A Unifying Response Time Analysis Framework for Dynamic Self-Suspending Tasks

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***Abstract*—For real-time embedded systems, self-suspending behaviors can cause substantial performance/schedulability degradations. In this paper, we focus on preemptive fixed-priority scheduling for the dynamic self-suspension task model on uniprocessor. This model assumes that a job of a task can dynamically suspend itself during its execution (for instance, to wait for shared resources or access co-processors or external devices). The total suspension time of a job is upper-bounded, but this dynamic behavior drastically influence the interference generated by this task on lower-priority tasks. The state-of-the-art results for this task model can be classified into three categories (i) modeling suspension as computation, (ii) modeling suspension as release jitter, and (iii) modeling suspension as a blocking term. However, several results associated to the release jitter approach have been recently proven to be erroneous, and the concept of modeling suspension as blocking was never formally proven correct. This paper presents a unifying response time analysis framework for the dynamic self-suspending task model. We provide a rigorous proof and show that the existing analyses pertaining to the three categories mentioned above are analytically dominated by our proposed solution. Therefore, all those techniques are in fact correct, but they are inferior to the proposed response time analysis in this paper. The evaluation results show that our analysis framework can generate huge improvements (an increase of up to 50% of the number of task sets deemed schedulable) over these state-of-the-art analyses.**

I. Introduction

The periodic/sporadic task model has been recognized as the basic model for real-time systems with recurring executions. A sporadic real-time task τ_i is characterized by its *worst-case execution time* C_i, its *minimum inter-arrival time* T_i and its *relative deadline* D_i. A sporadic task defines an infinite sequence of task instances, also called *jobs*, that arrive with the minimum inter-arrival time constraint. When a job of task τ_i arrives at time t, the job should finish no later than its *absolute deadline* $t + D_i$, and the next job of task τ_i can only be released no earlier than $t + T_i$. For the periodic task model, the next job is released at time $t + T_i$. T_i is then also referred to as the *period* of τ_i.

The seminal work by Liu and Layland [16] considered the scheduling of periodic tasks and presented the schedulability analyses based on utilization bounds to verify whether the deadlines are met or not. For decades, researchers in real-time systems have devoted themselves to effective design and efficient analyses of different recurrent task models to ensure that tasks can meet their specified deadlines. In most of these studies, *a task usually does not suspend itself*. That is, after a job is released, the job is either executed or stays in the ready queue, but it is not moved to the suspension state. Such an assumption is valid only under the following conditions: (1) the latency of the memory accesses and I/O peripherals is considered to be part of the worst-case execution time of a job, (2) there is no external device for accelerating the computation, and (3) there is no synchronization between different tasks on different processors in a multiprocessor or distributed computing platform.

If a job can suspend itself before it finishes its computation, self-suspension behaviour has to be considered. Due to the interaction with other system components and synchronization, self-suspension behaviour has become more visible in designing real-time embedded systems. Typically, the resulting suspension delays range from a few microseconds (e.g., a write operation on a flash drive [11]) to a few hundreds of milliseconds (e.g., offloading computation to GPUs [12], [18]).

There are two typical models for self-suspending sporadic task systems: 1) the dynamic self-suspension task model, and 2) the segmented self-suspension task model. In the *dynamic* self-suspension task model, in addition to the worst-case execution time C_i of sporadic task τ_i, we have also the worst-case self-suspension time S_i of task τ_i. In the *segmented* self-suspension task model, the execution behaviour of a job of task τ_i is specified by interleaved computation segments and self-suspension intervals. From the system designer's perspective, the dynamic self-suspension model provides a simple specification by ignoring the juncture of I/O access, computation offloading, or synchronization. However, if the suspending behaviour can be characterized by using a segmented pattern, the segmented self-suspension task model can be more appropriate.

In this paper, we focus on preemptive fixed-priority scheduling for the dynamic self-suspension task model on a uniprocessor platform. To verify the schedulability of a given task set, this problem has been specifically studied in [1], [2], [9], [14], [19]. The recent report by Chen et al.[1] and the report by Bletsas et al. [3] have shown that several analyses in the state-of-the-art of self-suspending tasks [1], [2], [14], [19] are in fact unsafe. Unfortunately, those misconceptions propagated to several works [4], [5], [8], [13], [15], [21]–[23] analyzing the worst-case response time for partitioned multiprocessor real-time locking protocols.

Furthermore, one of the seminal result presented by Jane

[1]The report, with a tentative title "Many Suspensions, Many Problems: A Review of Self-Suspending Tasks in Real-Time Systems" (all the authors in this paper are contributors to this review), has been finalized. However, it is under the final internal review by the co-authors. We hope the report to be filed in the early March 2016.

W. S. Liu in her book titled "Real-Time Systems" [17, p. 164-165] and implicitly used by Rajkumar, Sha, and Lehoczky [20, p. 267] for analyzing the self-suspending behaviour due to synchronization protocols in multiprocessor systems, was never proven correct.

Contributions. In this paper, we propose the following contributions:

- We provide a new response analysis framework for dynamic self-suspending sporadic real-time tasks executing on a uniprocessor platform. The key observation in our analysis is that the *interference from higher-priority self-suspending tasks can be arbitrarily modelled as jitter or carry-in terms*.
- We prove that the new analysis analytically dominates all the existing results in the state-of-the-art, excluding the flawed ones.
- We prove the correctness of the analysis initially proposed in [17, p. 164-165] and [20, p. 267], which were never proven correct in the state-of-the-art[2].
- The evaluation results presented in Section VIII show the huge improvement (an increase of up to 50% of the number of task sets that are deemed schedulable) over the state-of-the-art.

II. Task Model

We assume a system τ composed of n sporadic self-suspending tasks. A sporadic task τ_i is released repeatedly, with each such invocation called a job. The j^{th} job of τ_i, denoted by $\tau_{i,j}$, is released at time $r_{i,j}$ and has an absolute deadline at time $d_{i,j}$. Each job of task τ_i is assumed to have a worst-case execution time C_i. Furthermore, a job of task τ_i may suspend itself for at most S_i time units (across all of its suspension phases). When a job suspends itself, it releases the processor and another job can be executed. The response time of a job is defined as its finishing time minus its release time. Successive jobs of the same task are required to execute in sequence.

Each task τ_i is characterized by the tuple (C_i, S_i, D_i, T_i), where T_i is the period (or minimum inter-arrival time) of τ_i and D_i is its relative deadline. T_i specifies the minimum time between two consecutive job releases of τ_i, while D_i defines the maximum amount of time a job can take to complete its execution after its release. It results that for each job $\tau_{i,j}$, $d_{i,j} = r_{i,j} + D_i$ and $r_{i,j+1} \geq r_{i,j} + T_i$. In this paper, we focus on constrained-deadline tasks, for which $D_i \leq T_i$. The utilization of a task τ_i is defined as $U_i = C_i/T_i$.

The worst-case response time R_i of a task τ_i is the maximum response time among all its jobs. A schedulability test for a task τ_k is therefore to verify whether its worst-case response time is no more than its associated relative deadline D_k.

In this paper, we only consider *preemptive fixed-priority scheduling running on a single processor platform*, in which each task is assigned with a unique priority level. We assume that the priority assignment is given beforehand and that the tasks are numbered in a decreasing priority order. That is, a task with a smaller index has a higher priority than any task with a higher index, i.e., task τ_i has a higher-priority than task τ_j if $i < j$.

When performing the schedulability analysis of a specific task τ_k, we will implicitly assume that all the higher priority tasks (i.e., $\tau_1, \tau_2, \ldots, \tau_{k-1}$) are already verified to meet their deadlines, i.e., that $R_i \leq D_i, \forall \tau_i \mid 1 \leq i \leq k-1$.

III. Background

To analyze the worst-case response time (or the schedulability) of a task τ_k, one usually needs to quantify the worst-case interference exerted by the higher-priority tasks on the execution of any job of task τ_k. In the ordinary sequential sporadic real-time task model, i.e., when $S_i = 0$ for every task τ_i, the so-called critical instant theorem by Liu and Layland [16] is commonly adopted. That is, the worst-case response time of task τ_k (if it is less than or equal to its period) happens for the first job of task τ_k when (i) τ_k and all the higher-priority tasks release their first job synchronously and (ii) all their subsequent jobs are released as early as possible (i.e., with a rate equal to their period). However, this definition of the critical instant does not hold for self-suspending sporadic tasks.

