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

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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 influences 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.

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INTRODUCTION

The periodic/sporadic task model has been recognized as the basic model for real-time systems with recurring executions. The seminal work by Liu and Layland [24] 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 [17]) to a few hundreds of milliseconds (e.g., offloading computation to GPUs [18, 26]).

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, e.g., [1, 2, 20, 27, 23, 16, 10], in addition to the worst-case execution time  $C_i$  of sporadic task  $\tau_i$ , we have also the worst-case self-suspension time  $S_i$  of task  $\tau_i$ . In the *segmented* self-suspension task model, e.g., [5, 21, 15, 9, 14, 28], the execution behaviour of a job of task  $\tau_i$  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 [20, 27, 1, 2, 16]. The recent report by Chen et al. [11] and the report by Bletsas et al. [4] have shown that several analyses in the state-of-the-art of self-suspending tasks [1, 2, 20, 27] are in fact unsafe. Unfortunately, those misconceptions propagated to several works [32, 6, 30, 19, 13, 7, 31, 22] analyzing the worst-case response time for partitioned multiprocessor real-time locking protocols. Moreover, Liu and Chen in [23] provided a utilization-based schedulability test based

on a hyperbolic-form. Huang et al. [16] explored the priority assignment under the same system model.

Furthermore, one result presented by Jane W. S. Liu in her book "Real-Time Systems" [25, p. 164-165] and implicitly used by Rajkumar, Sha, and Lehoczky [29, p. 267] for analyzing the self-suspending behaviour due to synchronization protocols in multiprocessor systems, was never proven correct.

**Contributions.** The contributions of this paper are as follows:

- We provide a new response analysis framework for dynamic self-suspending sporadic real-time tasks on a uniprocessor platform. The key observation 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 state-ofthe-art results, excluding the flawed ones.
- We prove the correctness of the analysis initially proposed in [25, p. 164-165] and [29, p. 267], which were never proven correct in the state-of-the-art<sup>1</sup>.
- The evaluation results presented in Section 8 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.

A simplified version of the proof of Theorem 1 to support the correctness of [25, p. 164-165] and [29, p. 267] is provided in[8].

We assume a system  $\tau$  composed of n sporadic self-suspending tasks. A sporadic task  $\tau_i$  is released repeatedly, with each such invocation called a job. The  $j^{th}$  job of  $\tau_i$ , 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  $\tau_i$  is assumed to have a worst-case execution time  $C_i$ . Furthermore, a job of task  $\tau_i$  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 have to execute in sequence.

Each task  $\tau_i$  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  $\tau_i$  and  $D_i$  is its relative deadline.  $T_i$  specifies the minimum time between two consecutive job releases of  $\tau_i$ , 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  $r_{i,j+1} \geqslant r_{i,j} + T_i$ . In this paper, we focus on constrained-deadline tasks, for which  $D_i \leqslant T_i$ . The utilization of a task  $\tau_i$  is defined as  $U_i = C_i/T_i$ .

The worst-case response time (WCRT)  $R_i$  of a task  $\tau_i$  is the maximum response time among all its jobs. A schedulability test for a task  $\tau_k$  is therefore to verify whether its worst-case response time is no more than its 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  $\tau_i$  has a higher-priority than task  $\tau_j$  if i < j.

When performing the schedulability analysis of a specific task  $\tau_k$ , 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  $R_i \leq D_i, \forall \tau_i \mid 1 \leq i \leq k-1$ .

To analyze the worst-case response time (or the schedulability) of a task  $\tau_k$ , one usually needs to quantify the worst-case interference exerted by the higher-priority tasks on the execution of any job of task  $\tau_k$ . In the ordinary sequential sporadic real-time task model, i.e., when  $S_i=o$  for every task  $\tau_i$ , the so-called critical instant theorem by Liu and Layland [24] is commonly adopted. That is, the worst-case response time of task  $\tau_k$  (if it is less than or equal to its period) happens for the first job of task  $\tau_k$  when (i)  $\tau_k$  and all the higher-priority tasks release their first jobs synchronously and (ii) all their subsequent jobs are released as early as possible (i.e., with a rate equal to their periods). However, this definition of the critical instant does not hold for self-suspending tasks.

The analysis of self-suspending task systems requires to model the self-suspending behavior of both the task  $\tau_k$  under analysis and the higher priority tasks that interfere with  $\tau_k$ . The techniques employed to model the self-suspension are usually different for  $\tau_k$  and the higher priority tasks. The worst-case for  $\tau_k$  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  $\tau_k$  was converted into computation time (see [16] for more detailed 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  $\tau_k$ . There exist three 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 suspensionoblivious analysis (see Section 3.0.1);
- modeling the suspension as release jitter (see Section 3.1);
- modeling the suspension as blocking time (see Section 3.2).

We later prove in Section 6 that all these approaches are analytically correct.

## 3.0.1 Suspension-Oblivious Analysis

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

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

#### 3.1 MODELING THE SUSPENSION AS RELEASE JITTER

Another approach consists in modeling the impact of the self-suspension  $S_i$  of each higher priority task  $\tau_i$  as release jitter. Several works in the state-of-the-art [1, 2, 20, 27] upper bounded the release jitter with  $S_i$ . However, it has been recently shown in [4] that this upper bound is unsafe and the release jitter of task  $\tau_i$  can in fact be larger than  $S_i$ .

Nevertheless, it was proven in the same document [4] that the jitter of a higher-priority task  $\tau_i$  can be safely upper bounded by  $R_i - C_i$ . It results that a task  $\tau_k$  with a constrained deadline can be feasibly scheduled under fixed-priority if

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

#### 3.2 MODELING THE SUSPENSION AS BLOCKING TIME

In [25, p. 164-165], Liu proposed a solution to study the schedulability of a self-suspending task  $\tau_k$  by modeling the extra delay suffered by  $\tau_k$  due to the self-suspension behavior of each task in  $\tau$  as a blocking time.<sup>1</sup> This blocking time has been defined as follows:

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

An upper bound on the blocking time is therefore given by:  $B_k = S_k + \sum_{i=1}^{k-1} \min(C_i, S_i)$ . In [25], 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  $\frac{C_k + B_k}{T_k} + \sum_{i=1}^{k-1} U_i \leqslant k(2^{\frac{1}{k}} - 1)$ , then  $\tau_k$  can be feasibly scheduled with rate-monotonic scheduling.

The same concept was also implicitly used by Rajkumar, Sha, and Lehoczky in [29, p. 267] for analyzing the impact of the self-suspension of a task due to the utilization of synchronization protocols in multiprocessor systems. (See Appendix for details.) If the above argument is correct, we can further prove that a constrained-deadline task  $\tau_k$  can be feasibly scheduled under fixed-priority scheduling if

$$\exists t \mid o < t \leqslant D_k, \quad C_k + B_k + \sum_{i=1}^{k-1} \left\lceil \frac{t}{T_i} \right\rceil C_i \leqslant t.$$
 (3)

However, there is no proof in [25] nor in [29] to support the correctness of those tests. Therefore, in Section 6, we provide a proof (see Theorem 4) of the correctness of Equation (3).

<sup>1</sup> Even though the authors in this paper are able to provide a proof to support the correctness, the authors are not able to provide any rationale behind this method which treats suspension time as blocking time.

Even though it can be proven that the response time analysis associated with Eq.(3) dominates the suspension oblivious one (see Lemma 15 in Section 6), none of the analyses presented in Section 3 dominates all the others. Hence, Eqs. (2) and (3) are incomparable. That is, in some cases Eq. (3) performs better than Eq. (2), while in others Eq. (2) outperforms Eq. (3).

**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  $\tau_1$  is obviously 9 whatever the analysis employed. However, the upper bound on  $R_2$  obtained with Eq. (2) is 15, while it is 19 with Eq. (3). The solution obtained with Eq. (2) is tighter.

Now, let us consider one more task  $\tau_3=(4,0,50,50)$ . Using Eq. (2), the worst-case response time  $R_3$  of task  $\tau_3$  is upper bounded by the smallest t>0 such that  $t=4+\left\lceil\frac{t+9-4}{10}\right\rceil 4+\left\lceil\frac{t+15-6}{19}\right\rceil 6$ , which turns out to be 42. With Eq. (3) though,  $B_3=4+1=5$  and an upper bound on  $R_3$  is given by the smallest t>0 such that  $C_3+B_3+\sum_{i=1}^2\left\lceil\frac{t}{T_i}\right\rceil C_i\leqslant t$ . The solution to this last equation is t=37. Therefore, Eq. (3) provides a tighter bound on  $R_3$  than Eq. (2), while the opposite was true for  $\tau_2$ .

In addition to the fact that Eqs. (2) and (3) are incomparable, there are task sets for which both equations overestimate the worst-case response time, e.g., in the following example.

**Example 2** Consider the same three tasks as in Example 1. As explained in Section 3.1, the extra interference caused by the self-suspending behavior of  $\tau_1$  can be safely modeled by a release jitter equal to  $R_1 - C_1 = 5$ . Similarly, the extra interference caused by the self-suspension of  $\tau_2$  can be modeled by a blocking time equal to  $\min(C_2, S_2) = 1$  (see Section 3.2). Hence, the worst-case response time  $R_3$  of  $\tau_3$  is upper bounded by the smallest t > 0 such that  $t = 4 + 1 + \left\lceil \frac{t+5}{10} \right\rceil 4 + \left\lceil \frac{t}{19} \right\rceil 6$ , which turns out to be 33. This bound on  $R_3$  is smaller than the estimates obtained with both Eqs. (2) and (3) (see Example 1).

Example 2 shows that a tighter bound on the worst-case response time of a task can be obtained by combining the properties of the analyses discussed in both Section 3.1 and 3.2. Therefore, in this paper, we derive a response time analysis that draws inspiration from both Eqs. (2) and (3), combining the best of each of them. As further proven in Section 6, the resulting schedulability test dominates all the tests discussed in Section 3.

