Push Forward: Global Fixed-Priority Scheduling of Arbitrary-Deadline Sporadic Task Systems

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Abstract

The sporadic task model is often used to analyze recurrent execution of tasks in real-time systems. A sporadic task defines an infinite sequence of task instances, also called jobs, that arrive under the minimum inter-arrival time constraint. To ensure the system safety, timeliness has to be guaranteed in addition to functional correctness, i.e., all jobs of all tasks have to be finished before the job deadlines. We focus on analyzing arbitrary-deadline task sets on a homogeneous (identical) multiprocessor system under any given global fixed-priority scheduling approach and provide a series of schedulability tests with different tradeoffs between their time complexity and their accuracy. Under the arbitrary-deadline setting, the relative deadline of a task can be longer than the minimum inter-arrival time of the jobs of the task. We show that global deadline-monotonic (DM) scheduling has a speedup bound of $3 - 1/M$ against any optimal scheduling algorithms, where $M$ is the number of identical processors, and prove that this bound is asymptotically tight.

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1 Introduction

The sporadic task model is the basic task model in real-time systems, where each task $\tau_i$ releases an infinite number of task instances (jobs) under its minimum inter-arrival time (period) $T_i$ and is further characterized by its relative deadline $D_i$ and its worst-case execution time $C_i$. The sporadic task model has been widely adopted in real-time systems.
A sporadic task defines an infinite sequence of task instances, also called jobs, that arrive under the minimum inter-arrival time constraint, i.e., any two consecutive releases of jobs of task $\tau_i$ are temporally separated by at least $T_i$. When a job of task $\tau_i$ arrives at time $t$, it must finish no later than its absolute deadline $t + D_i$. If all tasks release their jobs strictly periodically with period $T_i$, the task model is the well-known Liu and Layland task model [33]. A sporadic task set is called with 1) implicit deadlines, if the relative deadlines are equal to their minimum inter-arrival times, 2) constrained deadlines, if the minimum inter-arrival times are no less than their relative deadlines, and 3) arbitrary deadlines, otherwise.

To schedule such task sets on a multiprocessor platform, three paradigms have been widely adopted: partitioned, global, and semi-partitioned multiprocessor scheduling. The partitioned scheduling approach partitions the tasks statically among the available processors, i.e., a task executes all its jobs on the assigned processor. The global scheduling approach allows a job to migrate from one processor to another at any time. The semi-partitioned scheduling approach decides whether a task is divided into subtasks statically and how each task/subtask is then assigned to a processor. A comprehensive survey of multiprocessor scheduling for real-time systems can be found in [23].

We focus on global fixed-priority preemptive scheduling on $M$ identical processors, i.e., unique fixed priority levels are statically assigned to the tasks and at any point in time the $M$ highest-priority jobs in the ready queue are executed. Hence, the schedule is workload-conserving. The response time of a job is defined as its finish time minus its arrival time. The worst-case response time of a task is an upper bound on the response times of all the jobs of the task and can be derived by a (worst-case) response time analysis for a sporadic task under a given scheduling algorithm. Verifying whether a set of sporadic tasks can meet their deadlines by a scheduling algorithm is called a schedulability test, i.e., verifying if the (worst-case) response time is smaller than or equal to the relative deadline.

### 1.1 Related Work

For uniprocessor systems, i.e., $M=1$, the exact schedulability test and the (tight) worst-case response time analysis by using busy intervals were provided by Lehoczky [32]. Several approaches have been proposed to reduce the time complexity, e.g., [35]. Bini and Buttazzo [12] proposed a framework of schedulability tests that can be tuned to balance the time complexity and the acceptance ratio of the schedulability test for uniprocessor sporadic task systems. To achieve polynomial-time schedulability tests and response time analyses, Lehoczky [32] proposed a utilization upper bound for a set of sporadic arbitrary-deadline tasks under fixed-priority scheduling. The linear-time response-time bound for fixed-priority systems was first proposed by Davis and Burns [22], and later improved by Bini et al. [14, 15] and Chen et al. [18]. The computational complexity of the schedulability test problem and the worst-case response time analysis in uniprocessor systems for different variances can be found in [16, 25, 24, 27, 26].

In this paper, we will implicitly assume multiprocessor systems, i.e., $M \geq 2$. Many results are known for constrained-deadline ($D_i \leq T_i$) and implicit-deadline task systems ($D_i = T_i$) on identical multiprocessor platforms, e.g., [2, 5, 30, 1, 7, 18]. For details, please refer to the survey by Davis and Burns [23]. Unfortunately, deriving exact schedulability tests under multiprocessor global scheduling is much harder than deriving them for uniprocessor systems due to the lack of concrete worst-case scenarios that can be constructed efficiently. Most results in the literature focus on sufficient schedulability tests. Exceptions are the exhaustive search under discrete time parameters by Baker and Cirinei [4], finite automata under discrete time parameters by Geeraerts et al. [29], and hybrid finite automata by
Sun and Lipari [36]. Specifically, Geeraerts et al. [29] showed that the schedulability test formulation by Baker and Cirinei [4] is PSPACE-Complete.

Regarding global fixed-priority scheduling for arbitrary-deadline task systems, several sufficient schedulability tests and safe worst-case response time analyses have been proposed, e.g., [3, 4, 8, 9, 30, 37, 31]. Baker [3] designed a test based on certain properties to characterize a problem window. Baruah and Fisher [8, 9] used different annotations to extend the analysis window and derived corresponding exponential-time schedulability tests. The first worst-case response-time analysis for arbitrary-deadline task systems was proposed by Guan et al. [30], where the authors used the insight proposed by Baruah [5] to limit the number of carry-in jobs, and then apply the workload function proposed by Bertogna et al. [11] to quantify the requested demand of higher-priority tasks. Unfortunately, it has recently been shown by Sun et al. [37] that this analysis in [30] is optimistic. In addition, Sun et al. [37] derived a complex carry-in workload function for the response time analysis where all possible combinations of carry-in and non-carry-in functions have to be explicitly enumerated. However, their method is computationally intractable since the time complexity is exponential. Huang and Chen [31] proposed a more precise quantification for the number of carry-in jobs of a task than the bounds used in the tests provided in [3, 9]. They also presented a response time bound for arbitrary-deadline tasks under global scheduling in multiprocessor systems with linear-time complexity.

1.2 Our Contribution

We consider arbitrary-deadline sporadic task systems, which is the most general case of the sporadic real-time task model. To quantify the performance loss due to efficient schedulability tests and the non-optimality of scheduling algorithms, we will adopt the notion of speedup factors/bounds, also known as resource augmentation factors/bounds. Table 1 summarizes the state-of-the-art speedup bounds for the global deadline-monotonic (DM) scheduling, one specific global fixed-priority scheduling algorithm. Under global DM, a task \( \tau_i \) has higher priority than task \( \tau_j \) if \( D_i \leq D_j \), in which ties are broken arbitrarily. The authors note that the proof by Lundberg [34] seems incomplete. However, the concrete task set in [34] provides the lower bound 2.668 of the speedup factors for global DM. Moreover, Andersson [1] showed that global slack monotonic scheduling has a speedup bound of \( \frac{3 + \sqrt{5}}{2} \approx 2.6181 \) for implicit-deadline task systems. However, no better global fixed-priority scheduling algorithms with respect to speedup factors are known for constrained-deadline and arbitrary-deadline task systems.

Our Contributions: Table 1 summarizes the related results and the contribution of this paper for multiprocessor global fixed-priority preemptive scheduling. We improve the best known results by Baruah and Fisher [8] with respect to the speedup bounds. Our contributions are:

- For any global fixed-priority preemptive scheduling, we provide a series of schedulability tests with different tradeoffs between time complexity and accuracy in Section 3 and Section 4.
- We show that the global deadline-monotonic scheduling algorithm has a speedup factor \( 3 - 1/M \) with respect to the optimal multiprocessor scheduling policies when considering task systems with arbitrary deadlines. This improves the analyses by Fisher and Baruah with respect to the speedup bounds, i.e., \( 4 - 1/M \) [9] and 3.73 [8].
- We show that all the schedulability tests we provide in this paper analytically dominate the tests by Baruah and Fisher [8] for global DM. We also show that global DM has
#### 2 System Model, Definitions, and Assumptions

We consider an arbitrary-deadline sporadic task set T with N tasks executed on M ≥ 2 identical processors based on global fixed-priority preemptive scheduling. We assume that the priority levels of the tasks are unique (and given) and that τ_i has higher priority than task τ_j if i < j. When there is only one processor, i.e., M = 1, the existing results discussed in Section 1.1 can be adopted, and our analysis here cannot be applied. We will implicitly use the assumption M ≥ 2 in the paper.