The analysis of self-suspending task systems requires to model the self-suspending behavior of both the task τ_k under analysis and the higher priority tasks that interfere with τ_k. The techniques employed to model the self-suspension are usually different for τ_k and the higher priority tasks. The worst-case for τ_k happens when its jobs suspend whenever there is no higher-priority job in the system. The resulting behavior is therefore similar as if the suspension time S_k of task τ_k was converted into computation time (see [10] for more detailled explanations). Second, for the higher-priority tasks, we need to consider the self-suspension behaviour that may result in the largest possible interference for task τ_k. There exist three different approaches in the state-of-the-art that are potentially sound to perform the schedulability analysis of self-suspending tasks:

- modeling the suspension as execution, also known as the suspension-oblivious analysis (see Section III-A);
- modeling the suspension as release jitter (see Section III-B);
- modeling the suspension as blocking time (see Section III-C).

We later prove in Section VI that all these approaches are analytically correct.

A. Suspension-Oblivious Analysis

The simplest analysis consists in converting the suspension time S_i of each task τ_i as a part of its computation time. Therefore, a constrained-deadline task τ_k can be feasibly scheduled by a fixed-priority scheduling algorithm if

$$\exists t \mid 0 < t \leq D_k, \quad C_k + S_k + \sum_{i=1}^{k-1} \left\lceil \frac{t}{T_i} \right\rceil (C_i + S_i) \leq t. \quad (1)$$

[2] A simplified version of the proof of Theorem 1 to support the correctness of [17, p. 164-165] and [20, p. 267] is provided in [6].

B. Modeling the Suspension as Release Jitter

Another approach consists in modeling the impact of the self-suspension S_i of each higher priority task τ_i as release jitter. Several works in the state-of-the-art [1], [2], [14], [19] upper bounded the release jitter with S_i. However, it has been recently shown in [3] that this upper bound is unsafe and the release jitter of task τ_i can in fact be larger than S_i.

Nevertheless, it was proven in the same document [3] that the jitter of a higher-priority task τ_i can be safely upper bounded by $R_i - C_i$. It results that a task τ_k with a constrained deadline can be feasibly scheduled under fixed-priority if

$$\exists t \mid 0 < t \leq D_k, \quad C_k + S_k + \sum_{i=1}^{k-1} \left\lceil \frac{t + R_i - C_i}{T_i} \right\rceil C_i \leq t. \quad (2)$$

C. Modeling the Suspension as Blocking Time

In [17, p. 164-165], Liu proposed a solution to study the schedulability of a self-suspending task τ_k by modeling the extra delay suffered by τ_k due to the self-suspension behavior of each task in τ as a blocking time. This blocking time has been defined as follows:

- The blocking time contributed from task τ_k is S_k.
- A higher-priority task τ_i can block the execution of task τ_k for at most $\min(C_i, S_i)$ time units.

An upper bound on the blocking time is therefore given by:

$$B_k = S_k + \sum_{i=1}^{k-1} \min(C_i, S_i). \quad (3)$$

In [17], the blocking time is then used to derive a utilization-based schedulability test for rate-monotonic scheduling. Namely, it is stated that, if $T_i = D_i$ for every task $\tau_i \in \tau$ and $\frac{C_k + B_k}{T_k} + \sum_{i=1}^{k-1} U_i \leq k(2^{\frac{1}{k}} - 1)$, then τ_k can be feasibly scheduled with rate-monotonic scheduling.

The same concept was also implicitly used by Rajkumar, Sha, and Lehoczky in [20, p. 267] for analyzing the impact of the self-suspendion of a task due to the utilization of synchronization protocols in multiprocessor systems. The statement in [20] reads as follows:[3]

> *"For each higher priority job $\tau_{i,j}$ that suspends on global semaphores or for other reasons, add the term $\min(C_i, S_i)$ to B_k, where S_i is the maximum duration that $\tau_{i,j}$ can suspend itself. [...] The sum [...] yields B_k, which in turn can be used in $\frac{C_k + B_k}{T_k} + \sum_{i=1}^{k-1} U_i \leq k(2^{\frac{1}{k}} - 1)$ to determine whether the current task allocation to the processor is schedulable."*

If the above argument is correct, we can further prove that a constrained-deadline task τ_k can be feasibly scheduled under fixed-priority scheduling if

$$\exists t \mid 0 < t \leq D_k, \quad C_k + B_k + \sum_{i=1}^{k-1} \left\lceil \frac{t}{T_i} \right\rceil C_i \leq t. \quad (4)$$

However, there is no proof in [17] nor in [20] to support the correctness of those tests. Therefore, in Section VI, we provide a proof (see Theorem 3) of the correctness of Equation (4).

[3] We rephrased the wording and notation in order to be consistent with the rest of this paper. Moreover, the multiprocessor scheduling in such a case is based on partitioned scheduling. Therefore, the schedulability analysis of a task set on a processor is the same as the uniprocessor problem by additionally considering the self-suspending behaviour due to the synchronization with other tasks on other processors.

IV. Rationale

Even though it can be proven that the response time analysis associated with Eq.(4) dominates the suspension oblivious one (see Lemma 3 in Section VI), none of the analyses presented in Section III dominates all the others. Hence, Eqs. (2) and (4) are incomparable. That is, in some cases Eq. (4) performs better than Eq. (2), while in others Eq. (2) outperforms Eq. (4).

Example 1. *Consider the two tasks $\tau_1 = (4, 5, 10, 10)$ and $\tau_2 = (6, 1, 19, 19)$. The worst-case response time of τ_1 is obviously 9 whatever the analysis employed. However, the upper bound on the WCRT of τ_2 obtained with Eq. (2) is 15, while it is 19 with Eq. (4). The solution obtained with Eq. (2) is therefore tighter.*

Now, let us consider one more task $\tau_3 = (4, 0, 50, 50)$. Using Eq. (2), the WCRT of task τ_3 is upper bounded by the smallest $t > 0$ such that $t = 4 + \left\lceil \frac{t+9-4}{10} \right\rceil 1 + \left\lceil \frac{t+15-6}{19} \right\rceil 6$, which turns out to be 42. With Eq. (4) though, $B_3 = 4+1 = 5$ (Eq. (3)) and an upper bound on the WCRT of τ_3 is given by the smallest $t > 0$ such that $C_3 + B_3 + \sum_{i=1}^{2} \left\lceil \frac{t}{T_i} \right\rceil C_i \leq t$. The solution to this last equation is $t = 37$. Therefore, Eq. (4) provides a tighter bound on the WCRT of τ_3 than Eq. (2), while the opposite was true for τ_2. □

In addition to the fact that Eqs. (2) and (4) are incomparable, there might be task sets for which both equations overestimate the WCRT. One such example is given below.

Example 2. *Consider the same three tasks than in Example 1. As explained in Section III-B, the extra interference caused by the self-suspending behavior of τ_1 can be safely modeled by a release jitter equal to $R_1 - S_1 = 5$. Similarly, the extra interference caused by the self-suspension of τ_2 can be modeled by a blocking time equal to $\min(C_2, S_2) = 1$ (see Section III-C). Hence, the WCRT of τ_3 is upper bounded by the smallest $t > 0$ such that $t = 4 + 1 + \left\lceil \frac{t+5}{10} \right\rceil 1 + \left\lceil \frac{t}{19} \right\rceil 6$, which turns out to be 33. This bound on the WCRT is smaller than the estimates obtained with both Eqs. (2) and (4) (see Example 1).* □

Example 2 shows that a tighter bound on the WCRT of a task can be obtained by combining the properties of the analyses discussed in both Section III-B and III-C. Therefore, in this paper, we derive a response time analysis that draws inspiration from both Eqs. (2) and (4), combining the best of each of them. As further proven in Section VI, the resulting schedulability test dominates all the tests discussed in Section III.

