig. 1. Fig. 1 was produced by a tool called STRESS [16], written by the Real Time Systems Research Group at the University of York; the Appendix describes the notation used in these diagrams. Fig. 1 shows part of a schedule of a system consisting of three tasks, displayed in priority order. Task 1 is a task with worst-case computational requirement of \( C_1 = 1 \), a deadline of \( D_1 = 4 \) and a worst-case inter-arrival time of \( T_1 = 50 \). The characteristics of Tasks 2 and 3 are defined in Table 1. Fig. 1 shows the worst-case scheduling point described by Liu and Layland, where all tasks are released simultaneously (at notational time 0).

As can be seen from Fig. 1, task 3 is prevented from executing by task 1 for 1 tick and by task 2 for 2 ticks, completing by time 8, giving \( R_3 = 8 \). Task 2 can never pre-empt task 3 more than once as task 3 finishes before tasks 2 can re-arrive (i.e. \( R_3 < T_2 \)).

If task 3 took a little longer to complete (because task 1 executes for an extra two ticks, for example), then a first guess at \( R_3 \) would be \( 8 + 2 = 10 \). However, now that task 3 is a little longer, task 2 can re-arrive and pre-empt task 3 a second time, giving a worst-case interference of 4 from task 2. Fig. 2 shows this situation. As can be seen, the worst-case response time of task 3 is now 12, just missing its deadline.

In general, given prior knowledge of the worst-case response time \( R_i \), the interference on task \( i \) from a task \( j \) is \( nC_j \), where \( n \) has a value such that \((n - 1)T_j < R_i \leq nT_j \).

As \( x(n) = n \) when \( n - 1 < x \leq n \), we can say that the worst-case interference from a task \( j \) on task \( i \) is given by:

\[
I_i = \sum_{j \in hp(i)} \left[ \frac{R_j}{T_j} \right] C_j \tag{2}
\]

Note that this value for the maximal interference holds regardless of whether \( j \) is periodic or sporadic. This is an important result as it means that runtime techniques, such as aperiodic servers [5], are not needed. In fact, a periodic task can be regarded as a sporadic task, released by a regular timing event.

The total interference \( I_i \) is given by:

\[
I_i = \sum_{j \in hp(i)} \left[ \frac{R_j}{T_j} \right] C_j \tag{2}
\]

where \( hp(i) \) is the set of tasks with higher base priorities.

<table>
<thead>
<tr>
<th>task</th>
<th>C</th>
<th>R</th>
<th>D</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>1</td>
<td>50</td>
<td>4</td>
</tr>
<tr>
<td>2</td>
<td>2</td>
<td>9</td>
<td>6</td>
</tr>
<tr>
<td>3</td>
<td>2</td>
<td>20</td>
<td>12</td>
</tr>
</tbody>
</table>

Software Engineering Journal  September 1993
Unfortunately, when eqns. 1 and 2 are combined, the unknown term \( R_i \) appears on both the left- and the right-hand sides of the equation:

\[
R_i = C_i + B_i + \sum_{j \neq i} (\bar{R}_j / T_j) C_j
\]

It is possible to solve this equation using an iterative technique. Let \( \bar{R}_i \) be the nth approximation to the true value of \( R_i \). These approximations are generated from the above equation:

\[
\bar{R}_i^{n+1} = C_i + B_i + \sum_{j \neq i} (\bar{R}_j / T_j) C_j
\]

The iteration starts with \( \bar{R}_i^0 = 0 \) and terminates when \( \bar{R}_i^{n+1} = \bar{R}_i^n \), and so the iteration can be halted early if either \( \bar{R}_i^{n+1} > D_i \) or if \( \bar{R}_i^{n+1} > T_i \). It can also be shown that the iteration is guaranteed to converge if the processor utilisation is <100% [9]. Note that if the priority of task \( a \) is greater than the priority of task \( b \), then \( R_a > R_b \); thus, the task set should be analysed in priority order, with the starting value \( R_a^0 \) set to \( R_a \). This enables the test to be evaluated more quickly.

In the above analysis, we have not made use of any information about priority assignment. Both the Rate Monotonic and Deadline Monotonic policies could be used. In more complex situations, for example in a distributed system with complex trade-offs, finding an optimal priority ordering may be NP-Hard and other sub-optimal techniques such as Simulated Annealing [17] are appropriate [12].

4 The release jitter problem

In this Section, we show how release jitter causes problems with the analysis presented above. We also show how the presented analysis can be extended to allow for such external blocking (and indicate how this type of blocking is often encountered in real systems).