By definition, M is an integer. In addition to C_i, T_i, D_i, we also define the utilization U_i of task τ_i as C_i/T_i. We will implicitly assume that D_i > 0, C_i > 0, T_i > 0, C_i/D_i ≤ 1, and U_i ≤ 1 ∀τ_i in this paper. Moreover, intra-task parallelism is not allowed. At most one job of task τ_i can be executed on at most one processor at each instant in time, regardless of the number of the jobs of task τ_i awaiting for execution and the number of idle processors. We denote the set of natural numbers as N.

##### 2.1 Resource Augmentation

We assume the original platform speed is 1. Therefore, running the platform at speed s implies that the worst-case execution time of task τ_i becomes C_i/s. A scheduling algorithm \( A \) has a speedup bound s with respect to the optimal schedule, if it guarantees to always produce a feasible solution when 1) each processor is sped up to run at s times of the original speed of the platform and 2) the task set T can be feasibly scheduled on the original M identical processors, i.e., running at speed 1.

We will use the negation of the above definition to quantify the failure of algorithm \( A \): If \( A \) fails to ensure that all the tasks in T meet their deadlines, then no feasible multiprocessor schedule exists when each processor is slowed down to run at speed 1/s.

##### 2.2 Definitions and Necessary Condition

We define the following notation according to the task system and the priority assignment:

- density \( \delta_i \) of task \( \tau_i \): \( \delta_i = C_i / \min\{D_i, T_i\} \)
- maximum density \( \delta_{\text{max}}(k) \) among the first k tasks: \( \delta_{\text{max}}(k) = \max_{i=1}^{k} \delta_i \)
- maximum between the utilization of the higher-priority tasks and the density of task \( \tau_k \): \( U_{\text{max}}^{D_k} = \max\{\text{max}_{i=1}^{k-1} U_i, \delta_k\} \)
- demand bound function \([10]\) \( \text{DBF}(\tau_i, t) \) of task \( \tau_i \), further explained in Definition 2.1
- load \( \text{LOAD}(k) \) of the first k tasks: \( \text{LOAD}(k) = \max_{t>0} \sum_{i=1}^{k} \text{DBF}(\tau_i, t) \)
The demand bound function $\text{DBF}(\tau_i, t)$ defines the execution time task $\tau_i$ must finish for any interval length $t$ to ensure its timing correctness.

Definition 2.2 (Workload function). Let $\text{work}_i(t)$ be a workload function, representing the maximum amount of time for sequentially executing the jobs of task $\tau_i$ released in time interval $[a, a + t)$, i.e., jobs released before $a$ are not considered. For any $t \geq 0$

$$\text{work}_i(t) = \left( \frac{t}{T_i} \right) C_i + \min \left\{ C_i, t - \left\lfloor \frac{t}{T_i} \right\rfloor T_i \right\}. \quad (2)$$

For notational brevity, we set $\text{work}_i(t)$ to $-\infty$ if $t < 0$.

The workload function $\text{work}_i(t)$ defined above is a piecewise function, i.e., linear in intervals $[\ell T_i, (\ell + 1)T_i]$ with a slope $1$ and constant, $(\ell + 1)C_i$, in intervals $[\ell T_i + C_i, (\ell + 1)T_i]$ for any non-negative integer $\ell$. Two examples of the workload function are illustrated in Figure 2 in Section 3. To prove the speedup bound, we will utilize the following necessary condition.

Lemma 2.3. A task set $\mathbf{T}$ with $N$ tasks is not schedulable by any multiprocessor scheduling algorithm when the $M$ processors are running at any speed $s$, if

$$\max \left\{ \max_{t \geq 0} \frac{\sum_{\tau_i \in \mathbf{T}} \text{DBF}(\tau_i, t)}{Mt}, \frac{\sum_{\tau_i \in \mathbf{T}} U_i}{M}, \delta_{\max}(N) \right\} > s. \quad (3)$$

Proof. This is widely used based on a reformulation in the literature, e.g., [8, 9].

2.3 Analysis Based on DBFs

Baruah and Fisher in [8] provided a schedulability test for task $\tau_k$ under global deadline-monotonic (DM) scheduling that is based on the Demand Bound Functions (DBF), assuming that the tasks are sorted according to DM order already; i.e., $D_1 \leq D_2 \leq \ldots \leq D_N$.

Theorem 2.4 (Baruah and Fisher [8], revised in [17]). Let $\mu_k$ be defined as $M - (M - 1)\delta_{\max}(k)$. Task $\tau_k$ is schedulable under global DM if

$$2\text{LOAD}(k) + \left( \mu_k - 1 \right) \delta_{\max}(k) \leq \mu_k. \quad (4)$$

3 Schedulability Test by Pushing Forward

In this section, we provide several conditions for the schedulability of task $\tau_k$ under a given preemptive global fixed-priority scheduling algorithm. They lead to a sufficient schedulability test for $\tau_k$, assuming that the schedulability of the tasks $\tau_1, \tau_2, \ldots, \tau_{k-1}$ under the given

\footnote{The original proof by Baruah and Fisher [8] had a mathematical flaw in their Lemma 3, i.e., setting $\mu_k$ to $M - (M - 1)\delta_k$. It can be fixed by setting $\mu_k$ to $M - (M - 1)\delta_{\max}(k)$.}
algorithm is already verified. This means that for all tasks \( \tau_i \) with \( i < k \) the worst-case response time is at most \( D_i \). Therefore, the test should be applied for all tasks, i.e., from the highest-priority task to the lowest-priority task, to ensure the schedulability of the task set under the (specified/given) global fixed-priority scheduling. As the test presented here has a high time complexity, we provide more efficient tests in Section 4.

### 3.1 Analysis Window Extension

We analyze the schedulability of \( \tau_k \) by looking at the intervals where \( \tau_k \) is active in the schedule \( S \) provided by the global fixed-priority scheduling algorithm according to the following definition:

#### Definition 3.1 (active task).

For a schedule \( S \), a task \( \tau_i \) is active at time \( t \), if there is (at least) one job of \( \tau_i \) that has arrived before or at \( t \) and has not finished yet at time \( t \). \( \blacksquare \)

The schedulability conditions are proved by using contrapositive. Suppose a schedule \( S \) produced by the given global fixed-priority scheduling algorithm and that \( t_d \) is the earliest (absolute) deadline at which a job of task \( \tau_k \) misses its deadline. Let \( t_a \) be the time instant in \( S \) such that \( \tau_k \) is continuously active in the time interval \([t_a, t_d)\) and is not active immediately prior to \( t_a \). By definition, \( t_a \) must be the arrival time of a job of task \( \tau_k \). Suppose that \( t_d \) is the absolute deadline of the \( \ell \)-th job of task \( \tau_k \) that arrived in the time interval \([t_a, t_d)\).

Therefore, as \( \tau_k \) is a sporadic task, \( t_d - t_a \geq (\ell - 1)T_k + D_k \). For notational brevity, we define \( D'_k = (\ell - 1)T_k + D_k \) and \( C'_k = \ell C_k \).

We remove all the jobs of task \( \tau_k \) that arrive before \( t_a \) and all the jobs with priorities lower than \( \tau_k \) from the schedule \( S \). The schedule of task \( \tau_k \) remains unchanged in the resulting (new) schedule \( S' \), due to the preemptiveness of the global fixed-priority scheduling algorithm. Let \( C''_k \) be the amount of time that task \( \tau_k \) is executed from \( t_a \) to \( t_d \). Since the \( \ell \)-th job of task \( \tau_k \) misses its deadline, we know that \( C''_k < \ell C_k = C'_k \). We now introduce three functions that are defined for any \( t \leq t_d \).

- Let \( E(t, t_d) \) be the amount of workload (sum of the execution times) of the higher-priority jobs, i.e., from \( \tau_1, \tau_2, \ldots, \tau_{k-1} \), executed in the time interval \([t, t_d)\) in schedule \( S \).
- Let \( W(t, t_d) \) be \( C''_k + E(t, t_d) \).
- Let \( \Omega(t, t_d) \) be \( \frac{W(t, t_d)}{t_d - t} \).

Those definitions and the deadline miss of task \( \tau_k \) at time \( t_d \) lead to the following lemma.

#### Lemma 3.2. Since \( \tau_k \) misses its deadline at \( t_d \) in \( S \), the following conditions hold:

\[ E(t_a, t_d) \geq M \times (t_d - t_a - C''_k) \]  \hspace{1cm} (5)
\[ W(t_a, t_d) > M \times (t_d - t_a) - (M - 1)C''_k \]  \hspace{1cm} (6)
\[ \Omega(t_a, t_d) > M - (M - 1) \times \frac{C''_k}{D'_k} \]  \hspace{1cm} (7)

**Proof.** Since task \( \tau_k \) is active from \( t_a \) to \( t_d \) and is only executed for exactly \( C''_k \) amount of time, we know that all \( M \) processors must be busy executing other higher-priority jobs for at least \( t_d - t_a - C''_k \) amount of time. Therefore, the amount of workload \( E(t_a, t_d) \) of the higher-priority jobs executed in the time interval \([t_a, t_d)\) must be at least \( M \times (t_d - t_a - C''_k) \),
This means, \( i.e., the condition in Eq. (7). \)

The condition in Eq. (5) is widely used in the form of \( W(t_a, t_d) \), where the last inequality is due to \( M \geq 2 \) and \( C_k > C^*_k \). This leads to the conditions in Eq. (6).

\[
\Omega(t_a, t_d) \geq M - (M - 1) \frac{C_k'}{C_k} \geq M - (M - 1) \frac{C_k'}{D_k'},
\]

i.e., the condition in Eq. (7).

