An Improved Schedulability Test for Uniprocessor Periodic Task Systems

UmaMaheswari C. Devi

Department of Computer Science, University of North Carolina, Chapel Hill, NC 27599-3175

December 2002

Abstract

We present a sufficient linear-time schedulability test for preemptable, asynchronous, periodic task systems with arbitrary relative deadlines, scheduled on a uniprocessor by an optimal scheduling algorithm. We show analytically and empirically that this test is more accurate than the commonly-used density condition. We also present and discuss the results of experiments that compare the accuracy and execution time of our test with that of a pseudo-polynomial-time schedulability test presented previously for a restricted class of task systems in which utilization is strictly less than one.

1 Introduction

We consider the problem of determining the schedulability of preemptable, asynchronous, periodic hard-real-time task systems with arbitrary relative deadlines, scheduled on a single processor by an online scheduling algorithm. This problem has been studied extensively, and there are now a number of well-known online scheduling algorithms with associated schedulability tests. It is also well-known that it is difficult to determine efficiently, i.e., in polynomial-time, whether a task system can be correctly scheduled by some scheduling algorithm if tasks exist that have relative deadlines less than their periods. Hence, when the amount of time available to determine schedulability is limited, such as in online admission-control tests, polynomial-time sufficient schedulability tests are generally used. However, sufficient tests are not capable of identifying every schedulable task set and result in schedulability loss. In this paper, we propose an improved linear-time sufficient schedulability test that can

*Work supported by NSF grants ITR 0082866 and CCR 0204312.
identify a larger percentage of task sets than other known polynomial-time tests.

2 Background and Related Work

A periodic task system consists of a set of tasks \( \tau = \{T_1, T_2, \ldots, T_n\} \), where each task \( T_i \) is a sequential program defined by a four-tuple \( (\phi_i, p_i, e_i, D_i) \), where \( \phi_i \geq 0, p_i > 0, e_i > 0, \) and \( D_i > 0 \) are the task’s phase, period, execution time, and relative deadline, respectively. Each task \( T_i \) issues periodic requests for \( e_i \leq \min(p_i, D_i) \) time units of execution, separated by \( p_i \) time units. Every such request, known as a job, must complete within \( D_i \) time units from the time it was issued. The phase, \( \phi_i \), denotes the time when \( T_i \) makes its first request. Such a task system is synchronous if all tasks release their initial jobs at the same time, and asynchronous, otherwise. In this paper we assume that task systems are specified by arranging tasks in order of non-decreasing relative deadlines.

The ratio of the execution cost of \( T_i \) to its period, \( e_i/p_i \), is defined as its utilization. If each task’s relative deadline is at least its period, then schedulability can be determined exactly by checking that total utilization is at most one, i.e., by checking that \( \sum_{i=1}^{n} e_i/p_i \leq 1 \) [8]. However, if relative deadlines are less than periods, then this linear-time utilization-based test is not sufficient. The only known method for exactly determining schedulability in such a case is by checking that no deadlines are missed in a schedule constructed using an optimal scheduling algorithm. The minimum time over which a schedule has to be constructed is known to be \( 2P + \max_{1 \leq i \leq n} \{D_i\} + \max_{1 \leq i \leq n} \{\phi_i\} \), where \( P \) is the least common multiple of the task periods, which is exponential in \( n \) [7]. Thus, this exact test runs in exponential time.

This problem of determining the schedulability of periodic task systems with arbitrary relative deadlines, which shall henceforth be referred to as ARD, is known to be co-NP-complete in the strong sense for asynchronous periodic task systems (ARD-Async), and in co-NP for synchronous periodic task systems (ARD-Sync) [3]. Hence, no exact polynomial-time utilization-based test or exact pseudo-polynomial-time demand-based test is possible for ARD-Async unless \( P = NP \). It is currently unknown whether ARD-Sync is co-NP-hard or whether it has a polynomial-time or pseudo-polynomial-time solution. Checking that total density is at most one, i.e., \( \sum_{i=1}^{n} e_i/\min(p_i, D_i) \leq 1 \), where \( e_i/\min(p_i, D_i) \) is the density of task \( T_i \), is the most commonly-used linear-time test, but it is only sufficient. Task systems with deadlines less than periods are not uncommon in
practice, and hence it is desirable that this sufficient condition be tightened. In this paper, we present one such schedulability test, which is tighter than the density condition and can be evaluated in time that is linear in the number of tasks. Note that the density test and our proposed test may not be applicable if the scheduling algorithm being used is not optimal. Our focus in this paper is to determine schedulability under optimal scheduling algorithms.

Baruah et al. showed that ARD-Sync can be solved in pseudo-polynomial-time if total utilization is capped to some fixed positive constant $c$, where $0 \leq c < 1$ [3]. Capping utilization limits the time interval over which demand has to be checked to $\frac{c}{1-c} \max_{1 \leq i \leq n}(p_i - D_i)$. Since $\frac{c}{1-c}$ is a constant, this test runs in $O(n \max_{1 \leq i \leq n}(p_i - D_i))$ time. We shall hereafter refer to this test as the PPT test (pseudo-polynomial-time test).