5

In all this section, we implicitly assume that  $R_i \leqslant D_i, \forall \tau_i \mid 1 \leqslant i \leqslant k-1$ . This assumption is implicitly used as a fact in all the theorems and lemmas. Therefore, the worst-case response time or the schedulability of task  $\tau_k$  has to be verified from  $k=1,2,\ldots,n$ . Here we only focus on the analysis of a certain task  $\tau_k$ , under the assumption that we have already validated that  $R_i \leqslant D_i \leqslant T_i, \forall \tau_i \mid 1 \leqslant i \leqslant k-1$  (by using any method in this section or Section 3). Our key result in this paper are the two following theorems:

**Theorem 1** Suppose that all tasks  $\tau_{\ell} \mid 1 \leqslant \ell \leqslant k$  are schedulable, (i.e.,  $R_{\ell} \leqslant T_{\ell}$ ). Then, for any arbitrary vector assignment  $\vec{x} = (x_1, x_2, \dots, x_{k-1})$ , in which  $x_i$  is either 0 or 1, the worst-case response time  $R_k$  of  $\tau_k$  is upper bounded by the minimum t larger than 0 such that

$$C_{k} + S_{k} + \sum_{i=1}^{k-1} \left[ \frac{t + Q_{i}^{\vec{x}} + (1 - x_{i})(R_{i} - C_{i})}{T_{i}} \right] C_{i} \leqslant t$$
(4)

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

**Theorem 2** Suppose that  $\tau_k$  is not schedulable (i.e.,  $R_k > T_k$ ). For any arbitrary vector assignment  $\vec{x} = (x_1, x_2, \dots, x_{k-1})$ , in which  $x_i$  is either 0 or 1, we have  $\forall t | 0 < t \leqslant T_k$ ,

$$C_k + S_k + \sum_{i=1}^{k-1} \left[ \frac{t + Q_i^{\vec{x}} + (1 - x_i)(R_i - C_i)}{T_i} \right] C_i > t$$

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

By Theorems 1 and 2, we can directly derive the following schedulability test.

**Corollary 1** If  $\forall \tau_i \mid 1 \le i < k$ ,  $R_i \le T_i$ , and if there is a vector  $\vec{x} = (x_1, x_2, \dots, x_{k-1})$  with  $x_i \in \{0, 1\}$ , such that

$$\exists t | o < t \leq D_{\nu}$$

$$C_{k} + S_{k} + \sum_{i=1}^{k-1} \left[ \frac{t + Q_{i}^{\vec{x}} + (1 - x_{i})(R_{i} - C_{i})}{T_{i}} \right] C_{i} \leqslant t$$
 (5)

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

**Proof.** Let  $t^*$  be the first positive value of t respecting Eq. (5). By the assumptions stated in the claim,  $t^*$  exists and  $t^*$  is smaller than or equal to the deadline  $D_k$ . By Theorems 1 and 2,  $t^*$  exists and is smaller than or equal to  $D_k$  only if  $\tau_k$  is schedulable.  $\square$ 

The proof of correctness of Theorems 1 and 2, and hence Corollary 1 is provided in Section 5.1. Moreover, we will later prove in Section 6, that Corollary 1 in fact dominates all the analyses discussed in Section 3.

We now use the same example as in Section 4, to demonstrate how Corollary 1 can be applied.

$\vec{x}$	Case 1: (0,0)	Case 2: (0,1)	Case 3: (1,0)	Case 4:(1,1)
$(Q_1^{\vec{x}},Q_2^{\vec{x}})$	(0,0)	(1,1)	(5, o)	(6,1)
condition of Eq. (5)	$4 + \left\lceil \frac{t+o+5}{10} \right\rceil 4 + \left\lceil \frac{t+o+9}{19} \right\rceil 6 \leqslant t$	$4 + \left\lceil \frac{t+1+5}{10} \right\rceil 4 + \left\lceil \frac{t+1+0}{19} \right\rceil 6 \leqslant t$	$4 + \left\lceil \frac{t+5+o}{10} \right\rceil 4 + \left\lceil \frac{t+o+9}{19} \right\rceil 6 \leqslant t$	$4 + \left\lceil \frac{t+6+o}{10} \right\rceil 4 + \left\lceil \frac{t+1+o}{19} \right\rceil 6 \leqslant t$
upper bound of R <sub>3</sub>	42	32	42	32

Table 1: Detailed procedure in Example 3 for deriving the upper bound of  $R_3$ , with  $R_1 - C_1 = 5$  and  $R_2 - C_2 = 9$ .

**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  $\tau_3 = (4, 0, 50, 50)$ . By the analysis in Example 1,  $R_1$  is upper bounded by 9 and  $R_2$  is upper bounded by 15. Let us assume  $R_1 = 9$  and  $R_2 = 15$  in the rest of this example. There are four possible vector assignments  $\vec{x}$  when considering the schedulability of task  $\tau_3$  with Corollary 1. The corresponding procedure to use these four vector assignments can be found in Table 1. Among the above four cases, the tests in Cases 2 and 4 are the tightest.

Note also that the upper bound on R<sub>3</sub> computed in Example 3, is lower than the estimated worst-case response time obtained in Example 2. The response time analysis presented in Corollary 1 is therefore tighter than the simple combination of existing analysis techniques as proposed in Example 2.

#### 5.1 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:  $T_0$  and  $T_1$ . A task  $\tau_i$  is in  $T_0$  if  $x_i$  is 0; otherwise, it is in  $T_1$ . 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  $\tau_i$  is no more than its period  $T_i$ , removing a job of task  $\tau_i$  does not affect the schedule of any other jobs of task  $\tau_i$ .

# **Proof.** The proof is in Appendix. $\Box$

We now present the detailed proof of Theorems 1 and 2 using the properties stated above. Since the proof is quite long, we will also provide examples to demonstrate the key steps in the proof and lemmas to support intermediate results.

Let  $\Psi$  be a fixed-priority preemptive schedule of the task system  $\tau$ . Suppose that a job  $J_k$  of task  $\tau_k$  arrives at time  $r_k$  and finishes at time  $f_k$ . By the assumption of  $R_k \leqslant T_k$ , we have  $f_k \leqslant r_k + R_k \leqslant r_k + T_k$ . We first prove that Eq. (4) gives us a safe upper bound on  $f_k - r_k$  for any job  $J_k$  in  $\Psi$  if  $R_k \leqslant T_k$ . The proof is built upon the three following steps:

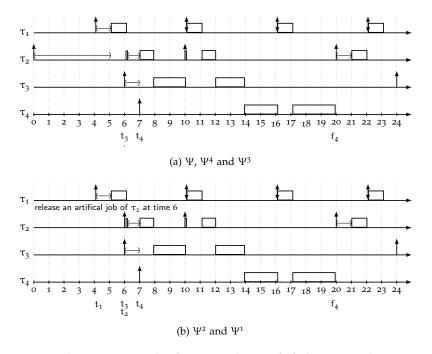


Figure 1: An illustrative example of Step 1 in the proof of Theorem 1 when  $\varepsilon=0.1$ .

- 1. We discard all the jobs that arrive before  $r_k$  and do not contribute to the response time of  $J_k$  in the schedule  $\Psi$ . We follow an inductive strategy by iteratively inspecting the schedule of the higher priority tasks in  $\Psi$ , starting with  $\tau_{k-1}$  until the highest priority task  $\tau_1$ . At each iteration, a time instant  $t_j$  is identified such that  $t_j \leqslant t_{j+1}$  ( $1 \leqslant j < k$ ). Then, all the jobs of task  $\tau_j$  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  $\tau_j$  at time  $t_j$ .
- 2. The final reduced schedule is analyzed to characterize important properties of the reduced schedule in Step 1.
- 3. We then prove that the response time analysis in Eq. (4) is indeed an upper bound on the worst-case response time  $R_k$  of  $\tau_k$ .

# Step 1: Reducing the schedule $\Psi$

Our purpose in this step is to discard all the jobs that arrive before  $\tau_k$  and have no impact on the response time of  $J_k$  in the schedule  $\Psi.$  During this step, we iteratively build the schedules from  $\Psi^k$  to  $\Psi^I$  mentioned above. Based on a given schedule  $\Psi^{j+1}$  (with  $1\leqslant j< k$ ), we build the fixed-priority schedule  $\Psi^j$  such that the response time of  $J_k$  remains identical. At each iteration, we define  $t_j$  for task  $\tau_j$  in the schedule  $\Psi^{j+1}$  and build  $\Psi^j$  by removing all the jobs released by  $\tau_j$  before  $t_j.$  We then prove that the response time of  $J_k$  in the reduced fixed-priority schedule  $\Psi^I$  remains the same as the response time of  $J_k$  in the original fixed-priority schedule  $\Psi.$  Basic step (definition of  $\Psi^k$  and  $t_k$ ):

We define  $\Psi^k$  as the schedule in which (i) all high-priority tasks  $\tau_1, \ldots, \tau_{k-1}$  release their jobs at the exact same instants as in  $\Psi$ , (ii)  $\tau_k$  releases only one job at time  $r_k$ , (iii) the low-priority tasks  $\tau_{k+1}, \ldots, \tau_n$  do not release any job, and

(iv) all jobs suspend their execution after the exact same execution time as in  $\Psi$ . Moreover, since  $J_k$  is released at time  $r_k$  and does not finish strictly before  $f_k$ , the total amount of idle time of the system from  $r_k$  to  $f_k$  is at most  $S_k$ . In the converted schedule  $\Psi^k$ , we further convert the idle time as part of the execution time of  $J_k$ . After the conversion by considering suspension as computation for job  $J_k$ , we know that the worst-case execution time of  $J_k$  is upper bounded by  $C_k' = S_k + C_k$ . As already discussed in Section 3, such a conversion has been widely used. For notational brevity, we denote this job  $J_k$  as a release of task  $\tau_k' = (C_k + S_k, o, D_k, T_k)$ . It is obvious that  $\Psi^k$  remains as a fixed-priority schedule.