V. A Unifying Analysis Framework

As already discussed in Section III, one can greedily convert the suspension time of task τ_k in computation time.

Let τ'_k be this converted version of task τ_k, i.e., $\tau'_k = (C_k + S_k, 0, D_k, T_k)$. Suppose that R'_k is the worst-case response time of τ'_k in the modified task set $\{\tau_1, \tau_2, \ldots, \tau_{k-1}, \tau'_k\}$. It was already shown in previous works, e.g., Lemma 3 in [18], that R'_k is a safe upper bound on the worst-case response time of task τ_k in the original task set.

Note that in all this section, we implicitly assume that $R_i \leq D_i, \forall \tau_i \mid 1 \leq i \leq k-1$. Our key result in this paper is the following theorem:

Theorem 1. *Suppose that $R_k \leq T_k$, then for any arbitrary vector assignment $\vec{x} = (x_1, x_2, \ldots, x_{k-1})$, in which x_i is either 0 or 1, the worst-case response time R_k of τ_k is upper bounded by the minimum t larger than 0 such that*

$$C_k + S_k + \sum_{i=1}^{k-1} \left\lceil \frac{t + Q_i^{\vec{x}} + (1 - x_i)(R_i - C_i)}{T_i} \right\rceil C_i \leq t \quad (5)$$

where $Q_i^{\vec{x}} \stackrel{\text{def}}{=} \sum_{j=i}^{k-1} (S_j \times x_j)$.

One can directly derive the following schedulability test from Theorem 1.

Corollary 1. *If there is a vector $\vec{x} = (x_1, x_2, \ldots, x_{k-1})$ with $x_i \in \{0, 1\}$, such that*

$$\exists t | 0 < t \leq D_k,$$
$$C_k + S_k + \sum_{i=1}^{k-1} \left\lceil \frac{t + Q_i^{\vec{x}} + (1 - x_i)(R_i - C_i)}{T_i} \right\rceil C_i \leq t \quad (6)$$

where $Q_i^{\vec{x}} \stackrel{\text{def}}{=} \sum_{j=i}^{k-1} (S_j \times x_j)$, then the constrained-deadline task τ_k is schedulable under fixed-priority.

The proof of correctness of Theorem 1 and hence Corollary 1 is provided in Section V-A. Moreover, we will later prove in Section VI, that Corollary 1 in fact dominates all the analyses discussed in Section III.

We now use the same example as in Section IV, to demonstrate how Corollary 1 can be applied.

Example 3. *Consider the same three tasks used in Examples 1 and 2, i.e., $\tau_1 = (4, 5, 10, 10)$, $\tau_2 = (6, 1, 19, 19)$ and $\tau_3 = (4, 0, 50, 50)$. There are four possible vector assignments $\vec{x}$ when considering the schedulability of task τ_3 with Corollary 1:*

Case 1. *$\vec{x} = (0, 0)$: In this case, Theorem 1 states that R_k is upper bounded by the minimum t under $0 < t \leq D_3$ such that $4 + \left\lceil \frac{t+5}{10} \right\rceil 4 + \left\lceil \frac{t+9}{19} \right\rceil 6 \leq t$. Note that this equation is identical to the schedulability test discussed in Section III-B, and hence, as shown in Example 1, we get that $R_k \leq 42$.*

Case 2. *$\vec{x} = (0, 1)$: In this case, Theorem 1 states that R_k is upper bounded by the minimum t under $0 < t \leq D_3$ that satisfies $4 + \left\lceil \frac{t+6}{10} \right\rceil 4 + \left\lceil \frac{t+1}{19} \right\rceil 6 \leq t$. As a solution, we get that $R_k \leq 32$.*

Case 3. *$\vec{x} = (1, 0)$: In this case, we look for the minimum t such that $4 + \left\lceil \frac{t+5}{10} \right\rceil \cdot 4 + \left\lceil \frac{t+9}{19} \right\rceil \cdot 6 \leq t$. Hence, we get $R_k \leq 42$.*

Case 4. *$\vec{x} = (1, 1)$: In this case, Theorem 1 states that R_k is upper bounded by the minimum t under $0 < t \leq T_3$ such that $4 + \left\lceil \frac{t+6}{10} \right\rceil \cdot 4 + \left\lceil \frac{t+1}{19} \right\rceil \cdot 6 \leq t$ leading to $R_k \leq 32$.*

Among the above four cases, the tests in Cases 2 and 4 are the tightest. Therefore, by Corollary 1, τ_3 is schedulable under fixed-priority. □

Note also that the upper bound on τ_3's WCRT computed in Example 3, is lower than the WCRT estimate obtained in Example 2. The response time analysis presented in Corollary 1 is therefore tighter than the simple combination of existing analysis techniques proposed in Example 2.

A. Proof of Correctness

We now provide the proof to support the correctness of the response time analysis presented in Theorem 1, whatever the binary values used in vector $\vec{x}$.

Throughout the proof, we consider any arbitrary assignment $\vec{x}$, in which x_i is either 0 or 1. For the sake of clarity, we classify the $k-1$ higher-priority tasks into two sets: $\mathbf{T}_0$ and $\mathbf{T}_1$. A task τ_i is in $\mathbf{T}_0$ if x_i is 0; otherwise, it is in $\mathbf{T}_1$.

Our analysis is also based on very simple properties and lemmas enunciated as follows:

Property 1. *In a preemptive fixed-priority schedule, the lower-priority jobs do not impact the schedule of the higher-priority jobs.*

Lemma 1. *In a preemptive fixed-priority schedule, if the worst-case response time of task τ_i is no more than its period T_i, preventing the release of a job of task τ_i does not affect the schedule of any other job of task τ_i.*

Proof: Since, by assumption, the worst-case response time of task τ_i is no more than its period, any job $\tau_{i,j}$ of task τ_i completes its execution before the release of the next job $\tau_{i,j+1}$. Hence, the execution of $\tau_{i,j}$ does not directly interfere with the execution of any other job of τ_i, which then depends only on the schedule of the higher priority jobs. Furthermore, as stated in Property 1, the removal of $\tau_{i,j}$ has no impact on the schedule of the higher-priority jobs, thereby implying that the other jobs of task τ_i are not affected by the removal of $\tau_{i,j}$. ■

With the above properties, we can present the detailed proof of Theorem 1. Since the proof is pretty long, we will also provide some examples to demonstrate the steps in the proof.

Proof of Theorem 1. Consider the modified task set τ' composed of $\{\tau_1, \tau_2, \ldots, \tau_{k-1}, \tau'_k, \tau_{k+1}, \ldots\}$ where $\tau'_k = (C_k + S_k, 0, D_k, T_k)$. Let Ψ be a schedule of τ' such that $R'_k \leq T_k$. Suppose that a job J_k of task τ'_k arrives at time r_k and finishes at time f_k. We prove that Eq. (5) gives us a safe upper bound on $f_k - r_k$ for any job J_k in Ψ.

The proof is built upon the three following steps:

1) We discard all the jobs that do not contribute to the response time of J_k in the schedule Ψ. We follow an inductive strategy by iteratively inspecting the schedule of the higher priority tasks in Ψ, starting with τ_{k-1} until the highest priority task τ_1. At each iteration, a time instant t_j is identified such that $t_j \leq t_{j+1}$ $(1 \leq j < k)$. Then, all the jobs of task τ_j released before t_j are removed from the schedule and, if needed, replaced by an artificial job mimicking the interference caused by the residual workload of task τ_j at time t_j.

2) The final reduced schedule is analyzed so as to characterize the worst-case response time of τ'_k in Ψ.
3) We then prove that the response time analysis in Eq. (5) is indeed an upper bound on the worst-case response time R'_k of τ'_k.

Step 1: Reducing the schedule Ψ

During this step, we iteratively build an artificial schedule Ψ^j from Ψ^{j+1} (with $1 \leq j < k$) so that the response time of τ'_k remains identical. At each iteration, we define t_j for task τ_j in the schedule Ψ^{j+1} (with $j = k-1, k-2, \ldots, 1$) and build Ψ^j by removing all the jobs released by τ_j before t_j.