To determine how well our test performs in comparison to the PPT test, we conducted a series of experiments involving randomly-generated task sets. These experiments showed that the accuracy of our test was usually comparable to that of the PPT test. However, at high utilizations, the PPT test was more accurate. On the other hand, at such high utilizations the running time of the PPT test was three orders of magnitude greater than our test. Hence, the choice of which test to use is largely a trade-off between schedulability and efficiency.

The rest of the paper is organized as follows. In Sec. 3, we present our more accurate schedulability test, prove it correct, and show that it is tighter than the density condition. In Sec. 4, we show how to adapt the test to account for context-switching overheads, nonpreemptivity of tasks, contention for resources among tasks, interrupt handlers, self-suspensions, and priority-space limitations. Then, in Sec. 5, we present the experimental results summarized above. We conclude in Sec. 6.

## 3 Improved Schedulability Test

Our proposed schedulability test is given by the following theorem.

**Theorem 1** Let $\tau = \{T_1, T_2, \ldots, T_n\}$ be a system of $n$ preemptable, asynchronous, periodic tasks, with arbitrary relative deadlines, arranged in order of non-decreasing relative deadlines. $\tau$ is schedulable using an optimal
Task $T_i$ \[ e_i \quad D_i \quad p_i \]

Task $T_k$ \[ e_k \quad D_k \quad p_k \]

Figure 1: Schedule with a deadline miss.

$t_d$: earliest time when a deadline is missed

$t_{-1}$: latest idle instance before $t_d$

$[t_{-1}, t_d)$: processor is not idle and no task whose active job’s deadline is later than $t_d$ executes in this interval.

$scheduling algorithm if$

\[
\left( \forall k: 1 \leq k \leq n : \sum_{i=1}^{k} \frac{e_i}{p_i} + \frac{1}{D_k} \sum_{i=1}^{k} \left( \frac{p_i - \min(p_i, D_i)}{p_i} \right) \cdot e_i \leq 1 \right). \tag{1}
\]

**Proof:** Since an *optimal* scheduling algorithm correctly schedules any schedulable task system, it is sufficient that one such algorithm be used in determining if a task system $\tau$ is schedulable. The *earliest-deadline-first* (EDF), which prioritizes jobs with earlier deadlines over those with later deadlines [8], is known to be optimal for scheduling the class of task systems considered in the theorem [4]. It is the optimal scheduling algorithm assumed in our proof.

We prove the above theorem by proving the contrapositive, *i.e.*, by showing that if $\tau$ is not schedulable by EDF, then (1) is false. Let $t_d$ be the first instant at which a job of some task $T_j$ misses its deadline under EDF, as shown in Fig. 1. Let $t_{-1}$ be the latest instant before $t_d$ at which the processor was idle or was executing a job whose deadline is after $t_d$. Since $T_j$ misses its deadline at $t_d$, the total demand placed on the processor in the interval $[t_{-1}, t_d)$ is greater than the length of this interval. The maximum demand placed on the processor in the interval is given by

\[
\sum_{1 \leq i \leq n \land D_i \leq (t_d - t_{-1})} \left( \left\lfloor \frac{t_d - t_{-1} - D_i}{p_i} \right\rfloor + 1 \right) \cdot e_i.
\]
If \( k \ (1 \leq k \leq n) \) is the highest index of any task whose relative deadline is not greater than the length of the interval \([t-1, t_d]\), then the total demand placed on the processor in the interval is bounded from above by

\[
\sum_{i=1}^{k} \left\lfloor \frac{t_d - t - 1 - D_i}{p_i} \right\rfloor \cdot e_i + e_i.
\]

Because \( T_j (j \leq k) \) misses its deadline at \( t_d \), we have the following:

\[
\begin{align*}
t_d - t - 1 &< \sum_{i=1}^{k} \left\lfloor \frac{t_d - t - 1 - D_i}{p_i} \right\rfloor \cdot e_i + e_i \\
&\leq \sum_{i=1}^{k} \left( \frac{t_d - t - 1 - D_i}{p_i} \right) \cdot e_i + e_i \\
&\leq \sum_{i=1}^{k} \left( \frac{t_d - t - 1 - \min(p_i, D_i)}{p_i} \right) \cdot e_i + e_i \\
&= \sum_{i=1}^{k} \frac{e_i}{p_i} (t_d - t - 1 + p_i - \min(p_i, D_i)).
\end{align*}
\]

Dividing both sides of the above inequality by \( t_d - t - 1 \), we have

\[
1 < \sum_{i=1}^{k} \frac{e_i}{p_i} \left( 1 + \frac{p_i - \min(p_i, D_i)}{t_d - t - 1} \right)
\]

\[
= \sum_{i=1}^{k} \frac{e_i}{p_i} + \sum_{i=1}^{k} \frac{e_i}{p_i} \left( \frac{p_i - \min(p_i, D_i)}{t_d - t - 1} \right)
\]

\[
\leq \sum_{i=1}^{k} \frac{e_i}{p_i} + \frac{1}{D_k} \sum_{i=1}^{k} \frac{e_i}{p_i} (p_i - \min(p_i, D_i)).
\]

The final step follows because \( D_k \leq (t_d - t - 1) \).