**Lemma 2** The response time of  $J_k$  in  $\Psi^k$  is the same as the response time of  $J_k$  in  $\Psi$ .

**Proof.** 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  $J_k$ . Therefore, not releasing them has no impact on the response time  $J_k$ . Moreover, since we assume that the worst-case response time of task  $\tau_k$  is no more than  $T_k$ , Lemma 1 proves that none of the jobs of task  $\tau_k$  except  $J_k$  impacts the schedule of  $J_k$ . Since all the other parameters (i.e., releases and suspensions) that may influence the scheduling decisions are kept identical between  $\Psi$  and  $\Psi^k$ , the response time of  $J_k$  in  $\Psi^k$  is identical to the response time of  $J_k$  in  $\Psi$ .  $\square$ 

To allow the induction defined below, we also define  $t_k$  as the release time of  $J_k$  (i.e.,  $t_k \stackrel{\text{def}}{=} r_k$ ).

Induction step (definition of  $\Psi^j$  and  $t_i$  with  $1 \le j < k$ ):

We define four cases in order to build  $\Psi^{j}$  from  $\Psi^{j+1}$ .

Case o. If all the jobs of task  $\tau_j$  are released at or after  $t_{j+1}$  in schedule  $\Psi^{j+1}$ , then we define  $\Psi^j$  as being identical to  $\Psi^{j+1}$  and set  $t_j \stackrel{\text{def}}{=} t_{j+1}$ .

Now, let us consider that task  $\tau_j$  releases at least one job before  $t_{j+1}$  in  $\Psi^{j+1}$ . Let  $r_j$  be the arrival time of the last job released by  $\tau_j$  before  $t_{j+1}$  in  $\Psi^{j+1}$  and let  $J_j$  denote that job. By definition,  $r_j < t_{j+1}$ . Let  $c_j^*$  be the remaining execution time of  $J_j$  at time  $t_{j+1}$  in  $\Psi^{j+1}$ . By definition,  $o \leqslant c_j^* \leqslant C_j$ . In the rest of the proof,  $c_i^*$  is called  $\tau_j$ 's residual workload.

We start by setting  $\Psi^j$  to be identical as  $\Psi^{j+1}$ . Then, all the jobs of task  $\tau_j$  that arrive before  $r_j$  are immediately removed from  $\Psi^j$ . That is, all jobs released in  $\Psi^j$  have identical suspension and execution behavior as in  $\Psi^{j+1}$ , and task  $\tau_j$  does not release any job before  $r_j$  in  $\Psi^j$ . There are three cases to decide how we include or exclude  $J_i$  in  $\Psi^j$  as follows:

**Case 1.** If  $\tau_j \in T_1$  and  $J_j$  does not complete its execution by  $t_{j+1}$  in the schedule  $\Psi^{j+1}$  (i.e.,  $c_j^* > o$ ), then  $t_j \stackrel{def}{=} r_j$  and  $J_j$  is included in  $\Psi^j$ . In this case, task  $\tau_j$  releases its jobs at exactly the same instants in  $\Psi^{j+1}$ , i.e., at and after  $r_j$ .

Case 2. If  $\tau_j \in T_1$  and  $J_j$  completes its execution before or at  $t_{j+1}$  in the schedule  $\Psi^{j+1}$  (i.e.,  $c_j^* = o$ ), then  $t_j \stackrel{def}{=} t_{j+1}$  and  $J_j$  is further removed and excluded from  $\Psi^j$ . In this case, task  $\tau_j$  releases its jobs at exactly the same instants in  $\Psi^{j+1}$  after  $r_j$ .

Case 3. If  $\tau_j \in T_o$ , then  $t_j \stackrel{def}{=} t_{j+1}$  and (i)  $\tau_j$  releases its jobs at the same instants in  $\Psi^{j+1}$  after  $r_j$  (i.e., exclude  $J_j$  in  $\Psi^j$ ), and (ii) an artificial (or additional) job  $J_\alpha$  with execution time  $C_\alpha \stackrel{def}{=} c_i^*$  and the same priority as  $\tau_j$  is released at time

 $t_{j+1}$ . This artificial job follows the same execution and suspension behavior as job  $J_j$  after  $t_{j+1}$ .

After the above procedures, it is obvious that the resulting schedule  $\Psi^j$  remains as a fixed-priority schedule.

**Lemma 3** The response time of  $J_k$  in  $\Psi^j$  is the same as the response time of  $J_k$  in  $\Psi^{j+1}$ .

**Proof.** If Case o is applied, then  $\Psi^j$  and  $\Psi^{j+1}$  are identical. The response time of  $J_k$  is thus unchanged and the claim trivially holds.

For the rest of the proof, we use the four following facts:

**Fact 1.** For any  $\ell$  such that  $j \leqslant \ell < k$ , there is  $t_{\ell} \leqslant t_{\ell+1}$  and  $\tau_{\ell}$  does not release any job before  $t_{\ell}$  in  $\Psi^{j}$ .

**Fact 2.** No job of  $\tau_k, \ldots, \tau_n$  are released before  $t_k$  in  $\Psi^k$ .

Fact 3. By the assumption that  $R_j \leqslant D_j \leqslant T_j$  for  $j=1,2,\ldots,k-1$ , removing all the jobs of task  $\tau_j$  arrived before  $r_j$  has no impact on the schedule of any other job released by  $\tau_j$  (Lemma 1) or any higher priority job released by  $\tau_1,\ldots,\tau_{j-1}$  (Property 1). Moreover, by Facts 1 and 2, no task with a priority lower than  $\tau_j$  (tasks  $\tau_{j+1},\ldots,\tau_n$ ) release jobs before  $t_{j+1}$  in  $\Psi^{j+1}$ . Therefore, removing the jobs released and completed by  $\tau_j$  before  $r_j$  does not impact the schedule of the jobs of  $\tau_{j+1},\ldots,\tau_n$ . Consequently, we can safely remove all the jobs of task  $\tau_j$  arrived and completed before  $t_{j+1}$  without impacting the response time of  $J_k$ .

**Fact 4.** Since by Facts 1 and 2, no task with a priority lower than  $\tau_j$  (tasks  $\tau_{j+1}, \ldots, \tau_n$ ) releases jobs before  $t_{j+1}$  in  $\Psi^{j+1}$ , replacing  $J_j$  with the created artificial job (which has the same execution and suspension behavior as  $J_j$  from  $t_{j+1}$ ) has no impact on the schedule of  $\tau_{j+1}, \ldots, \tau_n$  in  $\Psi^{j+1}$ .

We now consider the three remaining cases:

In Case 1,  $\Psi^j$  is built from  $\Psi^{j+1}$  by removing all the jobs released by  $\tau_j$  before  $r_j$ . All the excluded jobs have therefore completed their execution at  $t_{j+1}$  and by Fact 3, this has no impact on the execution of any job executed after  $t_{j+1}$  and thus on the response time of  $J_k$ .

In Case 2,  $\Psi^j$  is built from  $\Psi^{j+1}$  by removing all the jobs released by  $\tau_j$  before  $t_{j+1}$ . Since  $J_j$  completes before  $t_{j+1}$ , by Fact 3, none of the excluded jobs impacted the response time of  $J_k$ . The response time of  $J_k$  in  $\Psi^j$  thus remains unchanged in comparison to its response time in  $\Psi^{j+1}$ .

In Case 3, all the jobs of  $\tau_j$  released before  $t_{j+1}$  are removed and the job of task  $\tau_j$  arrived at time  $r_j$  is replaced by a new job released at time  $t_{j+1}$  with execution time  $c_j^*$  and the same priority as  $\tau_j$ . By Facts 3 and 4, the interference generated by  $\tau_j$  and the additional job on job released at or after  $t_{j+1}$  remains identical between  $\Psi^j$  and  $\Psi^{j+1}$ . Thus, the response time of  $J_k$  is unchanged.  $\square$  Conclusion of Step 1:

This iterative process is repeated until producing  $\Psi^{\rm I}$ . The procedures are well-defined and it is therefore guaranteed that  $\Psi^{\rm I}$  can be constructed. A pseudocode of  $\Psi^{\rm I}$ 's generation procedure can be found in Appendix. Note that after each iteration, the number of jobs considered in the resulting schedule has been reduced, yet without affecting the response time of  $J_k$ , as proven in the following lemma.

**Lemma 4** The response time of job  $J_k$  in  $\Psi^1$  is the same as the response time of  $J_k$  in  $\Psi$ .

**Proof.** By Lemma 2, the reponse time of  $\tau_k$  in  $\Psi^k$  is identical to the response time of  $J_k$  in  $\Psi$ . And by inductively applying Lemma 3, we get that the response time of  $\tau_k$  in  $\Psi^k$  is identical to the response time of  $\tau_k$  in  $\Psi^{k-1}$ ,  $\Psi^{k-2}$ , ...,  $\Psi^1$ . This proves the lemma.  $\square$ 

**Example 4** Consider 4 tasks  $\tau_1=(1,1,6,6)$ ,  $\tau_2=(1,6,10,10)$ ,  $\tau_3=(4,1,18,18)$  and  $\tau_4=(5,0,20,20)$ . We assume  $x_1=1$ ,  $x_2=0$  and  $x_3=1$ . Figure 1(a) depicts a possible schedule  $\Psi^4$  of those tasks. We assume that the first job of task  $\tau_1$  arrives at time  $4+\varepsilon$  with  $0<\varepsilon<0.5$ . The first job of task  $\tau_2$  suspends itself from time 0 to time  $5+\varepsilon$ , and is blocked by task  $\tau_1$  from time  $5+\varepsilon$  to time  $6+\varepsilon$ . After executing  $\varepsilon$  amount of time, the first job of task  $\tau_2$  suspends itself again from time  $6+2\varepsilon$  to 7. The schedule in Figure 1(a) is drawn for  $\varepsilon=0.1$ .