Basic step (definition of Ψ^k and t_k):

Recall that the job J_k of task τ'_k arrives at time r_k and finishes at time f_k in schedule Ψ. We know by Property 1 that the lower priority tasks $\tau_{k+1}, \tau_{k+2}, \ldots, \tau_n$ do not impact the response time of J_k. Moreover, since we assume that the worst-case response time of task τ'_k is no more than T_k, Lemma 1 proves that removing all the jobs of task τ'_k but J_k has no impact on the schedule of J_k. Therefore, let Ψ^k be a schedule identical to Ψ but removing all the jobs released by the lower priority tasks $\tau_{k+1}, \ldots, \tau_n$ as well as all the jobs released by τ'_k at the exception of J_k. The response time of J_k in Ψ^k is identical to the response time of J_k in Ψ.

We define t_k as the release time of J_k (i.e., $t_k = r_k$).

Induction step (definition of Ψ^j and t_j with $1 \leq j < k$):

Let r_j be the arrival time of the last job released by τ_j before t_{j+1} in Ψ^{j+1} and let J_j denote that job. Removing all the jobs of task τ_j arrived before r_j has no impact on the schedule of any other job released by τ_j (Lemma 1) or any higher priority job released by $\tau_1, \ldots, \tau_{j-1}$ (Property 1). Moreover, because by the construction of Ψ^{j+1}, no task with a priority lower than τ_j executes jobs before t_{j+1} in Ψ^{j+1}, removing the jobs released by τ_j before t_{j+1} does not impact the schedule of the jobs of $\tau_{j+1}, \ldots, \tau_k$. Therefore, we can safely remove all the jobs of task τ_j arrived before r_j without impacting the response time of J_k. Two cases must then be considered:

(a) $\tau_j \in \mathbf{T}_1$. In this case, we analyze two different subcases:
- J_j did not complete its execution by t_{j+1}. For such a case, t_j is set to r_j and hence Ψ^j is built from Ψ^{j+1} by removing all the jobs released by τ_j before r_j.
- J_j completed its execution before or at t_{j+1}. By Lemma 1 and Property 1, removing all the jobs of task τ_j arrived before t_{j+1} has no impact on the schedule of the higher-priority jobs (jobs released by $\tau_1, \ldots, \tau_{j-1}$) and the jobs of τ_j released after or at t_{j+1}. Moreover, because no task with lower priority than τ_j executes jobs before t_{j+1} in Ψ^{j+1}, removing the jobs released by τ_j before t_{j+1} does not impact the schedule of the jobs of $\tau_{j+1}, \ldots, \tau_k$. Therefore, t_j is set to t_{j+1} and Ψ^j is generated by removing all the jobs of task τ_j arrived before t_{j+1}. The response time of J_k in Ψ^j thus remains unchanged in comparison to its response time in Ψ^{j+1}.

(b) $\tau_j \in \mathbf{T}_0$. For such a case, we set t_j to t_{j+1}. Let c_j^* be the remaining execution time for the job of task τ_j at time t_j. By definition, $c_j^* \geq 0$, and we also know that c_j^* is at most C_j. Since by the construction of Ψ^j, all the jobs of τ_j released before t_j are removed, the job of task τ_j arrived at time r_j ($< t_j$) is replaced by a new job released at time t_j with execution time c_j^* and the same priority than τ_j. Clearly, this has no impact on the execution of any job executed after t_j and thus on the response time of J_k. The remaining execution time c_j^* of τ_j at time t_j is called the *residual workload* of task τ_j in the rest of the proof.

This iterative process is repeated until producing Ψ^1. The procedures are well-defined and it is therefore guaranteed that Ψ^1 can be constructed. Note that after each iteration, the number of jobs considered in the resulting schedule has been reduced, yet without affecting the response time of J_k.

Example 4. *Consider the 4 tasks $\tau_1 = (1, 1, 6, 6)$, $\tau_2 = (1, 6, 10, 10)$, $\tau_3 = (4, 1, 18, 18)$ and $\tau_4 = (5, 0, 20, 20)$.*

Figure 1 depicts a possible schedule of those tasks. We assume that the first job of task τ_1 arrives at time $4+\epsilon$ with a very small $\epsilon > 0$. The first job of task τ_2 suspends itself from time 0 to time $5+\epsilon$, and is blocked by task τ_1 from time $5+\epsilon$ to time $6+\epsilon$. After some very short computation with ϵ amount of time, the first job of task τ_2 suspends itself again from time $6+2\epsilon$ to 7.

In the schedule illustrated in Figure 1, f_4 is $20-\epsilon$. We define t_4 as 7. Then, we set t_3 to 6. When considering task τ_2, since it belongs to $\mathbf{T}_0$, we greedily set t_2 to $t_3 = 6$ and the residual workload c_2^* is 1. Then, t_1 is set to $4+\epsilon$. In the above schedule, the idle time from $4+\epsilon$ to $20-\epsilon$ is at most $2 = S_1 + S_3$. We have to further consider one job of task τ_2 arrived before time t_1 with execution time C_2.

Lemma 2. *Let σ_j be the amount of time during which the processor remains idle within $[t_j, t_{j+1})$ in Ψ^1. It holds that $\sum_{j=1}^{i-1} \sigma_j \leq \sum_{j=1}^{i-1} x_j S_j$.*

Proof: If $t_j = t_{j+1}$, which is the case when $x_j = 0$ (see Case (b) of the reduction process), then σ_j is obviously equal to 0.

If $t_j \neq t_{j+1}$, then we are in subcase 1 of Case (a) of the schedule reduction process (i.e., when $x_j = 1$). In that case, $t_j = r_j$ and the job J_j did not complete its execution yet. Therefore, the amount of time during which the processor may remain idle within $[t_j, t_{j+1})$ is at most S_j time units.

From those two cases, it results that $\sum_{j=1}^{i-1} \sigma_j = \sum_{j=1}^{i-1} x_j \sigma_j \leq \sum_{j=1}^{i-1} x_j S_j$. ■

Step 2: Analyzing the final reduced schedule Ψ^1

We now analyze the properties of the final schedule Ψ^1 in which all the unnecessary jobs have been removed. The proof is based on the fact that for any interval $[t_1, t)$, there is

$$\text{idle}(t_1, t) + \text{exec}(t_1, t) = (t - t_1) \tag{7}$$

where $\text{exec}(t_1, t)$ is the amount of time during which the processor executes tasks within $[t_1, t)$, and $\text{idle}(t_1, t)$ is the amount of time during which the processor remains idle within the interval $[t_1, t)$.

From Lemma 2, it holds that

$$\text{idle}(t_1, t) \leq \sum_{j=1}^{i-1} x_j S_j. \tag{8}$$

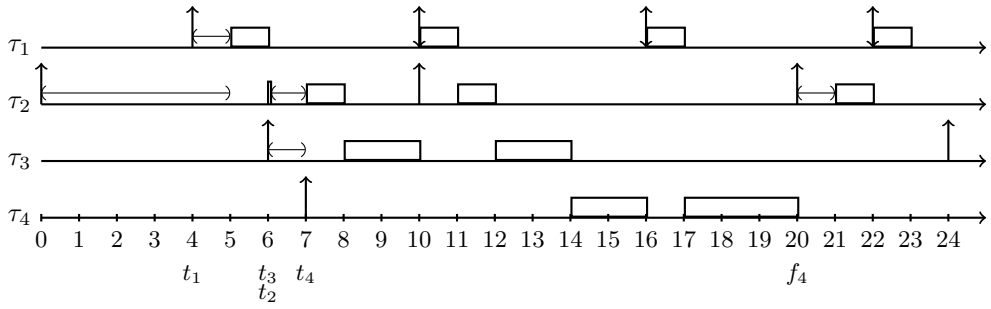


Fig. 1: An illustrative example of Step 1 in the proof of Theorem 1.