Thus, (1) is false as claimed. \( \square \)

This condition can be checked in \( O(n) \) time, if tasks are sorted by relative deadline. Otherwise, \( \Omega(n \log n) \) time is required.

A task system satisfying (1) may not satisfy the density condition, but if it satisfies the density condition, then (1) will necessarily be satisfied, as the following theorem implies.
Theorem 2  The schedulability test in Theorem 1 is tighter than the density condition.

Proof:  We prove this by showing that the left-hand side of Inequality (1) is less than or equal to the density of the system, i.e.,

\[
\sum_{i=1}^{k} \frac{e_i}{p_i} + \frac{1}{D_k} \sum_{i=1}^{k} \left( \frac{p_i - \min(p_i, D_i)}{p_i} \right) \cdot e_i \leq \sum_{i=1}^{k} \frac{e_i}{\min(p_i, D_i)}, \quad 1 \leq k \leq n.
\]

This is shown as follows.

\[
\sum_{i=1}^{k} \frac{e_i}{p_i} + \frac{1}{D_k} \sum_{i=1}^{k} \left( \frac{p_i - \min(p_i, D_i)}{p_i} \right) \cdot e_i \\
= \sum_{i=1}^{k} \frac{e_i}{p_i} \left( 1 - \frac{\min(p_i, D_i)}{D_k} \right) + \frac{1}{D_k} \sum_{i=1}^{k} e_i \\
\leq \sum_{i=1}^{k} \frac{e_i}{\min(p_i, D_i)} \left( 1 - \frac{\min(p_i, D_i)}{D_k} \right) + \frac{1}{D_k} \sum_{i=1}^{k} e_i \\
= \sum_{i=1}^{k} \frac{e_i}{\min(p_i, D_i)}
\]

\[\square\]

4  Extensions to Account for Practical Overheads

We now show how to extend the test in Theorem 1 to account for context-switching costs, nonpreemptivity of tasks, contention for resources among tasks, interference by interrupt handlers, self-suspensions, and priority-space limitations, when tasks are scheduled using EDF.

The terms *blocking*, *blocking time*, and *priority inversion* are used in this section with their usual meanings, and are defined here for completeness. *Blocking* occurs when a higher-priority job is prevented from executing by a lower-priority job. A *priority inversion* is said to occur when a low-priority job executes while a ready high-priority job waits. *Blocking time* is the duration for which a high-priority job is prevented from executing due to priority inversions.
4.1 Context Switches

In a preemptable system in which the priority of a job does not change over its lifetime, as in EDF, a job that does not self-suspend preempts at most one lower-priority job. Thus, each job suffers at most one context switch when it starts executing and another when it completes. It is well-known that any additional context switches a job may suffer due to preemptions by higher-priority jobs can be accounted for by charging them to the preemting higher-priority jobs. Hence, overhead due to context switches can be fully accounted for if the execution cost of each job is increased by twice the worst-case context-switching time of the system. Thus, we have the following theorem.

**Theorem 3** A preemptable, asynchronous, periodic task system consisting of $n$ tasks with arbitrary relative deadlines arranged in order of non-decreasing relative deadlines, and a worst-case context-switching time $CS$ is schedulable under EDF if

$$\left( \forall k : 1 \leq k \leq n : \sum_{i=1}^{k} \frac{e_i + 2 \cdot CS}{p_i} + \frac{1}{D_k} \sum_{i=1}^{k} \left( \frac{p_i - \min(p_i, D_i)}{p_i} \right) \cdot (e_i + 2 \cdot CS) \leq 1 \right).$$

4.2 Nonpreemptivity and Resource Contentions

Nonpreemptable sections and the use of shared locks can cause priority inversions. Though the relative priorities of tasks vary with their instances (jobs) under EDF, the set of tasks with jobs that can block $T_i$ is static; specifically, this set consists of only tasks with relative deadlines greater than $D_i$, as the following theorem proved in [2] shows.

**Theorem 4** [2] Under EDF scheduling, a job $J_k$ with relative deadline $D_k$ can block a job $J_i$ with relative deadline $D_i$ only if $D_k$ is larger than $D_i$.

In the absence of self-suspensions, any job can be blocked by at most one lower-priority job due to nonpreemptivity. Hence, when tasks are arranged in order of non-decreasing relative deadlines (as we have assumed), the maximum blocking time of $T_i$ due to nonpreemptivity, $b_i(np)$, is given by

$$b_i(np) = \max_{1 < k \leq n} \theta_k, \quad (2)$$
where \( \theta_k \) is the maximum execution time of the longest nonpreemptable section of \( T_k \) \[9\].