The release jitter problem arises when we change the assumption that a task is always released as soon as it arrives. With release jitter, a task may be released at any time up to a bounded time \( J_i \) after it arrives. This can occur if, for example, the scheduler mechanism takes a bounded time to recognise the arrival of a task.

The analysis presented above is not sufficient when tasks can experience release jitter. Consider the task set defined in Table 2.

Task \( T_1 \) is of higher priority than task \( T_2 \). In this example, we are concerned with the schedulability of \( T_2 \). \( T_1 \) experiences an external block because it needs a message before it can commence, for example. The message is sent at the same time as \( T_1 \) arrives (\( T_1 \) could be a sporadic task, for example, with the arrival triggered by an external event which also triggers the sending of a message from another processor). The message is guaranteed to arrive no later than 4 ticks after the arrival of \( T_1 \), and hence we have a release jitter of \( J_1 = 4 \).

Using our current analysis, we have (i.e. ignoring release jitter)

\[
\bar{r}_2 = 0
\]

\[
\bar{r}_2 = C_2 + \left( \frac{\bar{r}_1}{T_2} \right) C_1 = 6
\]

\[
\bar{r}_2 = C_2 + \left( \frac{\bar{r}_1}{T_2} \right) C_1 = 9
\]

\[
\bar{r}_2 = C_2 + \left( \frac{\bar{r}_1}{T_2} \right) C_1 = 9
\]

The equation has converged, and hence \( \bar{r}_2 = 9 \).

As \( \bar{r}_2 < D_2 \), \( T_2 \) would be deemed schedulable. Fig. 3 shows a schedule for the two tasks when both are released together (Li and Layland's worst-case). However, when release jitter is taken into account, there are situations when \( T_2 \) is not always schedulable. Fig. 4 shows such a situation.

Although \( T_1 \) arrives at time zero, it is suspended awaiting a message, which it receives at time 4 (this is also the time \( T_2 \) arrives and is released). On the next release of \( T_1 \), 12 ticks later, the next message is already available, and so the task can be released immediately. \( T_2 \) misses a deadline (indicated by the black circle in Fig. 4) because of the
release jitter of T1. The reason is that the worst-case scheduling point no longer occurs at Liu and Layland's critical instant (where all tasks are released together), but at the point when T2 is released at the same time as T1 finishes waiting. T1 can then effectively re-occur in a shorter time than the current analysis allows for, and so inflict a 'back to back hit'.

This phenomenon is described by Rajkumar[6], with reference to external blocking when locking remote semaphores in a distributed system; Rajkumar refers to this as an invasive effect due to deferred execution. This extra 'hit' can amount to an additional interference of Cj at most. The current analysis fails because the interference factor Int is not sufficient. An upper bound on the interference to allow for the extra 'hit' might therefore be obtained by simple adding in an extra computation time

\[ I = \sum_{i \in \mathcal{P}(j)} \left( \frac{T_i}{T_j} + 1 \right) C_i \]

In effect, we are saying here that an extra 'hit' occurs if \( r_i + T_i > T_j \). This is pessimistic, as the extra 'hit' is not certain to occur in all systems. Consider Fig. 4 again. If \( C_2 \) was 5 ticks, then \( r_2 \) would be 8, and all computation for T2 would be complete before T1 re-arrived and preempted hence

\[ I = \sum_{i \in \mathcal{P}(j)} \left( \frac{r_i + J_i}{T_j} \right) C_i \]  

(4)

In Fig. 4, \( r_2 \) would be 9, according to eqn. 3. As \( r_2 + J_2 > T_2 \) (or 9 + 4 > 12), T2 gets an extra 'hit'. But if \( C_2 \) = 5, then \( r_2 \) would be 8, and as 8 + 4 ≤ 12, no extra 'hit' occurs.

Eqn. 3 can thus be modified to allow for release jitter:

\[ r_i^{-1} = C_i + B_i + \sum_{i \in \mathcal{P}(j)} \left( \frac{r_i + J_i}{T_j} \right) C_i \]  

(5)

Recall that \( r_i \) is the worst-case response time measured from the point at which task \( i \) is released. A more reasonable and useful measure might be from the time task \( i \) arrives, so that the worst-case time from arrival to completion of task \( i \) is given by

\[ J_i + r_i \]  

(6)

Note that eqn. 5 still allows semaphores to be locked and unlocked according to the Priority Ceiling Protocol.

Having extended the scheduling analysis to handle release jitter, we now indicate how this can occur in a system, using two examples.