In the schedule illustrated in Figure 1(a),  $f_4$  is  $20-\varepsilon$ . Based on the definition of  $t_k$ ,  $t_4=7$ . Then, we set  $t_3$  to 6 by applying Case 1. The schedule  $\Psi^3$  is identical to the original schedule  $\Psi^4$ . When considering task  $\tau_2$ , we know that  $J_2$  is the job of task  $\tau_2$  arrived at time  $r_2=0< t_3$ . Since task  $\tau_2$  belongs to  $T_0$ , by applying Case 3, we have  $t_2=t_3=6$  and the residual workload  $c_2^*$  is 1. Then, we remove job  $J_2$  from the schedule and create an artificial job with execution time  $c_2^*$  that is released at time  $t_2$  and assign the artificial job the same priority level as task  $\tau_2$ . Note that this artificial job can still suspend itself. Therefore, the schedule  $\Psi^2$ , as drawn in Figure 1(b), is slightly different from  $\Psi^3$ , shown in Figure 1(a). Yet, the response time of  $J_4$  is unchanged. Finally,  $t_1$  is set to  $4+\varepsilon$  by applying Case 1 since  $J_1$  (arrived at time  $r_1=4+\varepsilon$ ) has not completed yet at time  $t_2=6$ . The schedule  $\Psi^1$  is identical to the schedule  $\Psi^2$ .  $\square$ 

### Step 2: Analyzing the reduced schedule $\Psi^1$

We now analyze the properties of the final fixed-priority schedule  $\Psi^{\text{I}}$  in which all the unnecessary jobs have been removed. This step is based on the simple fact that for any interval  $[t_1, t)$  with  $t \leq f_k$ , there is

$$idle(t_1, t) + exec(t_1, t) = (t - t_1)$$
 (6)

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

We first provide an upper bound on idle( $t_1$ , t) (see Lemma 5 and Corollary 2), then on exec( $t_1$ , t) (see Lemmas 6 to 9). Finally, in Lemma 10, we combine those results with Eq. (6) in order to characterise the schedule  $\Psi^1$  in  $[t_1, f_k)$ .

We start our analysis with  $idle(t_1, t)$  when  $t_1 < t \le f_k$ . Let  $\sigma_j$  be the amount of time during which the processor remains idle within  $[t_j, t_{j+1}]$  in  $\Psi^1$ .

**Lemma 5** For 
$$j = 1, 2, ..., k-1$$
,  $\sigma_j = x_j \times \sigma_j \leqslant x_j \times S_j$ .

**Proof.** If Case 1 is applied on  $\tau_j$  when we build  $\Psi^j$  in Step 1, (i)  $x_j = 1$ , (ii)  $t_j$  is set to the release time  $r_j$  of the job  $J_j$ , and (iii)  $J_j$  has not completed its execution yet at time  $t_{j+1}$ . By (ii) and (iii), the amount of time during which the processor may remain idle within  $[t_j, t_{j+1})$  is at most the suspension time  $S_j$  of  $\tau_j$ . Thus,  $\sigma_j \leq S_j$ . And by (i),  $\sigma_j = x_j \times \sigma_j \leq x_j \times S_j$ .

If Cases 0, 2 or 3 is applied on  $\tau_j$  when we build  $\Psi^j$  in Step 1, then  $t_j$  is equal to  $t_{j+1}$  and by definition,  $\sigma_j = 0$ . It results that  $\sigma_j = x_j \times \sigma_j \leqslant x_j \times S_j$ .  $\square$ 

**Corollary 2** For  $i = 1, 2, ..., k-1, \forall t | t_i < t \le t_{i+1}$ 

$$idle(t_1, t) \leqslant \sum_{j=1}^{i} x_j \sigma_j \leqslant \sum_{j=1}^{i} x_j S_j$$
 (7)

**Proof.** Since  $t_i < t \leqslant t_{i+1}$ ,  $idle(t_i,t) \leqslant \sum_{j=1}^i \sigma_j$ . And by Lemma 5,  $idle(t_i,t) \leqslant \sum_{j=1}^i x_j \sigma_j \leqslant \sum_{j=1}^i x_j S_j$ 

**Example 5** As shown in the schedule in Example 4, the total idle time from  $4 + \varepsilon$  to  $20 - \varepsilon$ , i.e., from  $4 + \varepsilon$  to  $5 + \varepsilon$  and from  $6 + 2\varepsilon$  to 7, is  $2 - 2\varepsilon$ , which is upper-bounded by  $S_1 + S_3 = 2$ .

We now consider  $\text{exec}(t_1,t)$  when  $t_1 < t \leqslant f_k$ . Because there is no job released by lower priority tasks than  $\tau'_k$  in  $\Psi^{\text{I}}$ , we only focus on the execution of the tasks  $(\tau_1,\tau_2,\ldots,\tau_{k-1},\tau'_k)$ . Let  $\text{exec}_j(t_1,t)$  be the (accumulative) amount of time that task  $\tau_j$  is executed in the schedule  $\Psi^{\text{I}}$  in the time interval  $(t_1,t]$ . By the construction of the schedule  $\Psi^{\text{I}}$ , we know that  $\text{exec}_j(t_1,t_j)$  must be equal to o since task  $\tau_j$  is not executed between  $t_1$  and  $t_j$ . Therefore,  $\text{exec}_j(t_1,t)$  is equal to  $\text{exec}_j(t_j,t)$  if  $t>t_j$ .

**Lemma 6**  $\forall t | t_k \leq t < f_k$ , the (accumulative) amount of time that task  $\tau_k'$  is executed from  $t_k$  to t is  $exec_k(t_k, t) < C_k'$ .

**Proof.** Since the finishing time of job  $J_k$  is at time  $f_k$  in schedule  $\Psi^1$ , the condition holds by definition.  $\square$ 

**Lemma 7** *If task*  $\tau_i \in T_1$ , then  $\forall \Delta \geq 0$  we have

$$\operatorname{exec}_{\mathbf{j}}(\mathbf{t}_{\mathbf{j}}, \mathbf{t}_{\mathbf{j}} + \Delta) \leqslant W_{\mathbf{j}}^{1}(\Delta)$$

where

$$W_{j}^{1}(\Delta) \stackrel{def}{=} \left| \frac{\Delta}{\mathsf{T}_{i}} \right| \mathsf{C}_{j} + \min \left\{ \Delta - \left| \frac{\Delta}{\mathsf{T}_{i}} \right| \mathsf{T}_{j}, \mathsf{C}_{j} \right\}. \tag{8}$$

**Proof.** If task  $\tau_j \in T_1$ , then Case 0, 1 or 2 is applied when building  $\Psi^1$  in Step 1. In this case,  $\Psi^1$  does not contain any job of task  $\tau_j$  arrived before  $t_j$  (i.e., no residual workload of  $\tau_j$  at time  $t_j$ ). Furthermore,  $\text{exec}_j(t_j, t_j + \Delta)$  is maximized when the jobs released by  $\tau_j$  after  $t_j$  are actually executing, and hence do not suspend themselves (i.e.,  $\tau_j$  acts as a sporadic tasks without self-suspension). Since, as shown in the literature [3],  $W_j^1(\Delta)$ , which is usually called *workload function*, is an upper bound on the amount of execution time that a sporadic task can execute without self-suspension, we know that  $\text{exec}_j(t_j, t_j + \Delta)$  of  $\tau_j$  from  $t_j$  to  $t_j + \Delta$  is upper bounded by  $W_j^1(\Delta)$ .  $\square$ 

**Lemma 8** *If*  $\tau_i \in T_o$ , then  $\forall \Delta \geqslant o$  we have

$$\operatorname{exec}_{j}(\mathsf{t}_{j},\mathsf{t}_{j}+\Delta)\leqslant\widehat{W}_{j}^{o}(\Delta,c_{j}^{*})$$

where

$$\widehat{W}_{j}^{o}(\Delta, \mathbf{c}_{j}^{*}) = \begin{cases} W_{j}^{1}(\Delta) & \text{if } \mathbf{c}_{j}^{*} = \mathbf{0} \\ \Delta & \text{if } \mathbf{c}_{j}^{*} > \mathbf{0} \text{ and } \Delta \leqslant \mathbf{c}_{j}^{*} \\ \mathbf{c}_{j}^{*} & \text{if } \mathbf{c}_{j}^{*} > \mathbf{0} \text{ and } \mathbf{c}_{j}^{*} < \Delta \leqslant \rho_{j} \\ \mathbf{c}_{j}^{*} + W_{j}^{1}(\Delta - \rho_{j}) & \text{if } \mathbf{c}_{j}^{*} > \mathbf{0} \text{ and } \rho_{j} < \Delta \end{cases}$$

$$(9)$$

and  $\rho_{j} = (T_{j} - R_{j} + c_{i}^{*}).$ 

**Proof.** If task  $\tau_j \in T_o$ , then Case o or 3 is applied when building  $\Psi^1$  in Step 1. Therefore, there might be a job  $J_j$  arrived before  $t_j$  with a residual workload  $o \leq c_j^* \leq C_j$  at time  $t_j$ . The case when  $c_j^* = o$  is identical to the proof of Lemma 7. We now consider the cases where  $c_j^* > o$ . Since by assumption  $R_j \leq D_j \leq T_j$ , task  $\tau_j$  respects all its deadlines and the worst-case response time, the absolute deadline of the job  $J_j$  of  $\tau_j$  that is not completed yet at  $t_j$ , must be at least  $t_j + c_j^*$ . Therefore, the earliest arrival time of a job of task  $\tau_j$  strictly after  $t_j$  is at least  $t_j + c_j^* + (T_j - R_j) = t_j + \rho_j$ . Since there is no other job of task  $\tau_j$  released in  $[t_j, \rho_j)$  except the artificial job with the residual workload  $c_j^*$  created based on  $J_j$ , we know that  $exec_j(t_j, t_j + \Delta)$  is upper bounded by min  $\{\Delta, c_j^*\}$  for  $\Delta \leq \rho_j$ , thereby proving cases 2 and 3 of Eq. (9). Furthermore, by assumption  $J_j$  completes its execution before or at  $t_j + \rho_j$ . Therefore, following the same proof as Lemma 7,  $exec_j(t_j + \rho_j, t_j + \Delta)$  is upper bounded by  $W_j^1(\Delta - \rho_j)$  when  $\Delta > \rho_j$ . This proves the fourth case of Eq. (9).  $\Box$ 

For notational brevity, let  $W_j^o(\Delta) \stackrel{\text{def}}{=} \widehat{W_j^o}(\Delta, C_j)$ . We also prove that, for any  $\Delta \geqslant o$ ,  $W_j^o(\Delta) \geqslant \widehat{W_j^o}(\Delta, c_j^*)$ :

**Lemma 9**  $\forall \Delta \geqslant 0$ ,  $W_{\mathbf{j}}^{0}(\Delta) \geqslant \widehat{W}_{\mathbf{j}}^{0}(\Delta, c_{\mathbf{j}}^{*})$ .