Blocking times of tasks due to contention for shared resources depend upon the synchronization mechanism and the resource access-control protocol used. In this subsection, we assume that jobs are synchronized using lock-based schemes. (We consider lock-free schemes in the next subsection.) We also assume that resources are allocated using the priority-ceiling protocol \[6\] or the stack-based resource allocation policy \[2\]. These protocols avoid deadlocks and limit the number of times a job is blocked to at most one for the duration of the longest outermost critical section of a lower-priority job.

The maximum blocking time, \( b_i(rc) \), of \( T_i \) due to resource contention is then given by

\[
\begin{align*}
  b_i(rc) = \max_{i<k \leq n} \gamma_{i,k},
\end{align*}
\]

where

\[
\gamma_{i,k} = \begin{cases} 
  0, & T_k \text{ does not have resource conflicts with } T_1, \ldots, T_i \\
  \gamma_k, & \text{otherwise}
\end{cases}
\]

In the above equation, \( \gamma_k \) is the maximum execution time of the longest outermost critical section of \( T_k \). (Here again, tasks are assumed to be arranged in order of non-decreasing relative deadlines.) Note that, to avoid deadlocks, a job of \( T_i \) may be blocked by a job of \( T_k, k > i \), even if \( T_i \) and \( T_k \) do not have common resource requirements, but if \( T_k \) shares a resource with \( T_j, j < i \).

A schedulability test that accounts for blocking times due to nonpreemptivity and resource contention described above is then given by the following theorem. (We assume that a nonpreemptable section does not access a shared resource.)

**Theorem 5** Let \( \tau \) be an asynchronous, periodic task system consisting of \( n \) tasks with arbitrary relative deadlines, arranged in order of non-decreasing relative deadlines. Let the tasks in \( \tau \) share resources using lock-based schemes and contain nonpreemptive sections that do not access shared resources, and let their blocking times due to nonpreemptivity and resource contention be given by Equations (2) and (3), respectively. \( \tau \) is schedulable under EDF if

\[
\left( \forall k : 1 \leq k \leq n : \sum_{i=1}^{k} \frac{e_i}{p_i} + \frac{1}{D_k} \sum_{i=1}^{k} \left( \frac{p_i - \min(p_i, D_i)}{p_i} \right) \cdot e_i + \frac{b_k(np) + b_k(rc)}{D_k} \leq 1 \right).
\]
Proof: The set-up is similar to that of Theorem 1. Let \( t_d \) be the first instant at which a job of \( T_j \) misses its deadline under EDF. Let \( t_{-1} \) be the latest instant before \( t_d \) when the processor was idle or no job with a deadline at or before \( t_d \) was active. (A job \( J_{k,l} \) of \( T_k \) is active at time \( t \), if \( \text{release time}(J_{k,l}) \leq t \) and \( J_{k,l} \) has executed for less than \( e_k \) time units.) Let \( J \) be the set of jobs with deadlines at or before \( t_d \) that are released at or after \( t_{-1} \). A job \( J_b \notin J \) may execute in \([t_{-1}, t_d)\) only if it is in a nonpreemptive section or in a critical section accessing a resource common to some \( J \in J \) at \( t_{-1} \) \([2\). (Note that it is obvious that there could be at most one job of the first kind and that it can be shown that there could be at most one job of the second kind under the resource access-control protocols assumed in this subsection \([2\).) This is illustrated in Fig. 2, which shows a sample schedule in which nonpreemptive and critical sections of jobs \( J_{b1} \) and \( J_{b2} \), respectively, lead to a deadline miss in a higher-priority job.

If \( T_k \) is the task with the highest index whose deadline is not greater than \( t_d - t_{-1} \), then the maximum durations that \( J_{b1} \) and \( J_{b2} \) could execute for in \([t_{-1}, t_d)\) are given by \( b_k(np) \) and \( b_k(rc) \), respectively. Hence, the maximum demand on the processor in \([t_{-1}, t_d)\) is given by

\[
\sum_{i=1}^{k} \left( \left\lfloor \frac{t_d - t_{-1} - D_i}{p_i} \right\rfloor e_i + e_i \right) + b_k(np) + b_k(rc).
\]
If a job of $T_j$ misses its deadline, then the demand in $[t_{-1}, t_d)$ is greater than the length of this interval, i.e.,

$$t_d - t_{-1} < k \sum_{i=1}^{k} \left( \left\lfloor \frac{t_d - t_{-1} - D_i}{p_i} \right\rfloor e_i + e_i \right) + b_k(np) + b_k(rc).$$

Manipulating the above inequality in the same way as in proof of Theorem 1, we arrive at

$$\sum_{i=1}^{k} e_i p_i + \frac{1}{D_k} \sum_{i=1}^{k} \left( p_i - \min(p_i, D_i) \right) e_i + \frac{b_k(np) + b_k(rc)}{D_k} > 1,$$

which is in contradiction to (4).