Because there is no job released by lower priority tasks than τ'_k in Ψ^1, we only focus on the execution patterns of the tasks $(\tau_1, \tau_2, \ldots, \tau_{k-1}, \tau'_k)$. According to Step 1, we should consider two cases:

- $\tau_j \in \mathbf{T}_1$. This corresponds to Case (a) in Step 1, which tells us that there is no job of task τ_j arrived before t_j in Ψ^1. In this case, for any $\Delta \geq 0$, the workload $W_j^1(\Delta)$, executed by τ_j on the processor from t_j to $t_j+\Delta$ is upper bounded by
$$W_j^1(\Delta) = \left\lfloor \frac{\Delta}{T_j} \right\rfloor C_j + \min\left\{ \Delta - \left\lfloor \frac{\Delta}{T_j} \right\rfloor T_j, C_j \right\}. \quad (9)$$
- $\tau_j \in \mathbf{T}_0$. This corresponds to case (b) in Step 1, which tells us that there might be a job arrived before t_j with a residual workload c_j^* at time t_j. Let c_j^* be the residual workload of task τ_j at t_j. Since by assumption τ_j respects all its deadlines, the absolute deadline of the job of τ_j active at t_j must be at least $t_j + c_j^*$; otherwise that job would miss its deadline. Therefore, the earliest arrival time of task τ_j arriving *strictly after* t_j is at least $t_j + (T_j - D_j + c_j^*)$. For notational brevity, let ρ_j be $(T_j - D_j + c_j^*)$. In this case, for any $\Delta \geq 0$ and $c_j^* > 0$, the workload $\widehat{W}_j^0(\Delta, c_j^*)$ executed by τ_j from t_j to $t_j+\Delta$ is upper bounded by
$$\widehat{W}_j^0(\Delta, c_j^*) = \begin{cases} \Delta & \text{if } \Delta \leq c_j^* \\ c_j^* & \text{if } c_j^* < \Delta \leq \rho_j \\ c_j^* + W_j^1(\Delta - \rho_j) & \text{otherwise.} \end{cases} \quad (10)$$
It is easy to see that Eq. (10) is maximized when c_j^* is maximum, that is, when $c_j^* = C_j$. It results that for all $\Delta \geq 0$, we have $\widehat{W}_j^0(\Delta, C_j) \geq \widehat{W}_j^0(\Delta, c_j^*)$. For the sake of notational brevity, for the rest of this proof, we use the notation $W_j^0(\Delta)$ to denote $\widehat{W}_j^0(\Delta, C_j)$.

Summing the workloads of the tasks in $\mathbf{T}_0$ and $\mathbf{T}_1$, we have for $i = 2, 3, \ldots, k-1$ that $\forall t \mid t_{i-1} \leq t < t_i$
$$\text{exec}(t_1, t) \leq \sum_{j=1}^{i-1} x_j \cdot W_j^1(t - t_j) + (1 - x_j) \cdot W_j^0(t - t_j). \quad (11)$$

Injecting Eqs. (8) and (11) in Eq. (7), we have for $i = 2, 3, \ldots, k-1$ that $\forall t \mid t_{i-1} \leq t < t_i$
$$\sum_{j=1}^{i-1} x_j \cdot (W_j^1(t-t_j)+\sigma_j) + (1-x_j) \cdot W_j^0(t-t_j) \geq t - t_1. \quad (12)$$

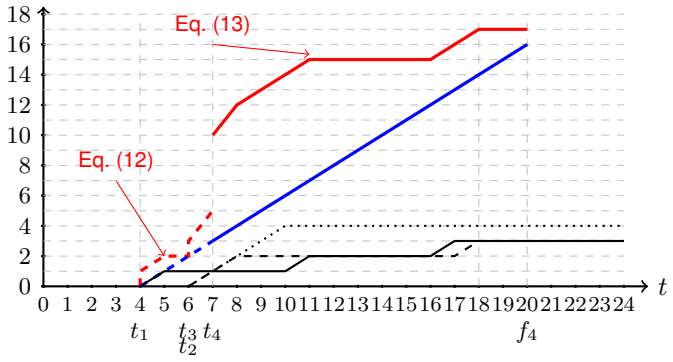


Fig. 2: The workload function for the three higher-priority tasks in Figure 2. Solid black line: $W_1^1(t-t_1)$, Dashed black line: $W_2^0(t-t_2)$, Dotted black line: $W_3^1(t-t_3)$, where the functions are 0 if $t - t_j < 0$ for $j = 1, 2, 3$, Blue line: $t - t_1$, Red line: left-hand side of Eq. (12) when $t < 7$ and left-hand side of Eq. (13) when $7 \leq t < 20$.

Moreover, since τ'_k does not complete its execution before f_k and because, by definition, τ'_k does not self-suspend, it holds that $\forall t \mid t_k \leq t < f_k$,
$$C'_k + \sum_{j=1}^{k-1} x_j \cdot (W_j^1(t-t_j)+\sigma_j) + (1-x_j) \cdot W_j^0(t-t_j) > t - t_1. \quad (13)$$

Example 5. *Consider the same 4 tasks as in Example 4, for which a possible schedule was depicted in Figure 1. We have $\sigma_1 = 1$, $\sigma_2 = 0$ and $\sigma_3 = 1$. The corresponding functions $W_1^1(t-t_1)$, $W_2^0(t-t_2)$, $W_3^1(t-t_3)$ are illustrated in Figure 2. As can be seen in the figure, the inequalities of Eqs.* (12) *and* (13) *clearly hold.*

Step 3: Creating a Safe Response-Time Upper Bound

This step constructs a safe response-time analysis based on the conditions specified by Eqs. (12) and (13). To do so, we construct another release pattern which moves t_i to t_i^* for $i = 2, 3, \ldots, k$ such that $t_i^* \leq t_i$ and the corresponding conditions in Eqs. (12) and (13) will become worse when we use t_i^*. We start the procedure as follows:

- Initial step: Let t_1^* be t_1.
- Iterative steps ($i = 2, 3, \ldots, k$): Let t_i^* be $t_{i-1}^* + x_{i-1} \cdot \sigma_{i-1}$.

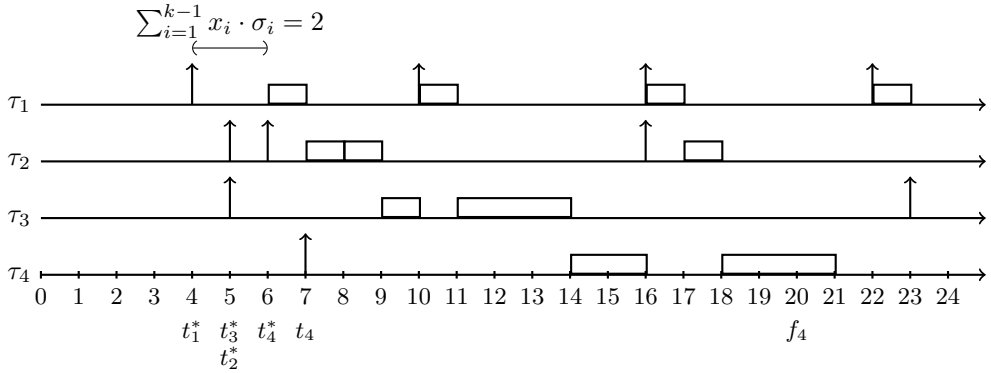


Fig. 3: An illustrative example of Step 3 in the proof of Theorem 1 based on an *imaginary* schedule.