Although the interval \([t_a, t_d]\) can already be used for constructing the schedulability tests, researchers have tried to push the interval of interest towards \([t_0, t_d]\) for some \( t_0 \leq t_a \) based on certain properties, e.g., \([31, 9, 8]\). Such extensions have been shown to provide better quantifications of the interfering workload from the higher-priority tasks. In our analysis, we will use a similar extension strategy as suggested by Baruah and Fisher [8] based on a user-specified parameter \( \rho \).

The following definition and lemmas are from [8]. Figure 1 provides an illustration of our notation based on the above definitions.

**Definition 3.3.** Suppose that \( \mu_k = M - (M - 1) \rho \) for a certain \( \rho \) with \( 1 \geq \rho \geq \frac{C_k'}{D_k'} \). For the schedule \( S \), let time instant \( t_0 \) be the smallest value of \( t \leq t_a \) such that \( \Omega(t, t_d) \geq \mu_k \). This means, \( \Omega(t, t_d) < \mu_k \) for any \( t < t_0 \).

**Lemma 3.4.** If \( \tau_k \) misses its deadline at \( t_d \), for any \( \rho \) with \( 1 \geq \rho \geq \frac{C_k'}{D_k'} \), the time \( t_0 \), as defined in Definition 3.3, always exists with \( \Omega(t_0, t_d) \geq \mu_k \) and \( t_0 \leq t_a \).

**Proof.** By Eq. (7) from Lemma 3.2 and \( \rho \geq \frac{C_k'}{D_k'} \), we know

\[
\Omega(t_a, t_d) > M - (M - 1) \times \frac{C_k'}{D_k'} \geq M - (M - 1) \rho = \mu_k.
\]

Therefore, such a time instant \( t_0 \leq t_a \) exists, at least when the system starts.

**Definition 3.5 (carry-in task).** A task \( \tau_i \) is a carry-in task in the schedule \( S \), if \( \tau_i \) is continuously active in a time interval \([t_i, t_0 + \varepsilon]\), for \( t_i < t_0 \) and an arbitrarily small \( \varepsilon > 0 \).

**Lemma 3.6.** For \( 1 \geq \rho \geq \frac{C_k'}{D_k'} \), there are at most \([M - (M - 1) \rho] - 1\) carry-in tasks at \( t_0 \) in schedule \( S \).

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2 The condition in Eq. (5) is widely used in the form of \( E(t_a, t_d) > M \times (t_d - t_a - tC_k) \). Here, since we will use \( C^*_k \), the correct form is with \( \geq \).
3.2 Analysis Based on Workload Functions

By extending the interval of interest to \([t_0, t_d]\), Baruah and Fisher provided the schedulability test shown in Theorem 2.4 in this paper. However, they analyzed the workload in \([t_0, t_d]\) based on the DBFs by using the function \(\text{LOAD}(k)\) as an approximation, which will be shown pessimistic in Corollary 5.1 in Section 5. Moreover, their final analysis can only be applied for global DM. We will carefully analyze the workload executed in \([t_0, t_d]\) to ensure that the analytical accuracy is better preserved and that the analysis can be used for any global fixed-priority preemptive scheduling. We will demonstrate that our analysis dominates the analysis by Baruah and Fisher [8] in Corollary 5.1.

For the analysis before Theorem 3.10, we will assume that \(\rho\) is given and \(t_0\) is already defined. According to Lemma 3.6, at time \(t_0\) at most \([M - (M - 1)\rho] - 1\) tasks are active in schedule \(S\). We quantify their contribution to the executed workload in time interval \([t_0, t_d]\) with two different forms from Lemma 3.7, denoted by \(\omega^{\text{heavy}}(t_d - t_0)\), and from Lemma 3.8, denoted by \(\omega^{\text{light}}(t_d - t_0)\). While Lemma 3.7 can be used in general, Lemma 3.8 only holds if \(U_i \leq \rho\). After these workload functions are detailed and explained, we will show their relationship in Lemma 3.9. Then, we will explain how they can be used and detail the constructed schedulability test in Theorem 3.10 based on the above concepts.

\textbf{Lemma 3.7.} If all jobs of a higher-priority task \(\tau_i\) meet their deadlines, the upper bound \(\omega^{\text{heavy}}(\Delta)\) on the workload of task \(\tau_i\) executed from \(t_0\) to \(t_d\) with \(\Delta = t_d - t_0\) in schedule \(S\) is at most:

\[\omega^{\text{heavy}}(\Delta) = \text{work}_i(\Delta + D_i).\] (8)

\textbf{Proof.} Since all jobs of \(\tau_i\) meet their deadlines, the jobs of \(\tau_i\) executed in \([t_0, t_d]\) must arrive in the time interval \((t_0 - D_i, t_d]\). Therefore, the workload of task \(\tau_i\) that can be sequentially executed is upper bounded by the workload function with length \(t_d - (t_0 - D_i) = \Delta + D_i\).

The key improvement achieved in this paper is due to the following Lemma 3.8 to safely bound the workload of a light task.

Figure 2 demonstrates the workload function for different cases in Lemma 3.8, together with a linear approximation that will be presented in Lemma 4.3. For the workload function defined in Eq. (9), informally speaking, the workload defined by \((p_2 + 1)C_i + \max\{0, C_i - \rho(T_i - q_2)\}\) can be imagined as if 1) there is an offset for \(C_i\) amount of execution time at beginning of the interval, and 2) the workload in each period starting from \(C_i + p_2 T_i\) to \(C_i + (p_2 + 1) T_i\) is pushed to the end of the period with a slope \(\rho\). For example, in Figure 2(b), the offset is 3, the workload increases from 3 at time 7 to 6 at time 13 with a slope \(\rho = 0.5\), the workload increases from 6 at time 17 to 9 at time 23 with a slope \(\rho = 0.5\), etc.

\textbf{Lemma 3.8.} If all jobs of a higher-priority task \(\tau_i\) meet their deadlines and \(U_i \leq \rho \leq 1\), the upper bound \(\omega^{\text{light}}(\Delta)\) on the workload of task \(\tau_i\) executed from \(t_0\) to \(t_d\) with \(\Delta = t_d - t_0\) in schedule \(S\) is:

\[\omega^{\text{light}}(\Delta) = \begin{cases} \Delta & \text{if } 0 < \Delta \leq C_i \\ \max \left\{ \text{work}_i(\Delta), \left( p_2 + 1 \right) C_i + \max\{0, C_i - \rho(T_i - q_2)\} \right\} & \text{if } \Delta > C_i \end{cases}\] (9)

where \(p_2 = \lfloor (\Delta - C_i)/T_i \rfloor - 1\) and \(q_2 = \Delta - C_i - p_2 T_i\).

\textbf{Proof.} As the case when \(0 < \Delta \leq C_i\) is due to the definition, let \(\Delta > C_i\) for the rest of the proof. Based on the schedule \(S\), let \(t_i < t_0\) be the time instant such that task \(\tau_i\) is
Let $\phi_i$ be $t_0 - t_i$. By the definition of $t_i$, if it exists, there are at most $\left\lceil \frac{t_i}{T_i} \right\rceil$ jobs of task $\tau_i$ executed in time interval $[t_i, t_0]$. For the rest of the proof, we only consider that $t_i$ exists and that $\Delta > C_i$. By definition, $t_i$ must be the arrival time of a job of task $\tau_i$. Moreover, due to the definition of $t_0$ in Definition 3.3, we know that $\Omega(t_i, t_d) < M - (M - 1)\rho$. Since $\Omega(t_i, t_d) < M - (M - 1)\rho$ and $\Omega(t_0, t_d) \geq M - (M - 1)\rho$, we have

$$W(t_0, t_d) = \Omega(t_0, t_d) \cdot (t_d - t_0) \geq (t_d - t_0)\mu_k = \Delta\mu_k$$

(10)

$$W(t_i, t_d) = \Omega(t_i, t_d) \cdot (t_d - t_i) < (t_d - t_i)\mu_k = (\Delta + \phi_i)\mu_k$$

(11)

Subtracting Eq. (11) by Eq. (10), we have $W(t_i, t_d) - W(t_0, t_d) < \phi_i\mu_k$, i.e., in schedule $S$ the workload executed in time interval $[t_i, t_0]$ is strictly less than $\phi_i\mu_k$. Suppose that $y_i$ is the amount of time that task $\tau_i$ is executed in time interval $[t_i, t_0]$, i.e., task $\tau_i$ is active but blocked by other higher-priority jobs for $\phi_i - y_i$ amount of time in this time interval. When task $\tau_i$ is blocked in global fixed-priority scheduling, all the $M$ processors are executing other jobs. The workload executed in time interval $[t_i, t_0]$ is at least $M(\phi_i - y_i) + y_i$. Therefore, in the above discussion, we know that

$$M(\phi_i - y_i) + y_i < \phi_i\mu_k = \phi_i(M - (M - 1)\rho) \Rightarrow y_i > \rho\phi_i,$$

(12)

since $M \geq 2$. At time $t_0$, the remaining execution time of the jobs of task $\tau_i$ that arrived before $t_0$ in schedule $S$ is at most $[\phi_i/T_i]C_i - \rho\phi_i$. Note that the existence of $t_i$ in our definition means that $[\phi_i/T_i]C_i - y_i > 0$, i.e., $[\phi_i/T_i]C_i - \rho\phi_i > 0$.