Note that, under the priority-ceiling protocol, a job may be blocked after it begins execution, and hence suffer two more context switches, one when it is blocked and a second when it resumes execution (in addition to the two when it begins and finishes executing).

### 4.3 Lock-free Synchronization

An alternative to lock-based protocols when implementing shared data objects is to use the *lock-free* algorithms. Lock-free algorithms work particularly well for simple objects like buffers, queues, and lists. In such algorithms, object calls are implemented using “retry loops.” Fig. 3 depicts a lock-free enqueue operation that is implemented in this way. An item is enqueued in this implementation by using a *two-word compare-and-swap* (CAS2) instruction to atomically update a tail pointer and either the “next” pointer of the last item in the queue or a head pointer, depending on whether the queue is empty. This loop is executed repeatedly until the CAS2 instruction succeeds. An important property of lock-free implementations such as this is that operations may *interfere* with each other. An interference results in this example when a successful CAS2 by one task causes another task’s CAS2 to fail.

Anderson *et al.* showed that the number of interferences that a job suffers is bounded under priority driven scheduling algorithms like EDF [1]. Assuming a worst-case execution cost of $s$ for each retry loop, they derived sufficient conditions for scheduling with lock-free shared objects under the EDF, rate-monotonic, and deadline-

---

1. The first two parameters of CAS2 specify addresses of two shared variables, the next two parameters are values to which these variables are compared, and the last two parameters are new values to assign to the variables if both comparisons succeed. Although CAS2 is uncommon, it makes for a simple example here.
typedef Qtype: record data: valtype; next: pointer to Qtype
shared var Head, Tail: pointer to Qtype
private var old, new: pointer to Qtype; input: valtype;
addr: pointer to pointer to Qtype

procedure Enqueue(input)
  *new := (input, nil);
do old := Tail;
  if old ≠ nil then addr := &(*(old).next) else addr := &Head fi
while ¬CAS2(&Tail, addr, old, nil, new, new)

Figure 3: Lock-free enqueue implementation.

monotonic scheduling algorithms, and formally showed that lock-free objects often incur less overhead than
lock-based schemes. When lock-free shared objects are included in our task model, a schedulability test under
EDF is given by:

\[
\left( \forall k : 1 \leq k \leq n \right) \sum_{i=1}^{k} \frac{e_i + s}{p_i} + \frac{1}{D_{\max}} \sum_{i=1}^{k} \left( \frac{p_i - \min(p_i, D_i)}{p_i} \right) \cdot (e_i + s) \leq 1 ,
\]

where \( s \) is the maximum execution cost of a retry-loop (and tasks are sorted by relative deadline).

To see that (5) is correct, note that demand placed on the processor in any interval is equal to the sum of
the execution cost of the jobs scheduled in that interval, plus any additional demand created by failed retry
loops in such jobs. Under the set-up of Theorem 1, the number of failed retry loops in \([t_{-1}, t_d)\) is bounded from
above by the number of preemptions in the interval, which in turn is bounded by

\[
\sum_{i=1}^{k} \left( \left\lceil \frac{t_d - t_{-1} - D_i}{p_i} \right\rceil + 1 \right)
\]

Hence, demand over the interval \([t_{-1}, t_d)\) is bounded from above by

\[
\sum_{i=1}^{k} \left\lceil \frac{t_d - t_{-1} - D_i + p_i}{p_i} \right\rceil \cdot e_i + \sum_{i=1}^{k} \left\lceil \frac{t_d - t_{-1} - D_i + p_i}{p_i} \right\rceil \cdot s.
\]

Given this expression, (5) can be established in a similar manner as before.

The exact schedulability test given in [1] requires checking that demand does not exceed the available
processor time over any interval whose length, $l$, is at least the least common multiple of the task periods. If total utilization is known to be strictly less than one, then $l$ can be made much shorter by using the ideas in the PPT test of Baruah et al. [3]. However, the test would still run in pseudo-polynomial-time, and hence, the sufficient test in Inequality (5) may be preferrable if efficiency is a concern.

We summarize the results of this subsection with the following theorem.

**Theorem 6** A preemptable, asynchronous, periodic task system consisting of $n$ tasks with arbitrary relative deadlines, arranged in order of non-decreasing relative deadlines, and sharing lock-free objects is schedulable under EDF if

$$\forall k : 1 \leq k \leq n :: \sum_{i=1}^{k} \frac{e_i + s}{p_i} + \frac{1}{D_k} \sum_{i=1}^{k} \left( \frac{p_i - \min(p_i, D_i)}{p_i} \right) \cdot (e_i + s) \leq 1,$$

where $s$ is the worst-case execution time of a retry-loop in the implementation of a lock-free object.