**Proof.** The proof is based on simple observations of the workload function. The proof is in Appendix.  $\Box$ 

Now that we have derived upper bounds on the idle time  $idle(t_1,t)$  and the execution time  $exec_j(t_j,t_j+\Delta)$  of each task  $\tau_j$  executed in  $\Psi^i$ , we inject those results in Eq. (6) in order to derive properties on the schedule in any interval  $[t_1,t)$  for any  $t_1 < t < f_k$ .

**Lemma 10**  $\forall t \mid t_i \leqslant t < t_{i+1} \text{ where } i = 1, 2, \dots, k-1$ 

$$\sum_{j=1}^{i} \left( x_j \cdot (W_j^{I}(t-t_j) + \sigma_j) + (I - x_j) \cdot W_j^{O}(t-t_j) \right) \geqslant t - t_1. \tag{10}$$

And,  $\forall t \mid t_k \leq t < f_k$ ,

$$C_k' + \sum_{j=1}^{k-1} \left( x_j \cdot (W_j^{\scriptscriptstyle \rm I}(t-t_j) + \sigma_j) + ({\bf 1} - x_j) \cdot W_j^{\scriptscriptstyle \rm O}(t-t_j) \right) > t - t_{\scriptscriptstyle \rm I}. \tag{11}$$

**Proof.** We combine the three following facts:

- **1.** By Eq. (6),  $idle(t_1, t) + exec(t_1, t) = t t_1$ .
- **2.** By Corollary 2, idle( $t_1$ , t)  $\leqslant \sum_{j=1}^{i-1} x_j \sigma_j$  for all  $t \mid t_i \leqslant t < t_{i+1}$  and  $i = 1, 2, \ldots, k-1$ .
- 3. By the construction of the schedule  $\Psi^1$ , we know that  $exec_j(t_1,t_j)=o$  since task  $\tau_j$  is not executed between  $t_1$  and  $t_j$ . Therefore,  $exec_j(t_1,t)=o$  if  $t< t_j$  and

<sup>1</sup> The readers may think of using the condition  $idle(t_1,t)\leqslant \sum_{j=1}^{i-1}x_jS_j$  in Eq. (7) to replace  $\sigma_j$  with  $S_j$ . But, this will create a serious problem in Step 3 later, since we cannot always guarantee that  $t_i^*\leqslant t_i$  for  $i=1,2,\ldots,k$  in Step 3 if we do so in Step 2. Such a treatment should not be applied at this moment here.

 $exec_j(t_1,t) = exec_j(t_j,t)$  if  $t > t_j$ . Since  $x_j = o$  if  $\tau_j \in T_o$  and  $x_j = 1$  if  $\tau_j \in T_1$ , by Lemmas 7, 8 and 9, we have for all  $t \mid t_i \leqslant t < t_{i+1}$  and  $i = 1,2,\ldots,k-1$ ,

$$\begin{aligned} & \operatorname{exec}(t_{1}, t) = \sum_{j=1}^{i} \operatorname{exec}_{j}(t_{1}, t) = \sum_{j=1}^{i} \operatorname{exec}_{j}(t_{j}, t) \\ & \leq \sum_{j=1}^{i} \left( x_{j} \cdot W_{j}^{1}(t - t_{j}) + (1 - x_{j}) \cdot \widehat{W}_{j}^{0}(t - t_{j}, c_{j}^{*}) \right) \\ & \leq \sum_{j=1}^{i} \left( x_{j} \cdot W_{j}^{1}(t - t_{j}) + (1 - x_{j}) \cdot W_{j}^{0}(t - t_{j}) \right) \end{aligned}$$

$$(12)$$

Therefore, combining Corollary 2, Eq. (12) and Eq. (6), we obtain Eq. (10). Moreover, since  $\tau_k'$  does not complete its execution  $\mathit{strictly}$  before  $f_k$  and because, by definition,  $\tau_k'$  does not self-suspend, we also know that idle( $t_k,t$ ) = o for  $t_k \leqslant t < f_k$ . Therefore, using Corollary 2, we get for all  $t \mid t_k \leqslant t < f_k$ 

$$idle(t_1,t)\leqslant \sum_{j=1}^{k-1}x_j\sigma_j. \tag{13}$$

Furthermore, by Lemma 6,  $\text{exec}_k(t_k,t) < C_k'$  for  $t < f_k.$  Therefore, adding  $\text{exec}_k(t_k,t)$  to Eq. (12), we get for all  $t \mid t_k \leqslant t < f_k$ 

$$\operatorname{exec}(\mathsf{t}_{1},\mathsf{t}) < C_{k}' + \sum_{j=1}^{k-1} \left( x_{j} \cdot W_{j}^{1}(\mathsf{t} - \mathsf{t}_{j}) + (\mathsf{1} - x_{j}) \cdot W_{j}^{0}(\mathsf{t} - \mathsf{t}_{j}) \right). \tag{14}$$

Combining Eqs. (13), (14) and (6), we obtain Eq. (11).  $\Box$ 

**Example 6** Consider the same 4 tasks as in Example 4, for which a possible schedule was depicted in Figure 1 when  $\varepsilon$  is very close to 0. We have  $x_1\sigma_1=1$ ,  $x_2\sigma_2=0$  and  $x_3\sigma_3=1-2\varepsilon$ . 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 when  $\varepsilon$  is close to 0 and  $R_2=10$ . As can be seen in Figure 2, the inequalities of Eqs. (10) and (11) clearly hold.  $\square$ 

Before moving to Step 3, the following lemma is useful for setting the upper bounds of the workload functions.

**Lemma 11** For any  $\Delta > 0$ , we have

$$W_{j}^{1}(\Delta) \leqslant \left\lceil \frac{\Delta}{\mathsf{T}_{j}} \right\rceil \mathsf{C}_{j} \tag{15}$$

$$W_{j}^{o}(\Delta) \leqslant \left\lceil \frac{\Delta + R_{j} - C_{j}}{T_{j}} \right\rceil C_{j} \tag{16}$$

**Proof.** The upper bound of  $W_j^1(\Delta)$  is trivial. Therefore, we focus on the upper bound of  $W_j^0(\Delta)$ .

If o  $< \Delta \le C_j$ , then by Eq. (9),  $W_j^0(\Delta) = \Delta \le C_j \le \left\lceil \frac{\Delta + R_j - C_j}{T_j} \right\rceil C_j$ . If  $\Delta > C_j$ , then by the third and fourth case of Eq. (9)

$$\begin{split} W_j^o(\Delta) &\leqslant C_j + W_j^r(\Delta - (T_j - R_j + C_j)) \\ &\leqslant C_j + \left\lceil \frac{\Delta - T_j + (R_j - C_j)}{T_i} \right\rceil C_j = \left\lceil \frac{\Delta + R_j - C_j}{T_i} \right\rceil C_j. \end{split}$$

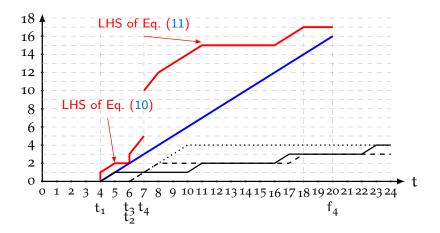


Figure 2: The workload function for the three higher-priority tasks in Example 4 when  $\varepsilon$  is very close to 0. Solid black line:  $W_1^t(t-t_1)$  when  $t\geqslant t_1$ , Dashed black line:  $W_2^0(t-t_2)$  when  $t\geqslant t_2$ , Dotted black line:  $W_3^t(t-t_3)$  when  $t\geqslant t_3$ , where the three workload functions are 0 if  $t-t_j<0$  for j=1,2,3, Blue line (the only linear function from t=4 in this figure):  $t-t_1$ , Red line (marked by Eq. (10) and Eq. (11)): left-hand side of Eq. (10) when t<7 and left-hand side of Eq. (11) when  $7\leqslant t<20$ .

#### Step 3: Creating a Safe Response-Time Upper Bound

The conditions in Lemma 10 cannot be used directly since the values  $t_j$  ( $j=1,2,\ldots,k$ ) are unknown in the general case. Therefore, Step 3 constructs a safe response-time analysis based on the conditions specified by Eqs. (10) and (11) in Lemma 10. Our goal in this step is to prove that Eq. (4) in Theorem 1 covers all the cases listed in Lemma 10 for any fixed-priority schedule  $\Psi^1$  generated from schedule  $\Psi$ .

Our proof strategy is to first artificially move  $t_i$  to  $t_i^*$  for  $i=1,2,\ldots,k$  such that  $t_i^* \leq t_i$ . We define  $t_i^*$  as follows:

$$t_{i}^{*} \stackrel{\text{def}}{=} \begin{cases} t_{1} & \text{if } i = 1\\ t_{i-1}^{*} + x_{i-1} \times \sigma_{i-1} & \text{if } i = 2, 3, \dots, k \end{cases}$$
 (17)

and we prove that t<sub>i</sub>\* is indeed smaller than or equal to t<sub>i</sub>.