This results in $t_i^* \leq t_i$ for $i = 2, 3, \ldots, k$. Moreover, by definition, t_j^* is $t_1^* + \sum_{i=1}^{j-1} x_i \cdot \sigma_i$ for $j = 2, 3, \ldots, k$. For any task τ_j in $\mathbf{T}_1$, $\forall \Delta \geq 0$, since $t_j \geq t_j^*$, we have

$$W_j^1(\Delta) \leq W_j^1(\Delta + (t_j - t_j^*)). \tag{14}$$

For any task τ_j in $\mathbf{T}_0$, $\forall \Delta \geq 0$, since $t_j \geq t_j^*$, we have

$$W_j^0(\Delta) \leq W_j^0(\Delta + (t_j - t_j^*)). \tag{15}$$

Therefore, for any $j = 1, 2, \ldots, k-1$, the contribution $W_j^1(t - t_j) \leq W_j^1(t - t_j^*)$ and $W_j^0(t - t_j) \leq W_j^0(t - t_j^*)$ for any $t \geq t_j$. Putting these into Eqs. (12) $\forall t \mid t_k^* \leq t < t_k$ leads to

$$\sum_{j=1}^{k-1} x_j \cdot (W_j^1(t - t_j^*) + \sigma_j) + (1 - x_j) \cdot W_j^0(t - t_j^*) \geq t - t_1,$$

$$\Rightarrow \sum_{j=1}^{k-1} x_j \cdot W_j^1(t - t_j^*) + (1 - x_j) \cdot W_j^0(t - t_j^*) \geq t - t_k^*. \tag{16}$$

Similarly, putting these into Eqs. (13) $\forall t \mid t_k \leq t < f_k$ leads to

$$C_k' + \sum_{j=1}^{k-1} x_j \cdot W_j^1(t - t_j^*) + (1 - x_j) \cdot W_j^0(t - t_j^*) > t - t_k^*. \tag{17}$$

By the fact that $C_k' \geq C_k > 0$, we can unify the above inequalities in Eq. (16) and Eq. (17) as follows: $\forall t \mid t_k^* \leq t < f_k$

$$C_k' + \sum_{j=1}^{k-1} x_j \cdot W_j^1(t - t_j^*) + (1 - x_j) \cdot W_j^0(t - t_j^*) > t - t_k^*. \tag{18}$$

By the definition of t_j^*, $\forall t \mid t_k^* \leq t < f_k$, we have $t - t_j^* = t - t_k^* + \sum_{\ell=j}^{k-1} x_\ell \sigma_\ell$ for every $j = 1, 2, \ldots, k-1$. Therefore, we know that $W_j^1(t - t_j^*) \leq \left\lceil \frac{t - t_k^* + \sum_{\ell=j}^{k-1} x_\ell \sigma_\ell}{T_j} \right\rceil C_j$ for task τ_j in $\mathbf{T}_1$. Moreover, $\forall t \mid t_k^* \leq t < f_k$, we have $W_j^0(t - t_j^*) \leq \left\lceil \frac{t - t_k^* + \sum_{\ell=j}^{k-1} x_\ell \sigma_\ell + (1 - x_j)(D_j - C_j)}{T_j} \right\rceil C_j$ for task τ_j in $\mathbf{T}_0$. Therefore, we can conclude that $\forall t \mid t_k^* \leq t < f_k$

$$C_k' + \sum_{j=1}^{k-1} \left\lceil \frac{t - t_k^* + X_j + (1 - x_j)(D_j - C_j)}{T_j} \right\rceil C_j > t - t_k^*, \tag{19}$$

where X_j is $\sum_{\ell=j}^{k-1} x_\ell \sigma_\ell$. We replace $t - t_k^*$ with θ. The above inequation implies that the minimum θ with $\theta > 0$ such that $C_k' + \sum_{j=1}^{k-1} \left\lceil \frac{\theta + X_j + (1 - x_j)(D_j - C_j)}{T_j} \right\rceil C_j = \theta$ is larger than or equal to $f_k - t_k^* \geq f_k - t_k$.

However, the above condition requires the knowledge of σ_i. It is straightforward to see that $\sum_{j=1}^{k-1} \left\lceil \frac{\theta + X_j + (1 - x_j)(D_j - C_j)}{T_j} \right\rceil C_j$ reaches the worst case if X_j is the largest. Since X_j is upper bounded by $Q_j^{\vec{x}}$ defined in Theorem 1, we reach the conclusion.

Example 6. *This can be demonstrated in Figure 3 based on the previous example in Figure 1. Figure 3 provides the imaginary workload and an imaginary execution plan based on the test behind the condition in Eq.* (18). Note that this is not an actual schedule since task τ_2 is artificially alerted to release two jobs within a short time interval. This is only for illustrative purposes. *For such a case,* $t_1^* = 4, t_2^* = 5, t_3^* = 5$, *and* $t_4^* = 6$. *The two idle time units are used between time* 4 *and time* 6. *The accumulated workload is then started to be executed at time* 6 *and the processor does not idle after time* 6. *Over here, we see that two jobs of task* τ_2 *are executed back to back from time 7 to time 9. As shown in the imaginary schedule in Figure 3, the processor is busy executing the workload from time* 6 *to time* 21, *which is more pessimistic than the actual schedule in Figure 1. The conclusion we have in the final statement of the theorem is that* $20 - 7 = f_k - r_k \leq 21 - 6$.

□

VI. Dominance over the State of the Art

In this section, we prove that the schedulability test presented in Corollary 1 dominates all the existing tests in the state-of-the-art, in the sense that if a task set is deemed schedulable by either of the tests presented in Section III, then it is also deemed schedulable by Corollary 1.

Lemma 3. *The schedulability test of task* τ_k *provided by Eq.* (4) *dominates that of Eq.* (1).

Proof: It is straightforward to see that

$$\begin{aligned} & C_k + S_k + \sum_{i=1}^{k-1} \left\lceil \frac{t}{T_i} \right\rceil (C_i + S_i) \\ \geq & C_k + S_k + \sum_{i=1}^{k-1} S_i + \sum_{i=1}^{k-1} \left\lceil \frac{t}{T_i} \right\rceil C_i \\ \geq & C_k + S_k + \sum_{i=1}^{k-1} \min(C_i, S_i) + \sum_{i=1}^{k-1} \left\lceil \frac{t}{T_i} \right\rceil C_i \end{aligned}$$

and by using the definition of B_k (i.e., Equation (3)), we get

$$C_k + S_k + \sum_{i=1}^{k-1} \left\lceil \frac{t}{T_i} \right\rceil (C_i + S_i) \geq C_k + B_k + \sum_{i=1}^{k-1} \left\lceil \frac{t}{T_i} \right\rceil C_i$$

Therefore, Eq. (4) will always have a solution which is smaller than or equal to the solution of Eq. (1). This proves the lemma. ■

Lemma 4. *The schedulability test presented in Corollary 1 dominates the schedulability test provided by Eq.* (2)*.*

Proof: Consider the case where $x_1 = x_2 = \cdots = x_{k-1} = 0$. Eq. (6) becomes identical to Eq. (2) for this particular vector assignment. Therefore, if Eq. (2) deems a task set as being schedulable, so does Corollary 1. This proves the lemma. ■

Lemma 5. *The schedulability test presented in Corollary 1 dominates the schedulability test provided by Eq.* (4)*.*

Proof: In this proof, we first transform the worst-case response time analysis presented in Corollary 1 in a more pessimistic analysis. We then prove that this more pessimistic version of Corollary 1 provides the same solution as Eq. (4), which then proves the lemma.

Since $Q_i^{\vec{x}} \stackrel{\text{def}}{=} \sum_{j=i}^{k-1} S_j \times x_j$, it holds that $Q_i^{\vec{x}} \leq Q_1^{\vec{x}}$ for $i = 1, 2, \ldots, k-1$. It follows that

$$\begin{aligned} & C_k + S_k + \sum_{i=1}^{k-1} \left\lceil \frac{t + Q_i^{\vec{x}} + (1 - x_i)(R_i - C_i)}{T_i} \right\rceil \\ \stackrel{(Q_i^{\vec{x}} \leq Q_1^{\vec{x}})}{\leq} & C_k + S_k + \sum_{i=1}^{k-1} \left\lceil \frac{t + Q_1^{\vec{x}} + (1 - x_i)(R_i - C_i)}{T_i} \right\rceil \\ \stackrel{(R_i \leq D_i \leq T_i)}{\leq} & C_k + S_k + \sum_{i=1}^{k-1} \left\lceil \frac{t + Q_1^{\vec{x}} + (1 - x_i)T_i}{T_i} \right\rceil \\ \stackrel{(x_i \in \{0,1\})}{=} & C_k + S_k + \sum_{i=1}^{k-1} (1 - x_i)C_i + \sum_{i=1}^{k-1} \left\lceil \frac{t + Q_1^{\vec{x}}}{T_i} \right\rceil \end{aligned}$$

Therefore, the smallest positive value t such that

$$C_k + S_k + \sum_{i=1}^{k-1} (1 - x_i)C_i + \sum_{i=1}^{k-1} \left\lceil \frac{t + Q_1^{\vec{x}}}{T_i} \right\rceil \leq t \qquad (20)$$

is always larger than or equal to the solution of Eq. (6).