### 4.4 Interrupt Handling Costs

Interrupt handlers execute in response to external events at priorities higher than the real-time tasks in the system. Jeffay et al. presented a pseudo-polynomial-time exact schedulability test for deadline-driven systems consisting of tasks with relative deadlines equal to periods when interrupt handlers are included and the total system utilization (including interrupt handlers) is less than one [5]. In this subsection, our schedulability test is extended to account for $m$ interrupt handlers, with an interrupt handler $I$ formally modeled as a tuple $(c, a)$, where $c$ is the maximum execution time of $I$ and $a$ is the minimum interval between invocations of $I$ [5]. We assume that all interrupts are of the same priority, they have no associated deadlines, and the order in which outstanding interrupts are handled is not significant. The following theorem gives conditions that are sufficient for a system with such interrupts to be schedulable.

**Theorem 7** Let $\tau$ be a system of $n$ preemptable, asynchronous, periodic tasks, with arbitrary relative deadlines, arranged in order of non-decreasing relative deadlines and $m$ interrupt handlers $\{(c_1, a_1) \ldots (c_m, a_m)\}$. $\tau$ is
schedulable using EDF if

\[
\left( \forall k : 1 \leq k \leq n : \sum_{i=1}^{k} \frac{e_i}{p_i} + \frac{1}{D_k} \sum_{i=1}^{k} \left( \frac{p_i - \min(p_i, D_i)}{p_i} \right) \cdot e_i + \sum_{j=1}^{m} \frac{c_j}{a_j} + \frac{1}{D_k} \sum_{j=1}^{m} c_j \leq 1 \right) .
\]  

(6)

**Proof:** As before, let \( t_d \) be the first instant at which a job of \( T_j \) misses its deadline under EDF. Let \( t_{-1} \) be the latest instant before \( t_d \) when the processor was idle or no job with a deadline at or before \( t_d \) was active. Let \( T_k \) be the task with the highest index whose relative deadline is less than or equal to the length of the interval \([t_{-1}, t_d)\). The demand placed on the processor in the interval \([t_{-1}, t_d)\) in this case is bounded from above by

\[
\text{demand} = \sum_{i=1}^{k} \left[ \frac{t_d - t_{-1} - D_i + p_i}{p_i} \right] \cdot e_i + \sum_{j=1}^{m} \left[ \frac{t_d - t_{-1}}{a_j} \right] \cdot c_j.
\]

Since \( T_j \) misses its deadline, \( t_d - t_{-1} < \text{demand} \), which implies that

\[
\sum_{i=1}^{k} \frac{e_i}{p_i} + \frac{1}{D_k} \sum_{i=1}^{k} \left( \frac{p_i - \min(p_i, D_i)}{p_i} \right) \cdot e_i + \sum_{j=1}^{m} \frac{c_j}{a_j} + \frac{1}{D_k} \sum_{j=1}^{m} c_j > 1,
\]

which contradicts (6).

\( \square \)

4.5 **Self-suspensions**

A self-suspension by a job, unlike blocking, affects both that job and lower-priority jobs. When a job is self-suspended, it is denied processor time, which reduces the time available for its completion. This leads to a task not behaving like a periodic task, in that it may demand more execution time over some interval than a periodic task. This may also cause a delay in the execution of lower-priority jobs, which may in turn miss their deadlines [9]. It is shown in [9] how to account for self-suspensions in fixed-priority systems by considering delays due to self-suspensions as blocking factors. We adopt a similar approach here to account for self-suspensions under EDF.

If \( \xi_k \) is the maximum self-suspension time of a job \( J_{k,l} \) of \( T_k \), then the maximum delay in the execution time of jobs of priority lower than \( J_{k,l} \) due to self-suspensions by \( J_{k,l} \) is equal to \( \min(\xi_k, e_k) \) [9]. Since the relative priorities of tasks vary with their instances (jobs) under EDF, a self-suspension by a task can impact all other tasks (unlike blocking due to nonpreemptivity or critical sections, in which a task can be blocked only by those with larger relative deadlines). Fig. 4 shows a schedule in which self-suspension in \( T_j \) causes \( T_i \) to
miss its deadline, where \( D_j > D_i \). The maximum delay in a job of \( T_j \), due to self-suspensions by jobs of tasks \( T_1, \ldots, T_k \), \( 1 \leq j \leq k \), is given by

\[
  s_j = \sum_{1 \leq i \leq k, i \neq j} \min(\xi_i, e_i) + \xi_j.
\]

Let

\[
  S_k = \sum_{i=1}^{k} \min(\xi_i, e_i)
\]

and

\[
  S'_k = \max_{1 \leq i \leq k} \left( \max(0, \xi_i - e_i) \right).
\]

Then, the maximum time that a job of tasks \( T_1, \ldots, T_k \) could be self-suspended in an interval in which only jobs of \( T_1, \ldots, T_k \) execute is upper-bounded by \( S_k + S'_k \). (Note that \( S_k + S'_k \geq s_j \), for all \( j, 1 \leq j \leq k \).) A schedulability test accounting for self-suspensions is then given by the following theorem.