The workload of task $\tau_i$ that is executed in the time interval $[t_i, t_d]$ in schedule $S$ is at most $\text{work}_k(t_d - t_i) = \text{work}_k(\Delta + \phi_i)$. The workload of task $\tau_i$ that is executed in the time interval $[t_i, t_0]$ is at least $y > \rho\phi_i$. Therefore, the workload of task $\tau_i$ that is executed in the time interval $[t_0, t_d]$ in schedule $S$ is upper bounded by $\text{work}_k(\Delta + \phi_i) - \rho\phi_i$.

The rest of the proof is to provide an upper bound of $\text{work}_k(\Delta + \phi_i) - \rho\phi_i$ for any arbitrary $\phi_i > 0$. The proof involves some detailed manipulations of the workload function. Before proceeding, we explain two basic properties of the workload function here by inspecting the periodicity of the workload function $\text{work}_k(t)$ where $p = \lceil t/T_i \rceil$, a non-negative integer:

For $t = pT_i + x$ with $0 \leq x$, the recursion $\text{work}_k(pT_i + x) = pC_i + \text{work}_k(x)$ holds.

For $t = pT_i + x$ with $0 \leq x \leq C_i$, the simplification $\text{work}_k(pT_i + x) = pC_i + x$ holds.
To identify the exact value of \( \text{work}_i(\Delta + \phi_i) \), we define the following variables \( p_1, p_2, q_1 \), and \( q_2 \) for brevity:

Let \( p_1 \) be \( \lceil \phi_i/T_i \rceil - 1 \) and \( q_1 \) be \( \phi_i - p_1 T_i \), i.e., \( p_1 + 1 \) is the number of jobs of task \( \tau_i \) that can be released in \( [t_i, t_0] \). By definition \( \phi_i > 0 \), which implies that \( p_1 \) is a non-negative integer, \( 0 < q_1 \leq T_i \), and \( \phi_i = p_1 T_i + q_1 \).

Let \( p_2 \) be \( \lceil (\Delta - C_i)/T_i \rceil - 1 \) and \( q_2 \) be \( \Delta - C_i - p_2 T_i \), i.e., \( p_2 + 1 \) is the number of jobs of task \( \tau_i \) that can be released in \( [t_0 + C_i, t_q] \). Due to the assumption \( \Delta > C_i \), we know that \( p_2 = 0 \) is a non-negative integer, \( 0 < q_2 \leq T_i \), and \( \Delta - C_i = p_2 T_i + q_2 \).

By the above definition, we achieve \( \phi_i + \Delta = (p_1 + p_2) T_i + q_1 + q_2 + C_i \), and

\[
\text{work}_i(\Delta + \phi_i) = \text{work}_i((p_1 + p_2) T_i + q_1 + q_2 + C_i) - \rho(p_1 T_i + q_1)
\]

\[
= \text{work}_i(p_2 T_i + q_1 + q_2 + C_i) + p_1 C_i - \rho(p_1 T_i + q_1)
\]

\[
= \text{work}_i(p_2 T_i + q_1 + q_2 + C_i) + p_1 U_i T_i - \rho(p_1 T_i + q_1)
\]

\[
\leq \text{work}_i(p_2 T_i + q_1 + q_2 + C_i) - \rho q_1
\]

(13)

where the inequality is due to the assumption that \( 0 \leq U_i \leq \rho \). We will prove that the right-hand side of Eq. (9) is a safe upper bound on the condition in Eq. (13). By the definition of \( q_1 \) and \( q_2 \), we know that \( 0 \leq q_1 + q_2 \leq 2T_i \), i.e., \( C_i \leq p_2 T_i + q_1 + q_2 + C_i \leq 2T_i + C_i \).

Depending on the value of \( q_1 + q_2 \), there are four cases for different (linear or constant) segments of \( \text{work}_i(p_2 T_i + q_1 + q_2 + C_i) \) to be analyzed:

**Case 1:** \( 0 \leq q_1 + q_2 \leq T_i - C_i \): That is, \( p_2 T_i + C_i \leq p_2 T_i + q_1 + q_2 + C_i \leq p_2 T_i + T_i \).

Therefore, \( \text{work}_i(p_2 T_i + C_i) \leq \text{work}_i(p_2 T_i + q_1 + q_2 + C_i) \leq \text{work}_i(p_2 T_i + T_i) \).

Since \( \text{work}_i(p_2 T_i + C_i) = \text{work}_i(p_2 T_i + T_i) = (p_2 + 1) C_i \), we have

RHS. of Eq. (13) =\( (p_2 + 1) C_i - \rho q_1 \leq \text{work}_i(p_2 T_i + C_i + q_2) = \text{work}_i(\Delta) \),

where \( \leq \) is due to \( \rho \geq 0 \) and \( q_1 > 0 \).

**Case 2:** \( T_i - C_i < q_1 + q_2 \leq T_i \): By definition, when \( p_2 \) is a non-negative integer and \( 0 < x \leq C_i \), \( \text{work}_i((p_2 + 1) T_i + x) = (p_2 + 1) C_i + x \). By \( T_i - C_i < q_1 + q_2 \leq T_i \), we know that \( (p_2 + 1) T_i < p_2 T_i + q_1 + q_2 + C_i \leq (p_2 + 1) T_i + C_i \). Therefore, \( \text{work}_i(p_2 T_i + q_1 + q_2 + C_i) = (p_2 + 1) C_i + (p_2 T_i + q_1 + q_2 + C_i) - (p_2 + 1) T_i = (p_2 + 1) C_i + (q_1 + q_2 + C_i - T_i) \). Let \( \eta \) be \( T_i - (q_1 + q_2) \). By definition \( \eta \geq 0 \).

Therefore, \( \text{RHS. of Eq. (13)} = (p_2 + 1) C_i + (C_i - \eta - \rho(T_i - q_2 - \eta)) = (p_2 + 1) C_i + (C_i - \rho(T_i - q_2)) + \eta(\rho - 1) \leq (p_2 + 1) C_i + \max\{0, C_i - \rho(T_i - q_2)\} \),

where \( \leq \) is due to \( 0 \leq \rho \leq 1 \) and \( \eta \geq 0 \).

**Case 3:** \( T_i < q_1 + q_2 \leq 2T_i - C_i \): Thus, \( \text{work}_i(p_2 T_i + q_1 + q_2 + C_i) = (p_2 + 2) C_i \), and

RHS. of Eq. (13) =\( (p_2 + 1) C_i + C_i - \rho q_1 \leq (p_2 + 1) C_i + \max\{0, C_i - \rho(T_i - q_2)\} \),

where \( \leq \) is due to \( \rho \geq 0 \) and \( q_1 + q_2 > T_i \).

**Case 4:** \( 2T_i - C_i < q_1 + q_2 \leq 2T_i \): In this case \( \text{work}_i(p_2 T_i + q_1 + q_2 + C_i) \) is equal to \( (p_2 + 2) C_i + (q_1 + q_2 + C_i - 2T_i) \), similar to the analysis in Case 2. Let \( \eta \) be \( 2T_i - (q_1 + q_2) \).

By definition \( \eta \geq 0 \).

Therefore, \( \text{RHS. of Eq. (13)} = (p_2 + 1) C_i + 2C_i - \eta - \rho(2T_i - q_2 - \eta) = (p_2 + 1) C_i + C_i + T_i(U_i - \rho) - \eta(1 - \rho) - \rho(T_i - q_2) \leq (p_2 + 1) C_i + \max\{0, C_i - \rho(T_i - q_2)\} \),
where \( \leq \) is due to \( 0 < U_i \leq \frac{C_i}{C_k} \leq \rho \leq 1 \) and \( \eta \geq 0 \), i.e., \( U_i - \rho \leq 0 \) and \( -\eta(1 - \rho) \leq 0 \).