**Lemma 12**  $t_i^* \leq t_i$  *for* i = 1, 2, ..., k.

**Proof.** By the definition of  $\sigma_i$ , we know that  $\sigma_i \leqslant t_{i+1} - t_i$  for i = 1, 2, ..., k-1. Therefore, for i = 2, 3, ..., k,

$$t_{i} = t_{1} + \sum_{j=1}^{i-1} (t_{j+1} - t_{j}) \geqslant t_{1} + \sum_{j=1}^{i-1} \sigma_{j} \geqslant t_{1} + \sum_{j=1}^{i-1} x_{j} \sigma_{j} = t_{i}^{*}$$

since  $x_j \in \{0,1\}$  for any  $j=1,2,\ldots,i-1$ . Finally, the property trivially holds for i=1.  $\square$ 

Lemma 13  $\forall t \mid t_k^* \leqslant t < f_k$ ,

$$C'_{k} + \sum_{j=1}^{k-1} x_{j} \cdot W_{j}^{1}(t - t_{j}^{*}) + (1 - x_{j}) \cdot W_{j}^{0}(t - t_{j}^{*}) > t - t_{k}^{*}.$$
(18)

**Proof.** Because  $t_j \geqslant t_j^*$  by Lemma 12, we have  $\forall \Delta \geqslant o$ 

$$W_{\mathbf{j}}^{1}(\Delta) \leqslant W_{\mathbf{j}}^{1}(\Delta + (\mathbf{t}_{\mathbf{j}} - \mathbf{t}_{\mathbf{j}}^{*})) \tag{19}$$

$$W_j^{o}(\Delta) \leqslant W_j^{o}(\Delta + (t_j - t_j^*)). \tag{20}$$

It results that, for  $j=1,2,\ldots,k-1$ ,  $W_j^1(t-t_j)\leqslant W_j^1(t-t_j^*)$  and  $W_j^0(t-t_j)\leqslant W_j^0(t-t_j^*)$  for any  $t\geqslant t_j$ . Injecting those two inequalities into Eq. (10)  $\forall t\mid t_k^*\leqslant t< t_k$  leads to<sup>2</sup>

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

and because by Eq. (17),  $t_k^* \stackrel{def}{=} t_{\scriptscriptstyle 1} + \sum_{j=\scriptscriptstyle 1}^{k-1} x_j \sigma_j$  , we get

$$\sum_{i=1}^{k-1} x_j \cdot W_j^{I}(t - t_j^*) + (\mathbf{1} - x_j) \cdot W_j^{O}(t - t_j^*) \geqslant t - t_k^*, \tag{21}$$

since  $C'_k \ge C_k > 0$ , it finally holds that

$$C'_{k} + \sum_{j=1}^{k-1} x_{j} \cdot W_{j}^{1}(t - t_{j}^{*}) + (1 - x_{j}) \cdot W_{j}^{0}(t - t_{j}^{*}) > t - t_{k}^{*}.$$
(22)

Similarly, injecting Eqs. (19) and (20) into Eq. (11)  $\forall t \mid t_k \leqslant t < f_k$  leads to

$$C'_{k} + \sum_{j=1}^{k-1} x_{j} \cdot W_{j}^{1}(t - t_{j}^{*}) + (1 - x_{j}) \cdot W_{j}^{0}(t - t_{j}^{*}) > t - t_{k}^{*}.$$
(23)

By Eq. (22) (valid for  $\forall t \mid t_k^* \leqslant t < t_k$ ) and Eq. (23) (valid  $\forall t \mid t_k \leqslant t < f_k$ ), we prove the lemma.  $\Box$ 

**Lemma 14**  $\forall \theta \mid o \leqslant \theta < f_k - t_{k'}^*$ 

$$C'_{k} + \sum_{j=1}^{k-1} \left\lceil \frac{\theta + X_{j} + (1 - x_{j})(R_{j} - C_{j})}{T_{j}} \right\rceil C_{j} > \theta,$$
 (24)

where  $X_j$  is  $\sum_{\ell=j}^{k-1} x_{\ell} \sigma_{\ell}$ .

**Proof.** By Eq. (17), we have  $t_j^* = t_k^* - \sum_{\ell=j}^{k-1} x_\ell \sigma_\ell$ . Therefore,  $\forall t \mid t_k^* \leqslant t < f_k$ , we have  $t - t_j^* = t - t_k^* + \sum_{\ell=j}^{k-1} x_\ell \sigma_\ell = t - t_k^* + X_j$  for every  $j = 1, 2, \ldots, k-1$ . By using Lemma 11 and  $t - t_j^*$  above, we can rewrite the condition in Lemma 13 as  $C_k' + \sum_{j=1}^{k-1} \left(x_j \left\lceil \frac{t - t_k^* + X_j}{T_j} \right\rceil C_j + (1 - x_j) \left\lceil \frac{t - t_k^* + X_j + R_j - C_j}{T_j} \right\rceil C_j \right) > t - t_k^*$ ,  $\forall t \mid t_k^* \leqslant t < f_k$ . Since  $x_j$  is either 0 or 1, this is equivalent to  $\forall t \mid t_k^* \leqslant t < f_k$ ,

$$C'_{k} + \sum_{j=1}^{k-1} \left\lceil \frac{t - t_{k}^{*} + X_{j} + (1 - x_{j})(R_{j} - C_{j})}{T_{j}} \right\rceil C_{j} > t - t_{k}^{*}$$

By replacing  $t - t_k^*$  with  $\theta$ , we reach the conclusion.  $\square$ 

<sup>2</sup> This holds since the interval  $[t_k^*, t_k]$  is fully covered by the interval  $[t_1, t_k]$ .

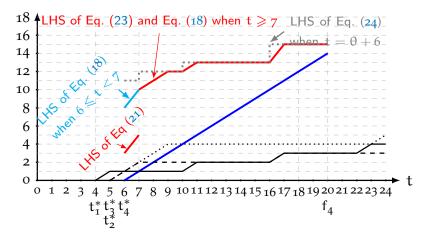


Figure 3: The workload function for the three higher-priority tasks in Example 4 when  $\varepsilon$  is very close to 0. Solid black line:  $W_1^1(t-t_1^*)$  when  $t\geqslant t_1^*$ , Dashed black line:  $W_2^0(t-t_2^*)$  when  $t\geqslant t_2^*$ , Dotted black line:  $W_3^1(t-t_3^*)$  when  $t\geqslant t_3^*$ , where the three workload functions are 0 if  $t-t_j^*<0$  for j=1,2,3, Blue line (the only linear function from t=6 in this figure):  $t-t_k^*=t-6$ , Red line (marked by Eq. (21) and Eq. (23)): left-hand side of Eq. (21) when t<7 and left-hand side of Eq. (23) and Eq. (18) when  $7\leqslant t<20$ , Purple line (marked by Eq. (18) when  $6\leqslant t<7$ ), Gray dotted line (marked by Eq. (24)) by setting  $\theta=t-6$ .

**Proof of Theorem 1.** The condition in Lemma 14 implies that the minimum  $\theta$  with  $\theta > 0$  and  $C_k' + \sum_{j=1}^{k-1} \left\lceil \frac{\theta + X_j + (1-x_j)(R_j - C_j)}{T_j} \right\rceil C_j = \theta$  is larger than or equal to  $f_k - t_k^* \geqslant f_k - t_k$  and therefore provides an upper bound on any job  $J_k$  released in any schedule  $\Psi$ . However, the condition in Lemma 14 still requires the knowledge of  $\sigma_i$ . Yet, it is straightforward to see that  $\sum_{j=1}^{k-1} \left\lceil \frac{\theta + X_j + (1-x_j)(R_j - C_j)}{T_j} \right\rceil C_j$  is maximized when  $X_j$  is the largest. Since by Lemma 5  $X_j = \sum_{\ell=j}^{k-1} x_\ell \sigma_\ell \leqslant \sum_{\ell=j}^{k-1} x_\ell S_\ell = Q_j^{\vec{x}}$ , we reach the conclusion of the correctness of Theorem 1 when replacing  $X_j$  with  $Q_j^{\vec{x}}$  in the previous equation.  $\square$ 

**Example 7** We consider Example 4 when  $\varepsilon$  is very close to 0 to illustrate Step 3 in the proof of Theorem 1. For such a case,  $t_1^* = 4$ ,  $t_2^* = 5$ ,  $t_3^* = 5$ , and  $t_4^* = 6$ . Figure 3 presents the corresponding relations of the inequalities in Step 3. As shown in Figure 3, all the inequalities in Eqs. (21), (23), (18), and (24) hold.  $\square$ 

The physical meaning of the above analysis in Theorem 1 can be found in Appendix.

As mentioned at the beginning of this section, all the lemmas and corollaries proven above are valid for any job in any schedule  $\Psi$  where all the jobs respect their deadlines. Yet, those results are still valid for the first job of task  $\tau_k$  that misses a deadline in a schedule  $\Psi$  (if such a job exists). This allows us to prove Theorem 2 below.

**Proof of Theorem 2.** By the assumption that  $R_k > T_k$ , there exists a schedule  $\Psi$  such that the response time of at least one job of  $\tau_k$  is strictly larger than  $T_k$ . Let  $J_k$  be the *first* job in the schedule  $\Psi$  that has response time larger than  $T_k$ . Suppose that  $J_k$  arrives at time  $r_k$ . When job  $J_k$  is released at time  $r_k$ , there is

no other unfinished job of task  $\tau_k$ . By Lemma 1, we can safely remove all the other jobs of task  $\tau_k$  arrived before  $r_k$  without affecting the response time of  $J_k$ . It is rather straightforward to see that removing all the other jobs of task  $\tau_k$  arrived after  $r_k$  also does not change the fact that  $J_k$  finishes after  $r_k + T_k$ . Let  $f_k$  be the time at which  $J_k$  finishes in the above schedule after removing the other jobs of task  $\tau_k$ . We know that  $f_k - r_k > T_k$ .

Then, we can follow all the procedures and steps in the proof of Theorem 1 to reach the same conclusion in Lemma 14, which implies Theorem 2 by setting  $X_j \leqslant Q_j^{\vec{x}}$  for  $j=1,2,\ldots,k-1$  since  $f_k-r_k>T_k$  and  $C_k'=C_k+S_k$ .  $\square$ 

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 3, then it is also deemed schedulable by Corollary 1.