Subtituting $(t + Q_1^{\vec{x}})$ by θ in Eq. (20), we get that R_k is upper bounded by the minimum value $(\theta - Q_1^{\vec{x}})$ greater than 0 (and therefore by the smallest $\theta > 0$) such that

$$\begin{aligned} & C_k + S_k + \sum_{i=1}^{k-1} (1 - x_i)C_i + \sum_{i=1}^{k-1} \left\lceil \frac{\theta}{T_i} \right\rceil \leq \theta - Q_1^{\vec{x}} \\ \Leftrightarrow \; & C_k + S_k + Q_1^{\vec{x}} + \sum_{i=1}^{k-1} (1 - x_i)C_i + \sum_{i=1}^{k-1} \left\lceil \frac{\theta}{T_i} \right\rceil \leq \theta \\ \Leftrightarrow \; & C_k + S_k + \sum_{i=1}^{k-1} (x_i S_i + (1 - x_i)C_i) + \sum_{i=1}^{k-1} \left\lceil \frac{\theta}{T_i} \right\rceil C_i \leq \theta. \end{aligned} \qquad (21)$$

Now, consider the particular vector assignment $\vec{x}$ in which

$$x_i = \begin{cases} 1 & \text{if } S_i \leq C_i \\ 0 & \text{otherwise,} \end{cases}$$

for $i = 1, 2, \ldots, k-1$. By the definition of B_k (i.e., Eq. (3)), we get that

$$B_k = S_k + \sum_{i=1}^{k-1} \min(C_i, S_i) = S_k + \sum_{i=1}^{k-1} (x_i S_i + (1 - x_i)C_i)$$

Eq. (21) thus becomes identical to Eq. (4). Therefore, if Eq. (4) deems a task set as being schedulable, so does Corollary 1. ■

Theorem 2. *The schedulability test presented in Corollary 1 dominates the schedulability tests provided by Equations* (1)*,* (2)*, and* (4)*.*

Proof: It is a direct application of Lemmas 3, 4 and 5. ■

As a corollary of this theorem, it directly follows that all the response time analyses discussed in Section III are in fact correct. This provides the first proof of correctness for Eq. (4), which was initially presented in [17] but never proven correct.

Theorem 3. *The schedulability tests provided by Eqs* (1)*,* (2)*, and* (4) *are all correct.*

Proof: It directly results from the two following facts,

(i) by Theorem 2, the schedulability test presented in Corollary 1 dominates the schedulability tests provided by Equations (1), (2), and (4);
(ii) as proven in Section V-A, Corollary 1 is correct.

■

VII. Linear Approximation

To test the schedulability of a task τ_k, Corollary 1 implies to test all the possible vector assignments $\vec{x} = (x_1, x_2, \ldots, x_{k-1})$. Therefore, 2^{k-1} possible combinations must therefore be tested, implying exponential time complexity. In this section, we thus provide a solution to reduce the time complexity associated to Corollary 1. Indeed, using a linear approximation of the test in Eq. (6), a good vector assignment can be derived in linear time.

By the definition of the ceiling operator, it holds that:

$$\begin{aligned}
&C_k + S_k + \sum_{i=1}^{k-1} \left\lceil \frac{t + \sum_{\ell=i}^{k-1} x_\ell S_\ell + (1-x_i)(R_i - C_i)}{T_i} \right\rceil C_i \\
\leq &C_k + S_k + \sum_{i=1}^{k-1} \left(\frac{t + \sum_{\ell=i}^{k-1} x_\ell S_\ell + (1-x_i)(R_i - C_i)}{T_i} + 1 \right) C_i \\
= &C_k + S_k + \sum_{i=1}^{k-1} \left(U_i \cdot t + C_i + U_i(1-x_i)(R_i - C_i) + U_i \sum_{\ell=i}^{k-1} x_\ell S_\ell \right)
\end{aligned} \tag{22}$$

Moreover, using the simple algebra property that for any two vectors $\vec{a}$ and $\vec{b}$ of size $(k-1)$ there is $\sum_{i=1}^{k} a_i \sum_{j=i}^{k} b_j = \sum_{j=1}^{k} b_j \sum_{i=1}^{j} a_i$, we get that $\sum_{i=1}^{k-1} U_i \sum_{\ell=i}^{k-1} x_\ell S_\ell = \sum_{i=1}^{k-1} x_i S_i \sum_{\ell=1}^{i} U_\ell$. Hence, injecting this last expression in Eq. (22), it holds that

$$\begin{aligned}
&C_k + S_k + \sum_{i=1}^{k-1} \left\lceil \frac{t + \sum_{\ell=i}^{k-1} x_\ell S_\ell + (1-x_i)(R_i - C_i)}{T_i} \right\rceil C_i \\
\leq &C_k + S_k + \sum_{i=1}^{k-1} \left(U_i \cdot t + C_i + U_i(1-x_i)(R_i - C_i) + x_i S_i \sum_{\ell=1}^{i} U_\ell \right)
\end{aligned}$$

It results that the minimum positive value for t such that

$$C_k + S_k + \sum_{i=1}^{k-1} \left(U_i \cdot t + C_i + U_i(1-x_i)(R_i - C_i) + x_i S_i \sum_{\ell=1}^{i} U_\ell \right) \leq t \tag{23}$$

is an upper bound on the WCRT of τ_k.

Observing Eq. (23), the contribution of x_i can be individually determined as $U_i(R_i - C_i)$ when x_i is 0 or $S_i(\sum_{\ell=1}^{i} U_\ell)$ when x_i is 1. Therefore, whether x_i should be set to 0 or 1 can be decided by individually comparing the two constants $U_i(R_i - C_i)$ and $S_i(\sum_{\ell=1}^{i} U_\ell)$. Eq. (23) is therefore minimized when $x_i = 1$ if $U_i(R_i - C_i) > S_i(\sum_{\ell=1}^{i} U_\ell)$ and when $x_i = 0$ otherwise. We denote the resulting vector by $\vec{x}^{lin}$, where, for each higher-priority task τ_i,

$$x_i^{lin} = \begin{cases} 1 & \text{if } U_i(R_i - C_i) > S_i(\sum_{\ell=1}^{i} U_\ell) \\ 0 & \text{otherwise} \end{cases} \tag{24}$$

The following properties directly follow.

Property 2. *For any $t > 0$, the vector assignment $\vec{x}^{lin}$ minimizes the solution to Eq.* (23) *among all 2^{k-1} possible vector assignments.*

Theorem 4. *Let $rbf_k^{lin}(t, \vec{x})$ bet the left hand side of Eq.* (23)*. Task τ_k is schedulable under fixed-priority if*

$$rbf_k(D_k, \vec{x}^{lin}) \leq D_k. \tag{25}$$

Proof: It directly follows from Corollary 1 and the fact that, by construction, Eq. (23) upper bounds Eq. (5). ■

Property 3. *The time complexity of both deriving $\vec{x}^{lin}$ and testing Eq. (23) is $O(k)$.*