Since \( 0 < q_1 + q_2 < 2T_i \), we know that \( w_{\text{heavy}}(\Delta) \) is a safe upper bound for Case 1 and that 

\( (p_1 + 1) C_i + \max\{0, C_i - \rho(T_i - q_2)\} \) is a safe upper bound for the other cases, and we reach the conclusion of this lemma.

\[ \blacktriangleleft \]

**Lemma 3.9.** If \( U_i \leq \rho \), then \( \omega_i^{\text{heavy}}(\Delta) \geq \omega_i^{\text{light}}(\Delta) \) for all \( \Delta > 0 \).

**Proof.** This inequality can be proved formally, but can also be derived by following the definitions. When \( 0 < \Delta \leq C_i \), the inequality holds naturally. In the proof of Lemma 3.8, the workload of task \( \tau_i \) that is executed in the time interval \([t_i, t_d)\) in schedule \( S \) is at most 

\[ \text{work}_i(t_d - t_i) = \text{work}_i(\Delta + \phi_i) \] 

Since \( \phi_i \leq D_i \), we know that \( \omega_i^{\text{light}}(\Delta) \leq \text{work}_i(\Delta + \phi_i) \leq \omega_i^{\text{heavy}}(\Delta) \). \[ \blacktriangleleft \]

Here is a short summary of the information provided by Lemmas 3.6, 3.7, and 3.8.

- According to Lemma 3.6, at time \( t_0 \), there are at most \( [M - (M - 1)\rho] - 1 = [\mu_k] - 1 \) carry-in tasks.
- Among the \([\mu_k] - 1\) carry-in tasks, there are two types of carry-in tasks, i.e., \( \text{heavy} \) and \( \text{light} \) tasks. A light carry-in task \( \tau_i \) can be described by \( \omega_i^{\text{light}}(\Delta) \) from Eq. (9) if the utilization is no more than \( \rho \) and a heavy carry-in task \( \tau_i \) can be described by \( \omega_i^{\text{heavy}}(\Delta) \) from Eq. (8). By observing the conditions in Eqs. (8) and (9), we know that 

\( \text{work}_i(\Delta) \leq \omega_i^{\text{light}}(\Delta) \leq \omega_i^{\text{heavy}}(\Delta) \).
- Since \( \rho \) is a user-defined parameter, a smaller \( \rho \) implies a larger \( \mu_k \), i.e., potentially more carry-in tasks and more heavy carry-in tasks. By contrast, a larger \( \rho \) implies a smaller \( \mu_k \), i.e., potentially less carry-in tasks and more light carry-in tasks. Therefore, a larger \( \rho \) is better for minimizing the carry-in workload.
- However, the window of interest \([t_0, t_d)\) is defined by the condition \( \Omega(t_0, t_d) \geq M - (M - 1)\rho \). The window of interest is smaller when \( \rho \) is larger. As a result, there is no monotonicity with respect to the schedulability test for setting the value of \( \rho \).

**Theorem 3.10.** Task \( \tau_k \) is schedulable by the given global fixed-priority scheduling if 

\[ \forall \ell \in \mathbb{N}, \exists 1 \geq \rho \geq \ell C_k / ((\ell - 1)T_k + D_k), \forall \Delta \geq (\ell - 1)T_k + D_k \]

\[ \ell C_k + \sum_{\tau_i \in \mathcal{T}^{\text{carry}}} \omega_i^{\text{diff}}(\Delta, \rho) + \sum_{i=1}^{k-1} \text{work}_i(\Delta) \leq \Delta \cdot \mu_k \] 

(14)

holds, where \( \mu_k = M - (M - 1)\rho \),

\[ \omega_i^{\text{diff}}(\Delta, \rho) = \begin{cases} \omega_i^{\text{heavy}}(\Delta) - \text{work}_i(\Delta) & \text{if } U_i > \rho \\ \omega_i^{\text{light}}(\Delta) - \text{work}_i(\Delta) & \text{if } U_i \leq \rho \end{cases} \] 

(15)

and \( \mathcal{T}^{\text{carry}} \) is the set of the \([\mu_k] - 1\) tasks among the \( k - 1 \) higher-priority tasks with the largest values of \( \omega_i^{\text{diff}}(\Delta, \rho) \). If \( D_k \leq T_k \), we only need to consider \( \ell = 1 \).

**Proof.** We prove this theorem by contrapositive, i.e., task \( \tau_k \) misses its deadline first at time \( t_d \) in a global fixed-priority preemptive schedule \( S \). We know that \( t_o \) can be defined for schedule \( S \), and \( t_0 \), i.e., \( \Omega(t_0, t_d) \geq M - (M - 1) \times \frac{C_k}{D_k} \) in Definition 3.3 can be defined for any \( \rho \) with \( 1 \geq \rho \geq \ell C_k / ((\ell - 1)T_k + D_k) \) due to Lemma 3.4.

By the existence of \( t_d \), the choice of \( \rho \), and the definition of \( t_0 \) in Definition 3.3, we know that the deadline miss of task \( \tau_k \) at time \( t_d \) in the schedule \( S \) implies 

\[ \exists \ell \in \mathbb{N}, \forall 1 \geq \rho \geq \ell C_k / ((\ell - 1)T_k + D_k), \exists \Delta = t_d - t_0, \quad \Omega(t_0, t_d) \geq M - (M - 1)\rho \] 

(16)
By the fact that $C_k < C'_k = \ell C_k$ and the definition of $\Omega()$, we have

$$
\Omega(t_0, t_d) = \frac{C_k + E(t_0, t_d)}{t_d - t_0} < \frac{\ell C_k + E(t_0, t_d)}{t_d - t_0}
$$

(17)

By Lemma 3.6, for a specific $\rho$, there are at most $\lceil M - (M - 1)\rho \rceil - 1 = \lfloor \mu_k \rfloor - 1$
higher-priority carry-in tasks at time $t_0$ and the other higher-priority tasks do not have any
unfinished job at time $t_0$. Suppose that $T_{heavy}$ and $T_{light}$ are the sets of the heavy and
light carry-in tasks at time $t_0$, respectively. By Lemma 3.6, $|T_{heavy}| + |T_{light}| \leq \lfloor \mu_k \rfloor - 1$.
Therefore, by using Lemmas 3.7 and 3.8 and 3.9, we have

$$
E(t_0, t_d) \leq \sum_{\tau_i \in T_{heavy}} \mu_i^{heav}(\Delta) + \sum_{\tau_i \in T_{light}} \mu_i^{light}(\Delta)
$$

$$
= \sum_{\tau_i \in T_{heavy}} (\mu_i^{heav}(\Delta) - \text{work}(\Delta)) + \sum_{\tau_i \in T_{light}} (\mu_i^{light}(\Delta) - \text{work}(\Delta)) + \sum_{i=1}^{k-1} \text{work}(\Delta)
$$

(18)

$$
\leq \sum_{\tau_i \in T_{carry}} \mu_i^{diff}(\Delta, \rho) + \sum_{i=1}^{k-1} \text{work}(\Delta)
$$

(19)

where $\mu_i^{diff}(\Delta, \rho)$ is defined in Eq. (15), and $T_{carry}$ is defined in the statement of the
theorem.

By Eqs. (16), (17), and (19), and the fact $t_d - t_a \geq D_k' = (\ell - 1)T_k + D_k$, the deadline
miss of task $\tau_k$ at time $t_d$ implies

$$
\exists \ell \in N, \forall 1 \geq \rho \geq \ell C_k/(\ell - 1)T_k + D_k, \exists \Delta \geq (\ell - 1)T_k + D_k
$$

$$
\ell C_k + \sum_{\tau_i \in T_{carry}} \mu_i^{diff}(\Delta, \rho) + \sum_{i=1}^{k-1} \text{work}(\Delta) > \Delta \cdot \mu_k
$$

(20)

Therefore, the negation of the above necessary condition for the deadline miss of task $\tau_k$
at time $t_d$ is a safe sufficient schedulability test. We reach the conclusion of the schedulability
test.

When $D_k \leq T_k$, since $t_d$ is the earliest moment in the schedule $S$ with a deadline miss
of task $\tau_k$, we know that $t_a$ is by definition $t_d - D_k$ and $\ell$ is 1. Therefore, we only have to
consider $\ell = 1$ when $D_k \leq T_k$.

The schedulability test described in Theorem 3.10 can be informally explained as follows:
1) it requires to test all the possible positive integers for $\ell$, like the busy-window concept, 2)
it has to find a $\rho$ value in the specified range, and 3) for the specified combination of $\ell$ and
$\rho$, we have to test whether the condition in Eq. (14) holds for every $\Delta \geq (\ell - 1)T_k + D_k$.