**Lemma 15** The schedulability test of task  $\tau_k$  provided by Eq. (3) dominates that of Eq. (1).

**Proof.** For any t > 0, it is straightforward to see that

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

and by using the definition of B<sub>k</sub> (i.e., in Section 3.2), we get

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

Therefore, Eq. (3) will always have a solution which is smaller than or equal to the solution of Eq. (1). This proves the lemma.  $\Box$ 

**Lemma 16** 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. (5) 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.  $\Box$ 

**Lemma 17** The schedulability test presented in Corollary 1 dominates the schedulability test provided by Eq. (3).

**Proof.** In this proof, we first transform the worst-case response time analysis presented in Corollary  ${\tt 1}$  in a more pessimistic analysis. We then prove that this more pessimistic version of Corollary  ${\tt 1}$  provides the same solution as Eq. (3), which then proves the lemma. Due to space limitation, the proof is in Appendix.  $\Box$ 

**Theorem 3** The schedulability test presented in Corollary 1 dominates the schedulability tests provided by Equations (1), (2), and (3).

**Proof.** It is a direct application of Lemmas 15, 16 and 17.

As a corollary of this theorem, it directly follows that all the response time analyses discussed in Section 3 are in fact correct. This provides the first proof of correctness for Eq. (3), which was initially presented in [25] but never proven correct.

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

**Proof.** It directly results from the two following facts,

- (i) by Theorem 3, the schedulability test presented in Corollary 1 dominates the schedulability tests provided by Equations (1), (2), and (3);
- (ii) as proven in Section 5.1, Corollary 1 is correct.

To test the schedulability of a task  $\tau_k$ , Corollary 1 implies to test all the possible vector assignments  $\vec{x} = (x_1, x_2, \dots, x_{k-1})$  to get the tightest result (under our analysis). Therefore,  $2^{k-1}$  possible combinations should 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. (5), a good vector assignment can be derived in linear time.

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

$$\begin{split} &C_{k} + S_{k} + \sum_{i=1}^{k-1} \left[ \frac{t + \sum_{\ell=i}^{k-1} x_{\ell} S_{\ell} + (1 - x_{i}) (R_{i} - C_{i})}{T_{i}} \right] C_{i} \\ \leqslant &C_{k} + S_{k} + \sum_{i=1}^{k-1} \left( \frac{t + \sum_{\ell=i}^{k-1} x_{\ell} S_{\ell} + (1 - x_{i}) (R_{i} - C_{i})}{T_{i}} + 1 \right) C_{i} \\ =&C_{k} + S_{k} + \sum_{i=1}^{k-1} \left( U_{i} \cdot t + C_{i} + U_{i} (1 - x_{i}) (R_{i} - C_{i}) + U_{i} \sum_{\ell=i}^{k-1} x_{\ell} S_{\ell} \right) \end{split} \tag{25}$$

Moreover, using the simple algebra property that for any two vectors  $\vec{a}$  and  $\vec{b}$  of size (k-1) there is  $\sum_{i=1}^{k-1} a_i \sum_{j=i}^{k-1} b_j = \sum_{j=1}^{k-1} b_j \sum_{i=1}^{j} a_i$ , 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. (25), it holds that

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

It results that the minimum positive value for t such that

$$C_k + S_k + \sum_{i=1}^{k-1} \left( U_i t + C_i + U_i (1 - x_i) (R_i - C_i) + x_i S_i \sum_{\ell=1}^i U_\ell \right) \leqslant t$$
 (26)

is an upper bound on the worst-case response time  $R_k$  of  $\tau_k$ .

Observing Eq. (26), 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. (26) 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  $\tau_i$ ,

$$x_{i}^{lin} = \begin{cases} 1 & \text{if } U_{i}(R_{i} - C_{i}) > S_{i}(\sum_{\ell=1}^{i} U_{\ell}) \\ 0 & \text{otherwise} \end{cases}$$
 (27)

The following properties directly follow.

**Property 2** For any t > 0, the vector assignment  $\vec{x}^{lin}$  minimizes the solution to Eq. (26) among all  $2^{k-1}$  possible vector assignments.

**Theorem 5** Let  ${\tt rbf}_k(t,\vec{x}^{\it lin})$  be the left hand side of Eq. (26). Task  $\tau_k$  is schedulable under fixed-priority if

$$rbf_k(D_k, \vec{x}^{lin}) \leqslant D_k. \tag{28}$$

**Proof.** It directly follows from Corollary 1 and the fact that, by construction, Eq. (26) upper bounds Eq. (4). Note that  ${\rm rbf}_k(t,\vec{x}^{lin})$  can be expressed as  $A+\sum_{i=1}^{k-1}U_it$  with a constant A>0 (independent from t). Therefore, if the condition in Eq. (26) holds for a certain  $0< t< D_k$  with  $A+\sum_{i=1}^{k-1}U_it\leqslant t$ , then the inequality  $A+\sum_{i=1}^{k-1}U_iD_k\leqslant D_k$  also holds.  $\square$ 

Property 3 The time complexity of both deriving  $\vec{x}^{lin}$  and testing Eq. (26) is O(k).

In this section, we present experiments conducted on randomly generated task sets. Five schedulability tests are compared, namely, the suspension oblivious approach (Section 3.0.1), the modeling of suspension as release jitter (Section 3.1), the analysis that models the suspension as a blocking term (Section 3.2), 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. (27) in Section 7 (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 [12]. Let  $C_i'$  denote the sum of  $C_i$  and  $S_i$  (i.e.,  $C_i' \stackrel{def}{=} C_i + S_i$ ). The modified utilization of  $\tau_i$  is then given by  $U_i' \stackrel{def}{=} C_i'/T_i$  and the total modified utilization is  $U' \stackrel{def}{=} \sum_{i=1}^n U_i'$ . 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 value  $C_i' = U_i' \times T_i$  is then divided in the two components  $C_i$  and  $S_i$  using a random ratio  $r_i$  from a uniform distribution between a value  $r_{min}$  and  $r_{max}$  depending of the specific experiment performed. That is,  $S_i \stackrel{def}{=} r_i \times C_i'$  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 experiments.

Four different types of experiments are reported in this paper. The first one is illustrated 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 performance of the tests is 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 7 slightly increases with the number of tasks. It is the opposite behavior in 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 performance 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 performance 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 100%. For the

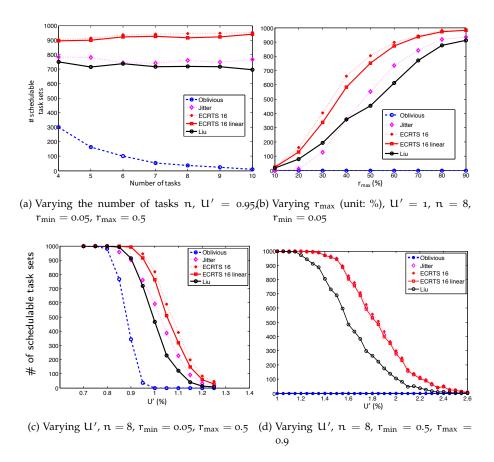


Figure 4: Number of schedulable task sets over 1000 randomly generated task sets.

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 witness the incomparability of the jitter-based and the blocking based schedulability tests.

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. 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.

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#### 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 Theorems 1 2 and Corollary 1. We show that this result analytically dominates all the existing analyses presented in Section 3, 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 7 can help accelerate the analysis while outputting good results.

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# 10

#### APPENDIX

How did Rajkumar, Sha, and Lehoczky in [29, p. 267] analyze dynamic self-suspending behaviour due to multiprocessor synchronization? The statement in [29] reads as follows:

"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  $\frac{C_k + B_k}{T_k} + \sum_{i=1}^{k-1} U_i \leqslant k(2^{\frac{1}{k}} - 1)$  to determine whether the current task allocation to the processor is schedulable."

We rephrased the wording and notation in order to be consistent with 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.

**Proof of of Lemma 1.** Since, by assumption, the worst-case response time of task  $\tau_i$  is no more than its period, any job  $\tau_{i,j}$  of task  $\tau_i$  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  $\tau_i$ , 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  $\tau_i$  are not affected by the removal of  $\tau_{i,j}$ .  $\square$ 

**Transformation from**  $\Psi$  **to**  $\Psi^1$ : Here, we present the pseudo-code to transform from the given schedule  $\Psi$  to  $\Psi^1$ :

**Proof of Lemma 9.** We first prove that  $\widehat{W}^o_j(\Delta, C_j) \geqslant \widehat{W}^o_j(\Delta, c_j^*)$  when  $c_j^* = o$ . That is, we prove that  $\widehat{W}^o_j(\Delta, C_j) \geqslant W^1_j(\Delta)$  (see Eq. (9)).

By definition,  $\rho_j = T_j - R_j + C_j$  when  $c_j^*$  is  $C_j$ . Because by assumption  $C_j \leq R_j \leq T_j$ , we have  $o \leq \rho_j \leq T_j$ . Therefore, for  $\Delta \geq T_j$ , we have  $W_j^T(\Delta) = C_j + W_j^T(\Delta - T_j) \leq C_j + W_j^T(\Delta - \rho_j) = \widehat{W}_j^0(\Delta, C_j)$  where the last equality is given by the fourth case of Eq. (9) when  $c_j^* = C_j$ . For  $o \leq \Delta < T_j$ , it is also obvious that  $\widehat{W}_j^o(\Delta, C_j) \geq \min\{\Delta, C_j\} = W_j^T(\Delta)$ .