4.1 Precedence-constrained distributed tasks

A common method of representing computations in a distributed system is as a collection of tasks with precedence relationships between their executions. Each task is statically allocated to a single processor. Such task sets can be analysed with theory which assumes release jitter. All tasks are defined to arrive at the same time, but a precedence-constrained task on one processor can have its release delayed awaiting an indication of termination of all direct predecessors on other processors (perhaps by the arrival of a message, in a similar way to the earlier example). The worst-case release jitter of such a subtask can be computed by knowing the worst-case response times of predecessor subtasks located on other processors, and by knowing the worst-case communications delay. By assuming a best-case response time of zero for the predecessors, and that best-case message transit times are zero, the release jitter (i.e. the variability in release) can be said...
to be the largest sum of the worst-case response time of each predecessor, computed by 
\[ J_i = \max_{s \in \text{dpred}(i)} (d_i + r_s + M_s) \]
where \( \text{dpred}(i) \) is the set of all tasks which are direct predecessors of task \( i \), and \( M_s \) is the worst-case transit time of the message sent from task \( i \) to task \( s \).

Note that the above equation only holds if all the predecessors of task \( i \) are on a different processor from task \( i \); to allow predecessors to be on the same processor, other analysis must be developed. For example, one approach is to assign a lower priority to task \( i \) than the local predecessors to ensure that task \( i \) never runs before a predecessor, and to assign a release jitter of task \( i \) such that it is greater than or equal to the release jitter of local direct predecessors (so that whenever a higher priority predecessor is deferred awaiting a message arrival, task \( i \) is also deferred, and hence prevented from running).

A more detailed analysis of distributed precedence-constrained tasks is beyond the scope of this paper and is the subject of current research.

4.2 Tick-driven scheduling

The implementation of a priority scheduler can also introduce release jitter. Consider a single processor where periodic and sporadic tasks are scheduled by a scheduler which is invoked by a periodic clock interrupt, the so-called tick-driven scheduling.

Assume the period of the scheduler is \( T_{\text{sys}} \) and that the scheduler, once invoked, takes no more than \( C_{\text{sys}} \) computation time. Consider the following sequence of events; the scheduler is released at time \( t = 0 \) and looks to see if the sporadic task \( s \) is to be released (in a real tick-driven system the scheduler might poll an I/O register for the condition for the release of \( s \)). Assume the condition for the arrival is not true and the scheduler continues executing (ultimately terminating after taking time \( C_{\text{sys}} \)). Just after the time the scheduler has polled, the sporadic \( s \) arrives (i.e. the condition becomes true). However, \( s \) cannot be released until the scheduler is next invoked at time \( t = T_{\text{sys}} \). Hence, the sporadic task is deferred for a maximum time \( T_{\text{sys}} \), awaiting the timer which invokes the scheduler. Fig. 5 illustrates how a sporadic task is deferred by a tick-driven scheduler.

The tick-driven scheduler executes for \( C_{\text{sys}} = 1 \), with \( T_{\text{sys}} = T \). The worst-case execution time of the sporadic task is \( 3 \) time units. As can be seen, the sporadic task is deferred for \( 7 \) time units.

Tasks that always arrive as the scheduler is released do not experience external blocking. In the case study described later, all periodic tasks have periods which are exact multiples of \( T_{\text{sys}} \), with release times measured in scheduler ticks, and hence these tasks can be considered to always arrive as the scheduler is released. However, a periodic task experiences release jitter if its period is not an integer multiple of the clock period.

5 Sporadically repeating tasks

Another illustration of the strength of our analytical approach is to adapt the scheduling analysis to more accurately describe the behaviour of so-called sporadically repeating tasks. Very often a task arrives at a particular time, executes and then re-arrives periodically a fixed number of times. This behaviour is then repeated sporadically during the execution of the system (Fig. 6).

The task illustrated has an 'inner' period of \( 4 \) ticks, a minimum 'outer' period of \( 15 \) ticks and a worst-case execution time of \( 1 \) tick. The task arrives periodically \( 3 \) times for each outer arrival. In Fig. 6 these occur at times \( 0, 15 \) and \( 40 \). This behaviour is quite common in real systems: a task is initiated in response to a particular event, and then for a short period of time periodically monitors or controls a part of the system.

The model also caters for bursty sporadic tasks. An interrupt that releases a sporadic task may be defined as having a very short minimum arrival time, but have a maximum number of arrivals over a larger interval; the...
maximum, being much lower than the minimum, interval would dictate. For example, in a satellite control system (to which this scheduling model has been applied [18]) bus interrupts can occur every 960 µs, but only 4 such interrupts can occur every 10 ms.