3.3 Remarks on Implementing Theorem 3.10

Unfortunately, due to the following issues, implementing the schedulability test in Theo-
rem 3.10 directly would lead to a high time complexity:

**Issue 1 due to $\Delta$:** For specific $\ell$ and $\rho$, testing the schedulability condition in Eq. (14)
requires to evaluate all $\Delta \geq (\ell - 1)T_k + D_k$. Suppose that $HP(k)$ is the hyper-period of
$\{\tau_1, \tau_2, \ldots, \tau_k\}$, i.e., the least common multiple of the periods of $\tau_1, \tau_2, \ldots, \tau_k$.
Since $work_i(\Delta) + HP(k)U_i = work_i(\Delta + HP(k))$, $\mu_i^{light}(\Delta) + HP(k)U_i = \mu_i^{light}(\Delta + HP(k))$, and
$\mu_i^{heav}(\Delta) + HP(k)U_i = \mu_i^{heav}(\Delta + HP(k))$, we only have to test $\Delta \in [(\ell - 1)T_k + D_k, (\ell - 1)T_k + D_k + HP(k)]$, as long as $\sum_{i=1}^{k} U_i \leq \mu_k$. However, the time complexity
can still be exponential. We will explain how to reduce this complexity by using safe
upper bounds in Section 4.
Lemma 4.1. When $0 \leq U_i \leq 1$, for any $\Delta \geq 0$,
\[
\text{work}_i(\Delta) \leq C_i - C_i U_i + U_i \Delta.
\] (21)

Proof. This inequality was already stated in Eq. (5) by Bini et al. [14] as a fact. Here, we provide the proof for completeness. Suppose that $\Delta$ is $p_3 T_i + q_3$, where $p_3$ is $\left\lceil \frac{\Delta}{T_i} \right\rceil$ and $q_3$ is
\[
\Delta - \left\lceil \frac{\Delta}{T_i} \right\rceil T_i.
\]
Therefore, we know $U_i \Delta = p_3 C_i + q_3 U_i$ and $\text{work}_i(\Delta) = p_3 C_i + \min\{C_i, q_3\}$. We have to consider two cases:

- If $q_3 \leq C_i$: we have
  \[
  \text{work}_i(\Delta) = p_3 C_i + q_3 \leq p_3 C_i + C_i - (C_i - q_3) \\
  \leq p_3 C_i + C_i - (C_i - q_3) U_i = C_i - C_i U_i + U_i \Delta,
  \]
  where $\leq$ is due to $0 \leq U_i \leq 1$ and $C_i - q_3 \geq 0$.

- If $q_3 > C_i$: we have
  \[
  \text{work}_i(\Delta) = p_3 C_i + C_i \leq p_3 C_i + C_i + (q_3 - C_i) U_i = C_i - C_i U_i + U_i \Delta,
  \]
  where $\leq$ is due to $0 \leq U_i \leq 1$ and $q_3 - C_i > 0$.

Lemma 4.2. For any $\Delta \geq 0$,
\[
\omega^\text{heavy}_i(\Delta) \leq C_i + U_i D_i - C_i U_i + U_i \Delta.
\] (22)

Proof. Due to Lemma 3.7 and Lemma 4.1, the inequality holds.

Lemma 4.3. If $U_i \leq \rho \leq 1$, for any $\Delta \geq 0$,
\[
\omega^\text{light}_i(\Delta) \leq C_i - C_i U_i + U_i \Delta.
\] (23)
**Lemma 3.8.** For specific $\ell, \rho$, and $\omega^\ell$, we have $\omega^\ell = \text{work}(\omega^\ell)$, as defined in Lemma 4.1 directly.

By the definition of $p_2$ and $q_2$, we have $p_2 + 1 = [(\Delta - C_i) / T_i]$ and $p_2 + 1 = \text{work}(p_2 T_i + C_i)$. Therefore, for any $\Delta > C_i$, if $C_i - \rho(T_i - q_2) \geq 0$, we get

$$\omega^\ell = (p_2 + 1)C_i + C_i - \rho(T_i - q_2)$$

$$= \text{work}(\omega^\ell) + C_i - \rho(T_i - q_2)$$

$$\leq C_i - C_i U_i + U_i(\Delta - q_2) + C_i - \rho(T_i - q_2)$$

$$= C_i - C_i U_i + U_i(\Delta - q_2) - T_i(\rho - U_i)$$

$$= C_i - C_i U_i + U_i(\Delta) - (T_i - q_2)(U_i - \rho)$$

$$\leq C_i - C_i U_i + U_i\Delta,$$

where $\leq_1$ is due to Lemma 4.1 and $\leq_2$ is due to $q_2 \leq T_i$ and $U_i \leq \rho$. For any $\Delta > C_i$, if $C_i - \rho(T_i - q_2) < 0$, similarly, we have

$$\omega^\ell = (p_2 + 1)C_i = \text{work}(p_2 T_i + C_i) \leq \text{work}(p_2 T_i + C_i + q_2) = \text{work}(\Delta)$$

Therefore, we reach the conclusion.

With the help of the above lemmas for safe approximations, we can now safely and efficiently handle the schedulability test for specific $\ell$ and $\rho$ in the following theorem. This handles Issue 1 explained at the end of Section 3.

**Theorem 4.4.** Task $\tau_k$ is schedulable by the given global fixed-priority scheduling if

$$\forall \ell \in \mathbb{N}, \exists \tau_i \geq \rho \geq \ell C_k / ((\ell - 1)T_k + D_k)$$

$$\frac{\ell C_k}{D_k} + \sum_{\tau_i \in T_{\text{carry-approx}}} \frac{\gamma_i U_i D_i}{D_k} + \sum_{i=1}^{k-1} \left( C_i - C_i U_i + U_i \right) \leq \mu_k,$$

where $\mu_k = M - (M - 1)\rho$ with $1 \geq \rho \geq \ell C_k / ((\ell - 1)T_k + D_k)$, $D_k'$ is $(\ell - 1)T_k + D_k$, and $T_{\text{carry-approx}}$ is the set of the $[\mu_k] - 1$ tasks among the $k - 1$ higher-priority tasks with the largest values of $\gamma_i U_i D_i$. Note that $|T_{\text{carry-approx}}|$ can be smaller than $[\mu_k] - 1$ if the number of tasks with $U_i > \rho$ is less than $[\mu_k] - 1$. If $D_k \leq T_k$, we only need to consider $\ell = 1$.

**Proof.** We prove that the condition in this theorem is a safe upper bound of that in Theorem 3.10. For specific $\ell, \rho, \Delta$, we can find $T_{\text{carry}}$ as defined in Theorem 3.10. By Lemmas 4.1, 4.2, and 4.3 and the assumptions $\Delta \geq (\ell - 1)T_k + D_k = D_k'$ and $0 < U_i \leq 1 \forall \tau_i$, we
have
\[ \ell C_k + \sum_{\tau_i \in T^\text{carry}} \omega_i^{\text{diff}}(\Delta, \rho) + \sum_{i=1}^{k-1} \text{work}_i(\Delta) \]
\[ \leq \ell C_k + \sum_{\tau_i \in T^\text{carry}} \gamma_i U_i D_i + \sum_{i=1}^{k-1} (C_i - C_i U_i + U_i \Delta) \]
\[ \leq \ell C_k + \sum_{\tau_i \in T^\text{carry-approx}} \gamma_i U_i D_i + \sum_{i=1}^{k-1} (C_i - C_i U_i + U_i \Delta) \]
\[ \leq \Delta \cdot \left( \frac{\ell C_k}{D_k} + \sum_{\tau_i \in T^\text{carry-approx}} \frac{\gamma_i U_i D_i}{D_k'} + \sum_{i=1}^{k-1} \left( \frac{C_i - C_i U_i}{D_k'} + U_i \right) \right) \]
\[ \forall \ell \in \mathbb{N}, \exists \exists \rho \geq \ell C_k/((\ell - 1)T_k + D_k)T_k + D_k \]
\[ \frac{\ell C_k}{D_k'} + \left( \sum_{\tau_i \in T^\text{carry-approx}} \frac{\gamma_i U_i D_i}{D_k'} \right) + \sum_{i=1}^{k-1} \left( \frac{C_i - C_i U_i}{D_k'} + U_i \right) \leq \mu_k \]

Therefore, the test in Theorem 3.10 can be safely over-approximated as follows:

Theorem 4.4 provides two interesting implications to handle Issue 2. Firstly, if \( U_i \leq \rho \), the linear approximation of \( \text{work}_i(\Delta) \) by considering task \( \tau_i \) as a non-carry-in task in Lemma 4.1 is the same as the linear approximation of \( \omega_i^{\text{light}}(\Delta) \) by considering task \( \tau_i \) as a carry-in task in Lemma 4.3. Therefore, the carry-in tasks are only effective for those tasks \( \tau_i \) with \( U_i > \rho \). Secondly, for a specific \( \ell \), deciding whether a specific \( \rho \) exists to pass the test in Eq. (24) can be done by only testing a finite number of \( \rho \) values, i.e., by starting from \( \rho = \ell C_k/((\ell - 1)T_k + D_k) \) and increasing \( \rho \) to the next (higher) values where \( T^\text{carry-approx} \) changes. This means either 1) \( \rho = U_i \) for certain higher-priority task \( \tau_i \), i.e., the summation can be larger with the same number of summands; or 2) \( \mu_k = M - (M - 1)\rho \) is an integer, i.e., the number of summands increases. This only has time complexity \( O((k + M) \log(k + M)) \), mainly due to the sorting, when proper data structures are used. Details can be found in Appendix of the full version [21].