We then prove that  $\widehat{W}^o_j(\Delta, C_j) \geqslant \widehat{W}^o_j(\Delta, c_j^*)$  for any  $o < c_j^* \leqslant C_j$  based on its definition in Eq. (9). Figure 5 provides an illustrative example for  $\widehat{W}^o_j(\Delta, c_j^*)$ . We consider three subcases:

- For  $0 \le \Delta \le C_j$ , it is obvious that  $\widehat{W}_j^o(\Delta, C_j) \ge \widehat{W}_j^o(\Delta, c_i^*)$ .
- For  $C_j < \Delta \leqslant T_j R_j + C_j$ , we have  $\widehat{W}^o_j(\Delta, C_j) = C_j$ , and from Eq (9),  $\widehat{W}^o_j(\Delta, c_j^*) = c_j^* + \max\{o, \Delta (T_j R_j + c_j^*)\} \leqslant c_j^* + C_j c_j^* = C_j$ .
- For  $T_j R_j + C_j < \Delta$ , we have  $\widehat{W}^o_j(\Delta, C_j) = C_j + W^1_j(\Delta (T_j R_j + C_j))$ . Moreover, by definition, we also know  $W^1_j(\Delta, c^*_j) \leqslant \delta + W^1_j(\Delta \delta, c^*_j)$  for

# **Algorithm 1** Transformation from $\Psi$ to $\Psi^1$

```
Input: \tau_k, T_0, T_1, and a fixed-priority preemptive schedule \Psi of \tau under the assumption R_k \leqslant T_k;
  1: pick one job J_k of task \tau_k and set r_k as the arrival time of J_k;
  2: remove all the jobs generated from \tau_k, \tau_{k+1}, \tau_{k+2}, \dots, \tau_n in the schedule \Psi, except J_k;
  3: \Psi^k \leftarrow \Psi and t_k \leftarrow r_k;
 4: for j \leftarrow k - 1 to 1 do
         let r_i be the arrival time of the last job released by \tau_i before t_{i+1} in \Psi^{j+1} and let J_i denote
  5:
         that job;
 6:
         if r<sub>i</sub> does not exist then
             \Psi^{j} \leftarrow \Psi^{j+1} and t_{i} \leftarrow t_{i+1}; {Case o}
  7:
 8:
             \Psi^{j} \leftarrow \Psi^{j+1} and remove all the jobs of task \tau_{j} released before r_{j} in schedule \Psi^{j};
  9:
             if \tau_i \in T_o then
10:
                 t_{j} \leftarrow t_{j+1}, remove J_{j}, and create an artificial job to represent the residual workload
                of J_j, executed at or after t_{j+1}; {Case 3}
12:
                 if J_j completes its execution at or before t_{j+1} then
13:
                     t_j \leftarrow t_{j+1}, remove J_j in schedule \Psi^j; {Case 2}
14:
15:
16:
                    t_j \leftarrow r_j; \{ \text{Case 1} \}
                 end if
17:
18:
             end if
         end if
19:
20: end for
21: return \Psi^1;
```

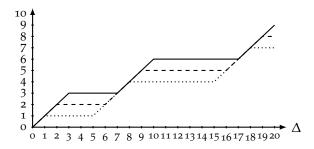


Figure 5: The workload function  $\widehat{W}_j^0(\Delta, c_j^*)$  when  $T_j = 10$ ,  $C_j = 3$ , and  $R_j = 6$ . Solid line:  $c_j^*$  is 3, Dashed line:  $c_j^*$  is 2, Dotted line:  $c_j^*$  is 1.

any  $\delta$  such that  $o < \delta \leqslant \Delta$ . Therefore, we conclude that  $\widehat{W_j^o}(\Delta, c_j^*) = c_j^* + W_j^1(\Delta - (T_j - R_j + c_j^*)) \leqslant C_j + W_j^1(\Delta - (T_j - R_j + C_j))$  by setting  $\delta$  to  $C_j - c_j^*$  in the previous inequality.

# Physical Meaning of Theorem 1

The rationale behind Theorem 1 may not be easy to be captured. A specific vector  $\vec{x}$  defines how we plan to set the release jitter for each task as follows:

- For task  $\tau_{k-1}$ , its release jitter is  $R_{k-1} C_{k-1}$  if  $x_{k-1}$  is 0 or  $S_{k-1}$  if  $x_{k-1}$  is 1.
- For task  $\tau_j$  with  $j=1,2,\ldots,k-2$ , its release jitter is  $Q_{j+1}^{\vec{x}}+R_j-C_j$  if  $x_j$  is 0 or  $Q_{j+1}^{\vec{x}}+S_j$  if  $x_j$  is 1.

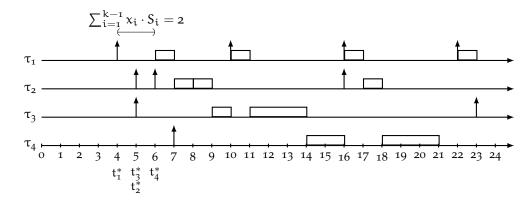


Figure 6: An illustrative example for the physical meaning of Theorem 1 for Example 4.

We use the following example to explain the physical meaning behind the setting of the release jitter of the tasks by referring to Step 3 in the proof of Theorem 1.

We consider Example 4 when  $\varepsilon$  is very close to o. For such a case,  $t_1^*=4$ ,  $t_2^*=5$ ,  $t_3^*=5$ , and  $t_4^*=6$ . We consider  $R_2=10$ . By the above setting with  $x_1=1, x_2=0, x_3=1$ , we know that

- the release jitter of task  $\tau_3$  is 1 with the first release at time  $t_4^* = 6$ ,
- the release jitter of task  $\tau_2$  is 1+10-1=10 with the first release at time  $t_4^*=6$ , and
- the release jitter of task  $\tau_1$  is 1+1=2 with the first release at time  $t_4^*=6$ . Or alternatively, we can equivalently rephrase it as follows:
  - the release jitter of task  $\tau_3$  is 0 with the first release at time  $t_3^* = 5$ ,
  - the release jitter of task  $\tau_2$  is 10-1=9 with the first release at time  $t_2^*=5$ , and
  - the release jitter of task  $\tau_1$  is 0 with the first release at time  $t_1^* = 4$ .

Therefore, the response time analysis in Lemmas 13 and 14 can be explained as follows:

A safe scenario to analyze the worst-case response time  $R_k$  of task  $\tau_k$  when  $R_k \leqslant T_k$  is 1) to release each higher-priority task  $\tau_j$  at time  $t_j^* \stackrel{\text{def}}{=} \sum_{i=1}^{j-1} x_j S_j$  with release jitter  $(1-x_j)(R_j-C_j)$ , and 2) to execute the accumulated work only after time  $t_k^*$ , where  $t_1^*$  is an arbitrary constant.

Figure 6 provides a schedule based on the above setting. Note that self-suspension does not have to be accounted any more after the above transformation. Task  $\tau_1$  is an ordinary periodic task with period 6 with the first release at time 4, and task  $\tau_3$  is an ordinary periodic task with period 18 with the first release at time 5. Task  $\tau_2$  is a jittered periodic task with period 10 and 9 time-unit jitter, starting at time 5. Therefore, the second job of task  $\tau_2$  is released at time 6 in Figure 6.

The two idle time units are used between time 4 and time 6. These two time units are *blocked* simply for accounting the self-suspension behavior in  $T_1$ , and no job is allowed to be executed in this time frame. 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  $\tau_2$  are executed back to back

from time 7 to time 9. As shown in Figure 6, the processor is busy executing the workload from time 6 to time 21. Therefore, we know that 21 - 6 = 15 is a safe upper bound of  $R_4$  in this example.

**Proof of of Lemma 17.** 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. (3), 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  $Q_i^{\vec{x}} \leqslant Q_1^{\vec{x}}$  for  $i=1,2,\ldots,k-1$ . It follows that

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

Therefore, the smallest positive value t such that

$$C_k + S_k + \sum_{i=1}^{k-1} (1 - x_i) C_i + \sum_{i=1}^{k-1} \left\lceil \frac{t + Q_1^{\vec{x}}}{T_i} \right\rceil C_i \leqslant t$$
 (29)

is always larger than or equal to the solution of Eq. (5). Suppose that  $R_k^*$  is the smallest positive value t satisfying Eq. (29) for a given vector assignment  $\vec{x}$ .

Subtituting  $(t+Q_1^{\vec{x}})$  by  $\theta$  in Eq. (29), we get that  $R_k^*$  is upper bounded by the minimum value  $(\theta-Q_1^{\vec{x}})$  greater than o (and therefore by the smallest  $\theta>0$ ) such that

$$C_{k} + S_{k} + \sum_{i=1}^{k-1} (1 - x_{i}) C_{i} + \sum_{i=1}^{k-1} \left\lceil \frac{\theta}{T_{i}} \right\rceil C_{i} \leqslant \theta - Q_{i}^{\vec{x}}$$

$$\Leftrightarrow C_{k} + S_{k} + Q_{i}^{\vec{x}} + \sum_{i=1}^{k-1} (1 - x_{i}) C_{i} + \sum_{i=1}^{k-1} \left\lceil \frac{\theta}{T_{i}} \right\rceil C_{i} \leqslant \theta$$

$$\Leftrightarrow C_{k} + S_{k} + \sum_{i=1}^{k-1} (x_{i} S_{i} + (1 - x_{i}) C_{i}) + \sum_{i=1}^{k-1} \left\lceil \frac{\theta}{T_{i}} \right\rceil C_{i} \leqslant \theta.$$
(30)

Suppose that  $R_k^{\dagger}$  is the smallest positive value  $\theta$  satisfying Eq. (30) for the given vector assignment  $\vec{x}$ . By definition,  $R_k^{\dagger} = R_k^* + Q_1^{\vec{x}} \geqslant R_k^*$  for any given vector assignment  $\vec{x}$ . Now, consider the particular vector assignment  $\vec{x}$  in which

$$x_{i} = \begin{cases} 1 & \text{if } S_{i} \leqslant C_{i} \\ 0 & \text{otherwise,} \end{cases}$$

for  $\mathfrak{i}=1,2,\ldots,k-1.$  By the definition of  $B_k$  (i.e., Section 3.2), 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. (30) thus becomes identical to Eq. (3). Therefore, if Eq. (3) deems a task set as being schedulable, so does Corollary 1.  $\Box$