**Theorem 8** Let \( \tau = \{T_1, T_2, \ldots, T_n\} \) be a system of \( n \) preemptable, asynchronous, periodic tasks, with arbitrary relative deadlines, arranged in order of non-decreasing relative deadlines. Let \( \xi_i \) be the maximum self-suspension time of \( T_i \). \( \tau \) is schedulable using EDF if

\[
  \left( \forall k : 1 \leq k \leq n :: \sum_{i=1}^{k} \frac{e_i}{p_i} + \frac{1}{D_k} \sum_{i=1}^{k} \left( \frac{p_i - \min(p_i, D_i)}{p_i} \right) \cdot e_i + \frac{S_k + S'_k}{D_k} \leq 1 \right),
\]

where \( S_k \) and \( S'_k \) are given by Equations (7) and (8), respectively.

**Proof:** As before, let a job \( J_{j,t} \) of \( T_j \) miss its deadline at \( t_d \). The delay in \( J_{j,t} \) due to self-suspensions in \([t_d, t_{-1})\) is at most \( S_k + S'_k \). The theorem can be proved by inflating demand by this quantity, and following the same approach as in the proof of Theorem 1.

---

Figure 4: Schedule with a deadline miss due to self-suspension.
Self-suspensions also result in an increase in the number of context switches and blockings due to nonpreemptivity and resources contentions. If the maximum number of times each task self-suspends is the same, say $M$, then increasing the overhead due to context switches by a factor of $2M$ and blocking overheads by a factor of $M$ would suffice. However, if the maximum number of self-suspensions varies, then determining blocking overheads accurately becomes complicated.

### 4.6 Limited Priorities

The number of priorities that an actual system supports may be less than the number of distinct priorities of the tasks that are to be scheduled in that system. As a result, jobs that would otherwise be assigned distinct priorities may be assigned the same priority, leading to a loss in schedulable utilization.

Let $n$ denote the number of distinct task priorities required. Then, under EDF, the number of distinct relative deadlines among the tasks is equal to $n$. One way to implement EDF is using $n$ FIFO queues, one for each distinct relative deadline [9]. Ready jobs with the same relative deadline are appended to the queue associated with that deadline, which ensures that these jobs are ordered among themselves according to their absolute deadlines. Hence, the job with the earliest deadline can be found by examining the job, if any, at the head of each queue. Let $m$ be the number of system priorities. If $m < n$, then the number of FIFO queues is less than $n$ and EDF ceases to support naturally occurring priorities. In such a case, jobs of tasks with different relative deadlines are made to share a single FIFO queue. This artificially advances the absolute deadlines of some jobs, thereby increasing their priorities and scheduling them in preference to other jobs that would be scheduled if there were $n$ queues. The jobs with higher priorities in the unlimited-priority domain could miss their deadlines, as a result. Hence, if $m < n$, then the schedulability test of Theorem 1 is not sufficient. We now show how to extend it in a straightforward manner.

Some notation is in order. If $m < n$, then the $n$ task priorities (or relative deadlines, in the case of EDF) have to be mapped onto $m$ system priorities (or relative deadlines). Let $\pi_1, \pi_2, \ldots, \pi_m$ denote the $m$ system deadlines, where $1 \leq \pi_k \leq n$, $1 \leq k \leq m$, and $\pi_j < \pi_k$, if $j < k$. The set $\{\pi_1, \pi_2, \ldots, \pi_m\}$ can be thought of as a deadline mapping grid [9]. The tasks are mapped onto the grid such that tasks with indices less than or equal to $\pi_1$ (tasks $T_1, \ldots, T_{\pi_1}$) are assigned to the first FIFO queue and are given a relative deadline equal to that of
and all tasks with indices in the range $[\pi_{k-1} + 1, \pi_k]$ are assigned to the $k^{th}$ FIFO queue and are given a relative deadline equal to that of $T_{\pi_{k-1} + 1}$. $\pi_0$ is set to 0 and $\pi_m$ to $n$.

A sufficient test for scheduling tasks with $n$ distinct priorities on a system supporting $m < n$ priorities is given by the following theorem.

**Theorem 9** A system of $n$ preemptable, asynchronous, periodic tasks, with distinct arbitrary relative deadlines, arranged in order of non-decreasing relative deadlines and mapped onto $m < n$ relative deadlines supported by the system per the mapping grid $\{\pi_1, \pi_2, \ldots, \pi_m\}$, is schedulable under EDF if

\[
\left( \forall g : 1 \leq g \leq m :: \sum_{i=1}^{\pi_g} \frac{e_i}{p_i} + \frac{1}{D_{\pi_{g-1} + 1}} \sum_{j=1}^{g} \sum_{i=\pi_{j-1} + 1}^{\pi_j} \left( \frac{p_i - \min(p_i, D_{\pi_{i-1} + 1})}{p_i} \right) \cdot e_i \leq 1 \right).
\]

**Proof:** This is similar to the proof of Theorem 1. $\square$

The time complexity of the test in Theorem 1 is not altered when it is extended to account for one or more of the practical factors described in this section. (It still requires only $O(n)$ time if tasks are arranged in non-decreasing order of relative deadlines, and $O(n\log n)$ time otherwise.)