### 4.1 Linear-Time Schedulability Tests

The time complexity of Theorem 4.4 is due to the search of possible \( \rho \) values. Nevertheless, we can directly set \( \rho \) to \( U^{\text{max}}_{\delta, k} \) which implies that there is no carry-in task in the linear-approximation form. With this simplification, we can conclude different schedulability tests in Theorems 4.5, 4.6, and 4.7. Although these tests are not superior to Theorem 4.4, our main target is the test in Theorem 4.7, which will be used mainly to derive the speedup bounds later in Theorem 5.2.

**Theorem 4.5.** Task \( \tau_k \) is schedulable by the given global fixed-priority scheduling if \( \forall \ell \in \mathbb{N} \)

\[ \frac{\ell C_k}{D_k'} + \sum_{i=1}^{k-1} \left( \frac{C_i - C_i U_i}{D_k'} + U_i \right) \leq (M - (M - 1)U^{\text{max}}_{\delta, k}) \]

holds, where \( D_k' = (\ell - 1)T_k + D_k \).
Proof. This comes directly from Theorem 4.4 by setting $\rho$ to $U_{\delta,k}^{\text{max}}$ and the facts that $U_{\delta,k}^{\text{max}} \geq U_i$ for $i = 1, 2, \ldots, k - 1$ and $U_{\delta,k}^{\text{max}} \geq \delta_k \geq \ell C_k/((\ell - 1)T_k + D_k)$ by definition.

► Theorem 4.6. Suppose that $D_k > T_k$. Let $b = \frac{D_k - T_k}{\delta_k}$. Task $\tau_k$ is schedulable by the given global fixed-priority scheduling algorithm if:

$$\sum_{i=1}^{k} U_i \leq (M - (M - 1)U_{\delta,k}^{\text{max}}), \quad \text{when } b U_k - \sum_{i=1}^{k-1} \frac{C_i - C_i U_i}{T_k} > 0 \quad (31)$$

$$C_k + \sum_{i=1}^{k-1} \left( \frac{C_i - C_i U_i}{D_k} + U_i \right) \leq (M - (M - 1)U_{\delta,k}^{\text{max}}), \quad \text{otherwise} \quad (32)$$

Proof. For a given $\ell$, the left-hand side in Eq. (30) can be rephrased as:

$$F(\ell) = \frac{\ell C_k}{D_k} + \sum_{i=1}^{k-1} \left( \frac{C_i - C_i U_i}{D_k} + U_i \right) = \frac{\ell U_k}{\ell + b} + \sum_{i=1}^{k-1} \frac{C_i - C_i U_i}{T_k} + \sum_{i=1}^{k-1} U_i \quad (33)$$

The first order derivative of $F(\ell)$ with respect to $\ell$ is:

$$\frac{\partial F(\ell)}{\partial \ell} = \frac{b U_k - \sum_{i=1}^{k-1} \frac{C_i - C_i U_i}{T_k}}{(\ell + b)^2}. \quad (34)$$

We have to consider two cases:

- Case 1: if $b U_k - \sum_{i=1}^{k-1} \frac{C_i - C_i U_i}{T_k} > 0$, then $F(\ell)$ is an increasing function with respect to $\ell$. Therefore, $F(\ell)$ is maximized when $\ell \to \infty$, i.e., $F(\ell) \leq \sum_{i=1}^{k} U_i$.

- Case 2: if $b U_k - \sum_{i=1}^{k-1} \frac{C_i - C_i U_i}{T_k} \leq 0$, then $F(\ell)$ is a non-increasing function with respect to $\ell$. Therefore, $F(\ell)$ is maximized when $\ell \to 1$, i.e., $F(\ell) \leq \frac{C_k}{D_k} + \sum_{i=1}^{k-1} \left( \frac{C_i - C_i U_i}{D_k} + U_i \right)$.

► Theorem 4.7. Task $\tau_k$ is schedulable by the given global fixed-priority scheduling if

$$\delta_k + \sum_{i=1}^{k-1} \left( \frac{C_i - C_i U_i}{D_k} + U_i \right) \leq M - (M - 1)U_{\delta,k}^{\text{max}} \quad (35)$$

Proof. Based on Theorem 4.5 and the two facts that $D_k = (\ell - 1)T_k + D_k \geq D_k$ and $\delta_k \geq \ell C_k/((\ell - 1)T_k + D_k)$ for all $\ell \in \mathbb{N}$, we reach the conclusion.

4.2 Dominance

We now show analytical dominance among the tests presented above and in Theorem 3.10 in the following corollary. A test $B_1$ analytically dominates another test $B_2$ if the schedulability condition in $B_1$ always dominates that in $B_2$. This means, if task $\tau_k$ is deemed schedulable by $B_2$, task $\tau_k$ is also deemed schedulable by $B_1$.

► Corollary 4.8. For arbitrary-deadline sporadic real-time systems under global fixed-priority scheduling, the schedulability tests have the following dominance relations:

- Theorem 3.10 analytically dominates Theorem 4.4.
- Theorem 4.4 analytically dominates Theorem 4.5.
- Theorem 4.5 is equivalent to the test in Theorem 4.6.
- Theorem 4.6 analytically dominates Theorem 4.7.
Proof. They follow directly from the above analyses. The reason why Theorems 4.5 and 4.6 are equivalent is because the conditions in Theorem 4.6 represent exactly the worst-case selection in Theorem 4.6. The other cases are obvious. ▶

Although we will show in Theorem 5.3 that all the above schedulability tests have the same speedup bound for global DM, the performance of the schedulability tests in this section can be very different in practice. Chen et al. [19] have recently shown that “Speedup factors... often lack the power to discriminate between the performance of different scheduling algorithms and schedulability tests even though the performance of these algorithms and tests may be very different when viewed from the perspective of empirical evaluation.” To avoid concluding an algorithm with a reasonable speedup bound but practically not useful, we performed a series of experiments and present the results in Section 6. Moreover, according to the experimental results, the domination relations among Theorems 4.4, 4.5, and 4.7 are strict, i.e., there is a concrete input instance that is deemed schedulable by a dominating schedulability test but is not deemed schedulable by a dominated schedulability test.

5 Global Deadline-Monotonic (DM) Scheduling

After presenting the schedulability tests for any global fixed-priority scheduling algorithms, we focus ourselves on global DM in this section. We will discuss the speedup upper bound and the speedup lower bound. Baruah and Fisher [8] showed that global DM has a speedup upper bound of $2 + \sqrt{3} \approx 3.73$ compared to the optimal schedules, based on the test restated in Theorem 2.4. This is the best known upper bound on speedup factors for arbitrary-deadline sporadic task systems under global fixed-priority scheduling. Evaluating $\text{load}(k)$ in Theorem 2.4 requires to calculate $\sum_{i=1}^{k} \text{DBF}(\tau_i, t) / t$ at all time points $t$. This means, the naïve implementation has an exponential-time complexity. There are more efficient methods, as discussed by Baruah and Bini [6], but the time complexity remains exponential. Although it is possible to approximate $\text{load}(k)$ by using approximate demand bound functions in polynomial time, this is at a price of higher $\text{load}(k)$. We show that the test in Theorem 2.4 is over-pessimistic and is analytically dominated by our linear-time schedulability test in Theorem 4.7 under global DM.

Corollary 5.1. For global DM, the schedulability test in Theorem 4.7 analytically dominates the schedulability test in Theorem 2.4 proposed by Baruah and Fisher [8].

Proof. This is due to the following facts:

By definition, $\text{load}(k) \geq \lim_{t \to \infty} \sum_{i=1}^{k} \text{DBF}(\tau_i, t) / t = \sum_{i=1}^{k} U_i$.

Since $D_i \leq D_k$ in global DM for $i = 1, 2, \ldots, k - 1$, we know that $\sum_{i=1}^{k} \text{DBF}(\tau_i, D_k) / D_k \geq \sum_{i=1}^{k} C_i / D_k$. Therefore, $\text{load}(k) \geq \text{load}(k) \geq \sum_{i=1}^{k} C_i / D_k$.

Combining these facts, we get

$$\delta_k + \sum_{i=1}^{k-1} \left( \frac{C_i - C_i U_i}{D_k} + U_i \right) \leq \sum_{i=1}^{k} C_i / D_k + U_i \leq 2 \text{load}(k).$$

Since we know that the right-hand side in Eq. (4), i.e., $M - (M-1)\delta_{\text{max}}(k)$, is less than or equal to $M - (M-1)U_{\text{max}}^{\text{max}}$ in Eq. (35), we reach the conclusion. ▶

Theorem 5.2. Global DM has a speedup bound of $3 - \frac{2}{M}$, with respect to the optimal schedule, when $M \geq 2$. 