If the analysis developed so far is applied to these situations, the predictions would be pessimistic as the theory has to assume that the task executed continually. This might result in a higher assumed interference than could actually occur. However, the general analytical approach is well suited to extending the current analysis to remove this pessimism.

Our general approach to ascertaining the schedulability of a task is to determine the interference over a given window (usually the worst-case response time of a task). This interference is summed, and the window widened if necessary. We require that a wider window always leads to a higher interference. Hence, to ascertain the schedulability in the presence of higher priority sporadically periodic tasks, we need to find an upper bound on the interference over a window of size \( r_j \). We adopt the following additional notation:

- \( n_j \) = the number of times task \( j \) executes for each 'outer' arrival (in Fig. 6, \( n = 3 \)).
- \( t_j \) = the 'inner' period of task \( j \) (in Fig. 6, \( t = 4 \)).
- \( T_j \) = the 'outer' period of task \( j \) (in Fig. 6, \( T = 15 \)).
- \( C_j \) = the worst-case computation time required by the 'inner' task (in Fig. 6, \( C = 1 \)).

For the moment, we assume that tasks do not experience release jitter. The number of full outer periods completing within the window of size \( r_j \) is bounded by

\[
\left\lfloor \frac{r_j}{T_j} \right\rfloor
\]

The total interference due to full outer arrivals is therefore bounded by

\[
n_j \left\lfloor \frac{r_j}{T_j} \right\rfloor C_j
\] (7)

At most, one partially complete outer arrival can interfere over the remaining part of the window not already accounted for by whole arrivals. This remaining time amounts to

\[
\left( r_j - \left\lfloor \frac{r_j}{T_j} \right\rfloor \right) \frac{C_j}{T_j}
\]

and lies in the range \( \left\lfloor \frac{r_j}{T_j} \right\rfloor \). We shall denote this value by \( Q_{ij} \). The interference over this remaining time is bounded by

\[
\left\lfloor \frac{Q_{ij}}{t_j} \right\rfloor C_j
\] (8)

The above equation assumes that task \( j \) executes as a continual periodic task (with period \( t_j \) ) over the remaining interval. However, task \( j \) cannot execute for more than \( n_j \) periods in this interval (since the interval covers only a partially complete outer arrival), and another bound can be obtained:

\[
n_j \frac{C_j}{t_j}
\] (9)

The least upper bound can therefore be used:

\[
\min \left( \left\lfloor \frac{Q_{ij}}{t_j} \right\rfloor, n_j \right) C_j
\] (10)

Combining eqns. 10 and 7, and summing over all higher priority tasks, we obtain

\[
l = \sum_{j \in \text{seq}(i)} \min \left( \left\lfloor \frac{Q_{ij}}{t_j} \right\rfloor, n_j \right) C_j + n_j \left( \frac{C_i}{T_i} \right) C_j
\] (11)

If a task \( j \) is not sporadically periodic, then we choose \( n_j = 1 \) and \( t_j = T_j \). As a check for eqn. 11, we assume that all tasks are not sporadically periodic, and hence for all tasks \( n_j = 1 \) and \( t_j = T_j \). From eqn. 11, we have

\[
l = \sum_{j \in \text{seq}(i)} \left( \frac{C_j}{T_j} \right) C_j
\]

which is equal to eqn. 2. Hence, eqn. 11 is a generalisation of eqn. 2.

We now return to the problem of release jitter. There are two potential places where release jitter could occur; on an
outer arrival (where the first arrival of a succession of \( n_j \) inner arrivals of a task \( j \) is deferred), and on an inner arrival (where each of the \( n_j \) arrivals could experience delay). For simplicity, we assume that the outer arrival jitter and the inner arrival jitter are the same. For a task \( j \), we assume that this jitter is denoted \( J_j \). Following the same argument as for the derivation of jitter in eqn. 5, we can modify eqn. 11 to include release jitter:

\[
l_i = \sum_{i \in \text{tasks}} \left( \min \left( \frac{J_j + n_j \cdot r_j}{T_j} \right) + n_i \cdot \frac{J_j + r_j}{T_j} \right) C_i
\]

(12)

As with eqn. 3, an iterative equation to find \( r_i \) can be formulated. The worst-case response time of a task, measured from arrival to termination, is again given by \( J_j + r_i \).

6 Discussion and case study

In this Section, we analyse and discuss the task set of a small avionics case study undertaken by Locke et al. [19].