VIII. Experiments

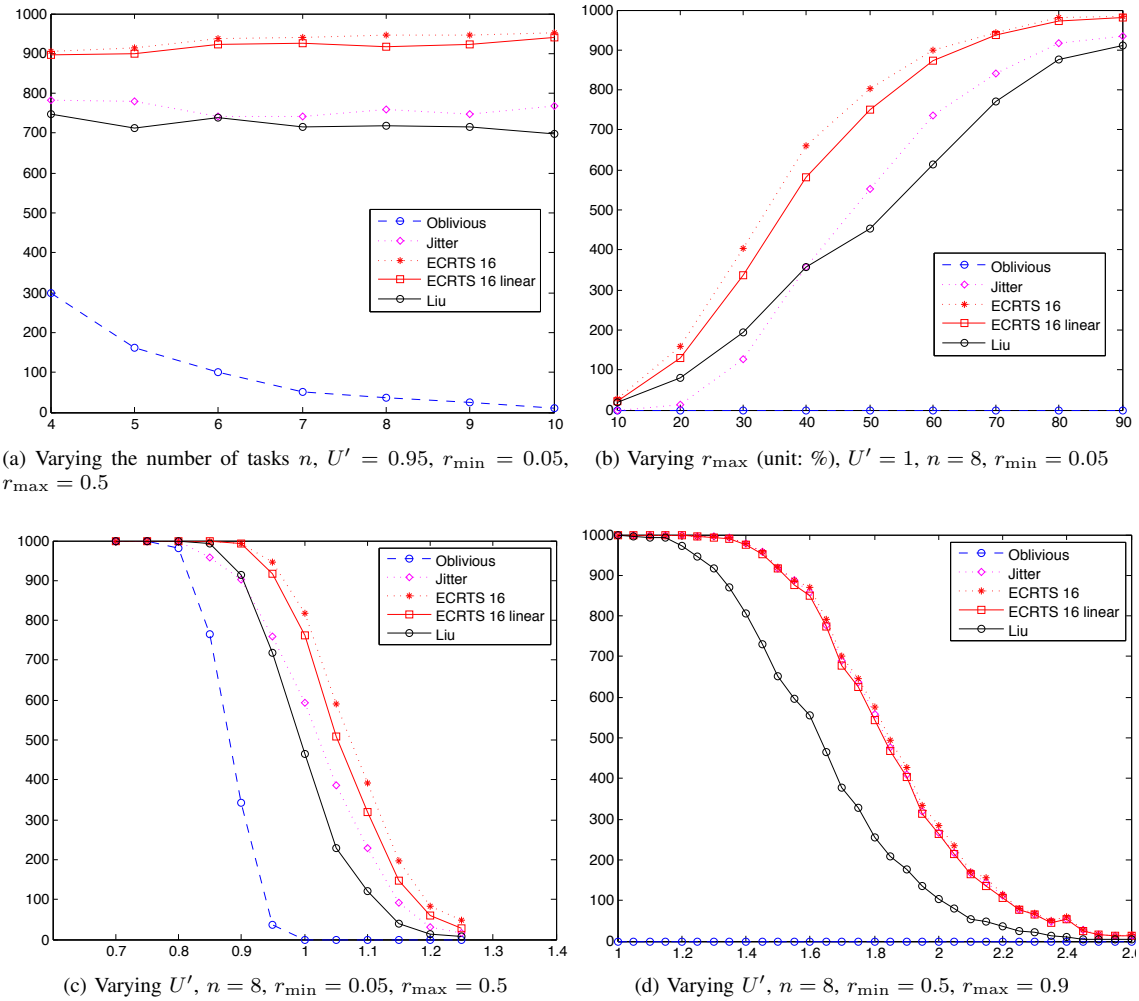
In this section, we present experiments conducted on randomly generated task sets. Five schedulability tests for dynamic self-suspending tasks are compared, namely, the suspension oblivious approach (Section III-A), the modeling of suspension as release jitter (Section III-B), the analysis proposed by Jane W.S. Liu and proven correct in this paper that models the suspension as a blocking term (Section III-C), the generic framework of Corollary 1 (called ECRTS 16 in the plots) and the schedulability test of Theorem 1 based on the vector defined in Eq. (24) in Section VII (called ECRTS 16 linear in the plots). In those experiments, the tasks are assumed to be scheduled with rate monotonic and have implicit deadlines (i.e., $D_i = T_i$).

The task sets were generated using the `randfixedsum` algorithm presented in [7]. Let C'_i denote the sum of C_i and S_i (i.e., $C'_i \stackrel{\text{def}}{=} C_i + S_i$). That is, C'_i is the WCET of a task for which the suspension time would be considered as execution time like it is the case in the test of Section III-A. The modified utilization of τ_i is then given by $U'_i \stackrel{\text{def}}{=} \frac{C'_i}{T_i}$ and the total modified utilization is $U' \stackrel{\text{def}}{=} \sum_{i=1}^{n} U'_i$. The task generator uses the `randfixedsum` algorithm to generate n values of U'_i (one for each task) with total modified utilization U'. A period T_i is then randomly generated from a uniform distribution spanning from 100 to 10000. The resulting value $C'_i = U'_i \times T_i$ is then divided in the two components C_i and S_i using a random ratio r_i obtained from a uniform distribution between a value $r_{\min}$ and $r_{\max}$ depending of the specific experiment performed. That is, $S_i \stackrel{\text{def}}{=} r_i \times C'_i$ and $C_i = (1 - r_i) \times C'_i$.

Each point in the plots of Figure 4 represents the number of task sets that were deemed schedulable by the respective algorithm over 1000 different experiments.

Four different types of experiments are reported in this paper. The first one is presented in Figure 4a. It presents the evolution of the number of task sets deemed schedulable when the number of self-suspending tasks increases. The number of tasks n is varied from 4 to 10 for a total modified utilization U' of 0.95. As can be seen in Figure 4a, at the exception of the suspension oblivious analysis, the performances of the tests are barely influenced by the number of tasks. In fact, the number of task sets found schedulable by the test of Corollary 1 and the linear test of Section VII slightly increases with the number of tasks. It is the opposite behavior than the suspension oblivious approach. One can already conclude from this plot that the tests developed in this paper perform way better than the state-of-the-art. Furthermore, the difference between the performances of the complete test of Corollary 1 and its linear version is quite small, thereby making the linear test a practical and useful analysis.

The second experiment is presented in Figure 4b and shows the evolution of the performances of the tests with respect to the length of the total suspension time of a task when the total modified utilization U' and the number of tasks are kept constant. The value of $r_{\max}$ is then varied from 10% to 90%, hence increasing the number of tasks with high suspension times. The value $r_{\min}$ is kept constant at 5%, so as to keep a certain diversity in the suspension behavior of each task. As expected, the suspension oblivious approach does not accept any task set since the total modified utilization is equal to



(a) Varying the number of tasks n, $U' = 0.95$, $r_{\min} = 0.05$, $r_{\max} = 0.5$

(b) Varying $r_{\max}$ (unit: %), $U' = 1$, $n = 8$, $r_{\min} = 0.05$

(c) Varying U', $n = 8$, $r_{\min} = 0.05$, $r_{\max} = 0.5$

(d) Varying U', $n = 8$, $r_{\min} = 0.5$, $r_{\max} = 0.9$

Fig. 4: Number of schedulable task sets over 1000 randomly generated task sets.

100%. For the other tests however, the number of schedulable task sets increases when the suspension times become larger. Indeed, the actual workload, which accounts only for the WCET C_i, decreases when S_i increases. Again, one can see the improvement of the tests of this paper over the state-of-the-art. Interestingly, one can also witnesses the incomparability of the jitter based and the blocking based schedulability tests. The jitter based test performs better for large blocking times while the blocking based test has better results for lower blocking times.

The last two plots (Figures 4c and 4d), present the results obtained when the total modified utilization increases but the distribution of suspension times and the number of tasks remain identical. As expected, the number of schedulable task sets decreases when the utilization increases. Again, the incomparability of the jitter- and blocking-based tests can be seen. The test based on blocking usually performs better for lower utilizations. The improvement of Corollary 1 over the state-of-the-art is still high when suspension times are in average smaller than the execution times of the tasks (see Figures 4c). However, when the suspension time becomes larger than the execution time of the task (see Figures 4d), the release jitter-based test performs almost as well as Corollary 1 since the best vector assignment is usually to set all the x_i to 0 for such cases.

IX. Conclusion

In this paper, we studied the preemptive fixed-priority scheduling of dynamic self-suspending tasks running on a uniprocessor platform. This paper presents a unifying response time analysis framework in Theorem 1 and Corollary 1. We show that this result analytically dominates all the existing analyses presented in Section III, and, by doing such, we also implicitly proved the correctness of all these analyses. Although Corollary 1 requires exponential time complexity, we show that a simpler algorithm presented in Section VII can help accelerate the analysis while outputting good results.

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