## 5 Experimental Evaluation

In this section, we describe experiments that were conducted to evaluate our proposed test.

### 5.1 Description of Test Data and Experiments

We compared the density test, our proposed test, and the PPT test (pseudo-polynomial-time test) of Baruah et al. [3] by generating two groups of independent random task sets and subjecting the task sets in both groups to all three tests. Over 16,000 random task sets were generated for each group. Each task set consisted of 100 tasks in Group 1 and 1000 tasks in Group 2. The total utilization of each task set was uniformly distributed between 1% and 100%. For conciseness, we shall refer to the difference between a task’s period and relative difference as its *gap*. Each task set’s average gap was selected using a uniform distribution to be between 0% and 80% of its period. The number of task sets declared schedulable by each test and its running time were calculated. Results are shown in the plots in Figs. 5, 6, and 7.
Figure 5: (a) Schedulability by average gap. (b) Relative schedulability by average gap. (c) Schedulability by utilization. (d) Relative schedulability by utilization.

5.2 Comparison with the Density Test

The variations in the percentage of task sets declared schedulable for both increasing utilizations and increasing gap values for the density test and the proposed test are shown in Fig. 5. The figure also depicts these variations for the PPT test, which we discuss in the next subsection.

When utilization is less than 20%, both the tests admit almost the same percentage of task sets, as can be observed in Fig. 5(a). However, with increasing utilizations, the difference in effectiveness between the two tests increases, with our proposed test finding a larger percentage of task sets to be schedulable. This difference reaches a peak of around 20% when utilization is 50%, and then starts decreasing slowly. The plot normalized
Figure 6: Average running times for PPT with 100 tasks per task set. (a) Running time by utilization. (b) Running time by average gap.

with respect to the PPT test is shown in Fig. 5(b).

Similarly, when the gap is negligible, the difference in effectiveness of the two tests is not appreciable, as can be observed from Fig. 5(c). Here again, as the difference increases, the new test starts admitting a larger percentage of task sets. The difference reaches a peak of 15% when the average gap is 60%, and decreases very slowly after that. The same plot normalized with respect to the PPT test is shown in Fig. 5(d).

From the plots we can infer that when utilization is high and the gap is large, the new test correctly predicts the schedulability of a significantly higher percentage of task sets (than the density test).

5.3 Comparison with the Pseudo-polynomial-time Test

Fig. 5 shows the variation in the percentage of task sets declared schedulable for increasing utilizations and increasing gap values for the PPT test also (in addition to those for the density test and the proposed test). The plots show that the proposed test admits more than 80% of the schedulable task sets at all gap values. However, it does not perform well at high utilization, e.g., it admits only slightly over 20% of the task sets when utilization is higher than 95%. The tests were also timed on a 933MHz Dell dual processor machine. The
average running time for the PPT test is plotted against utilization and average gap for Group 1 (hundred tasks per task set) and Group 2 (thousand tasks per task set) in Figs. 6 and 7, respectively. For Group 1, the running time of the PPT test was of the order of hundreds of milliseconds, at high utilizations. The running time is higher by more than two orders of magnitude (tens of seconds) for Group 2. On the other hand, the proposed test took less than a tenth of a millisecond for Group 1 and less than a millisecond for Group 2. High test execution times may not be acceptable in online admission control tests with stringent timing requirements, and hence, we can conclude that the choice of which test to use is largely a trade-off between schedulability and efficiency. If the practical overheads described in Sec. 4 have to be accounted for, then the time complexity of the PPT test deteriorates to $O(n^2 \max(p_i - D_i))$ and it also ceases to be an exact schedulability test. Using our proposed test under such conditions may be advantageous.
6 Concluding Remarks

In this paper, we have presented a new sufficiency test for determining the schedulability of preemptable, asynchronous, periodic task systems with arbitrary relative deadlines. We have also shown that this test is more accurate than the density condition, both analytically and empirically. We have also empirically compared our test to the pseudo-polynomial-time test of Baruah et al. (PPT test) [3] for a restricted class of task systems in which utilization is strictly less than one. We have found that at high utilizations the relative schedulability of the proposed test is low while the execution time of the PPT test is higher by more than three orders of magnitude. Hence, we conclude that the choice of which test to use is a trade-off between accurate and efficient determination of schedulability. In addition, the running time of the PPT test may be quite high if overheads due to practical factors have to be accounted for.

Acknowledgements: I am grateful to Jim Anderson for his encouragement and helpful suggestions with this work, and for his comments on earlier drafts of this paper.

References