Several mostly periodic tasks implement an avionics weapons management subsystem. There is a single sporadic task and a single task where the deadline of the task is less than the period (for reduced outer 'jitter' requirements). Task priorities are assigned according to the Deadline Monotonic policy. Originally, the tasks were analysed using the Rate Monotonic schedulability analysis derived by Shao et al. [2]. In the case study [19], Locke et al. report that, using this analysis, only the 8 highest priority tasks out of a set of 18 tasks could be guaranteed to meet their deadlines. In simulations, nearly all tasks were found to meet their timing requirements (two tasks were reported as missing their deadlines).

Eqn. 5 was applied to the task set described by Locke et al. [19], using the given priority assignment. For the single sporadic task, a release jitter of 1000 \( \mu \)s was assumed, to account for the worst-case delay due to the operation of the tick-driven scheduler (see the above discussion of induced jitter from tick-driven scheduling). The other tasks are all periodic, with periods that are multiples of \( T_{aw} = 1000 \mu \)s, and hence do not experience release jitter. Table 3 lists the tasks in priority order (task 1, the tick-driven scheduler, is the highest priority task), and supplies the attributes and the derived response times of the tasks using eqn. 5. All times are given in micro-seconds.

As can be seen from Table 3, our analysis predicts that all deadlines can be met except for task 11. Locke et al. found that task 11 did indeed miss its deadline occasionally. They also found that task 16 missed a deadline once. This discrepancy can be explained if the scheduler implementation does not exactly agree with the assumptions made in this paper.

The case study was implemented in Ada. Most Ada runtime systems make use of two queues; a run-queue which holds all runnable tasks and a delay-queue which holds all (periodic) task that are waiting for their next release. At any particular tick, the number of tasks to be moved from the delay queue to the run-queue varies between none and sixteen. A standard runtime system will not undertake this at a constant cost (in computation time); hence, the value of \( C_i \) of 51 \( \mu \)s is potentially an underestimation. Furthermore, the costs of context switches must be accounted for accurately, along with any blocking factors due to the operation of the system (for example, calls to the Ada runtime in most implementations are generally not pre-emptible, and hence can induce a blocking factor on all tasks). It is therefore unlikely that the results of experiments and analysis can be made. In

![Diagram](image-url)
general, however, our analysis agrees with the observed behaviour. Moreover, it matches the observed behaviour more closely than the original Rate Monotonic analysis.

7 Summary and conclusions

We have presented results which provide simple exact analysis for systems scheduled at runtime with a static priority pre-emptive dispatcher. The analysis has been extended to include release jitter, allowing tasks to arrive and then be deferred for a bounded amount of time. The analysis has been further extended to permit sporadically repeating tasks to be analysed exactly. A case study [19], already analysed according to Rate Monotonic scheduling theory, has been re-analysed using this theory. The basis of the analysis is the development of formulae which predict the worst-case interference a task can suffer from higher priority tasks; utilisation-based analysis is not used as this cannot cater for systems which contain tasks with deadlines less than periods.

The most important aspects of our scheduling theory are that older scheduling theory can be considered a special case of the analysis presented in this paper (systems previously analysed by the Rate Monotonic approach can now be re-analysed using more powerful techniques), and that the analysis presented here can be extended in a straightforward manner to allow for more complex and powerful systems to be investigated.

8 Acknowledgments

The authors would like to thank the authors of the case study for help with its analysis and for comments on an earlier draft of this paper.

9 References


10 Appendix: A brief description of STRESS diagrams

STRESS diagrams illustrate the execution of tasks under the STRESS simulator. In these diagrams, time increases from left to right.

Task execution is represented by boxes. A task which is pre-empted is shown by a line at the level of the bottom of the boxes; a task which is deferred is shown by a line at the level of the top of the boxes. These states are annotated by a variety of symbols.

Task release is marked by an open low-level circle, and successful task completion by an open high-level circle. If a task fails to meet its deadline, or otherwise fails to complete, then a solid high-level circle is used. Task deadlines are marked by a vertical line with a ▲ mark at the bottom.

An example is shown in Fig. 7. Task 0 and task 1 are released at times 2 and 0, respectively, with deadlines at times 10 and 8, respectively, and require 6 and 3 computation ticks, respectively. Task 1 is deferred for 4 ticks, executes for 3 further ticks and then completes. Task 0 executes for 2 ticks, before being pre-empted at tick 4 and resumed at tick 7; it fails to meet its deadline and is killed.

The paper was first received 15 April 1992 and in its revised form 9 February 1993.

The authors are with the Department of Computer Science, University of York, Heslington, York YO1 5DD, U.K.

Software Engineering Journal September 1993