Proof. We only prove the speedup bound by using the schedulability test in Theorem 4.7.

Due to the dominance properties in Corollary 4.8, such a bound also holds for the schedulability tests from Theorems 3.10, 4.4, 4.5, and 4.6.

Suppose that task \( \tau_k \) is not schedulable by global DM. Since \( D_i \leq D_k \) for any \( i = 1, 2, \ldots, k-1 \) under global DM, we know \( \text{DBF}(\tau_i, D_k) \geq C_i \). Therefore, under global DM,\n
\[
\sum_{i=1}^{k} \frac{C_i}{MD_k} \leq \sum_{i=1}^{k} \frac{\text{DBF}(\tau_i, D_k)}{MD_k} \leq \max_{t > 0} \frac{\sum_{\tau_i \in T} \text{DBF}(\tau_i, t)}{MD_k}.
\]

By the assumption that task \( \tau_k \) is also deemed not schedulable by Theorem 4.7, we have

\[
\delta_k + \sum_{i=1}^{k-1} \left( \frac{C_i - C_i U_i}{D_k} + U_i \right) > M - (M - 1) U_{\delta, k}^{\text{max}}
\]

\[
\Rightarrow \sum_{i=1}^{k} \frac{C_i}{MD_k} + \sum_{i=1}^{k} \frac{U_i}{M} > 1 - \left(1 - \frac{1}{M}\right) U_{\delta, k}^{\text{max}}
\]

\[
\Rightarrow \sum_{i=1}^{k} \frac{C_i}{MD_k} + \sum_{i=1}^{k} \frac{U_i}{M} + \left(1 - \frac{1}{M}\right) U_{\delta, k}^{\text{max}} > 1
\]

\[
x + x + (1 - 1/M)x > 1 \Rightarrow \max_{t > 0} \left\{ \sum_{i=1}^{k} \frac{C_i}{MD_k}, \sum_{i=1}^{k} \frac{U_i}{M} \right\} > \frac{1}{3 - 1/M} \tag{37}
\]

Therefore, either \( \max_{t > 0} \sum_{\tau_i \in T} \frac{\text{DBF}(\tau_i, t)}{MD_k} > \frac{1}{3 - 1/M} \), or \( \sum_{i=1}^{k} \frac{U_i}{M} > \frac{1}{3 - 1/M} \). By Lemma 2.3, we reach the conclusion of the speedup bound for global DM with respect to the optimal schedule. \(\blacktriangleright\)

Theorem 5.3. For global DM, the schedulability tests in Theorems 3.10, 4.4, 4.5, 4.6, and 4.7 have a speedup bound of \( 3 - \frac{1}{M} \), with respect to the optimal schedule, when \( M \geq 2 \).

Proof. This is due to Theorem 5.2 and Corollary 4.8, because all of the tests in Theorems 3.10, 4.4, 4.5, 4.6 dominate the test in Theorem 4.7 as presented in Corollary 4.8. \(\blacktriangleright\)

Theorem 5.4. The speedup bound of global DM for arbitrary-deadline task systems is at least \( 3 - \frac{3}{M+1} \).

Proof. The proof is based on a concrete task set. We specifically use the following task set \( T^{ad} \) with \( N = 2M + 1 \) tasks. Let \( \varepsilon \) be an arbitrarily small positive real number such that \( 1/\varepsilon \) is an integer. Let \( \eta \ll \varepsilon \) be an arbitrarily small positive number, that is used to enforce the priority assignment under global DM:

\[
\begin{align*}
C_i &= \frac{\kappa}{3}, T_i = \varepsilon, D_i = 1, \text{ for } i = 1, 2, \ldots, M. \\
C_i &= \frac{1}{\eta}, T_i = \infty, D_i = 1 + \eta, \text{ for } i = M + 1, M + 2, \ldots, 2M. \\
C_i &= \frac{1}{3\eta}, T_i = \infty, D_i = 1 + 2\eta, \text{ for } i = 2M + 1
\end{align*}
\]

As the setting of \( \eta \ll \varepsilon \) is just to enforce the indexing, we will directly take \( \eta \to 0 \) here. In the Appendix, we prove two properties: 1) \( T^{ad} \) is not schedulable by global DM under a concrete instance which releases all the tasks at time 0 and the subsequent jobs periodically. 2) There exists a feasible schedule for task set \( T^{ad} \) at any speed no lower than \( \frac{1+\varepsilon}{1+\varepsilon} + \frac{1+\varepsilon}{3M} \) under a concrete semi-partitioned multiprocessor schedule, i.e., \( \{ \tau_m, \tau_{m+1} \} \) assigned to processor \( m \) for \( m = 1, 2, \ldots, M \) and task \( \tau_{2M+1} \) executed partially on each of the \( M \) processors. Therefore, a lower bound on the speedup bound of global DM is:

\[
\lim_{\varepsilon \to 0} \frac{1}{\frac{1+\varepsilon}{3M} + \frac{1+\varepsilon}{3M}} = \lim_{\varepsilon \to 0} \frac{3M}{(1+\varepsilon) \times (M+1)} = \frac{3M}{M+1} = 3 - \frac{3}{M+1}.
\]

\(\blacktriangleright\)
By Theorems 5.2 and 5.4, we can reach the conclusion that all the schedulability tests from Theorems 3.10, 4.4, 4.5, 4.6, and 4.7 are asymptotically tight with respect to speedup bounds. However, these tests have different performance with respect to the schedulability.

6 Evaluation

We evaluated the scheduling tests provided in this paper by comparing their acceptance ratio to the acceptance ratio of other algorithms, i.e., comparing the percentage of task sets accepted for the different schedulability tests, using different settings for the number of processors, the scheduling policy, and the ratio of the relative deadline to the period.

Evaluation Setup: We conducted evaluations for homogeneous multiprocessor systems with $M = 4$, $M = 8$, and $M = 16$ processors. We generated 100 task sets with cardinality of both $N = 5 \times M$ and $N = 10 \times M$, and utilization ranging from $M \times 5\%$ to $M \times 100\%$ in steps of $M \times 5\%$. The UUniFast-Discard method [13] was adopted to generate the utilization values of a set of $N$ tasks under the target utilization. As suggested by Emberson et al. [28], the periods were generated according to a log-uniform distribution, with 1, 2, and 3 orders of magnitude, i.e., $[1 \text{ms} - 10 \text{ms}]$, $[1 \text{ms} - 100 \text{ms}]$, and $[1 \text{ms} - 1000 \text{ms}]$. For each task, the relative deadline was set to the period multiplied with a value randomly drawn under a uniform distribution from a given interval $I$. We conducted evaluations using different intervals, i.e., $I = [0.8, 2]$, $[0.8, 5]$, $[0.8, 10]$, $[1, 2]$, $[1, 5]$, or $[1, 10]$. To schedule the task sets, we applied global deadline-monotonic (DM) and global slack-monotonic (SM) [1] scheduling.

Whether the task set is schedulable under the given scheduling approach or not was tested using the following schedulability tests:

- **LOAD**: The load-based analysis by Baruah and Fisher in [9], only for DM scheduling.
- **BAK**: The test by Baker in Theorem 11 in [3].
- **HC**: The sufficient test in Corollary 2 by Huang and Chen in [31].
- **OUR-4.4**: The sufficient test in Theorem 4.4 in this paper.
- **OUR-4.6**: The sufficient test in Theorem 4.6 in this paper.
- **OUR-4.7**: The sufficient test in Theorem 4.7 in this paper.

We also checked if a task set was schedulable according to at least one of the tests, denoted as **ALL**. We only present a small set of the conducted tests here. The diagrams of all conducted evaluations can be found in [20].

Evaluation Results: Figure 3 shows the evaluations under the setting used in the paper by Huang and Chen [31]. They used DM scheduling on $M = 8$ processors, a task set containing 40 tasks and ratios of $\frac{D_i}{T_i} \in [0.8, 2]$ and analyzed the schedulability for $T_i$ values that differ up to 1, 2, and 3 orders of magnitude, i.e., $T_i$ in a range of $[1 \text{ms}, 10 \text{ms}]$, $[1 \text{ms}, 100 \text{ms}]$, or $[1 \text{ms}, 1000 \text{ms}]$. The test by Baruah and Fisher [9] is clearly outperformed by Theorem 4.6, Theorem 4.7, and Baker’s test [3], which provide similar acceptance ratios. The test by Huang and Chen [31] outperforms those three tests and is worse than the test in Theorem 4.4 in these settings. However, there is no dominance relation between Theorem 4.4 and the test by Huang and Chen [31], as some task sets are schedulable under the test by Huang and Chen [31] but not schedulable under Theorem 4.4 and vice versa.

There are other configurations where the test by Huang and Chen [31] performs better than Theorem 4.4. One example is shown in Figure 4, analyzing the impact of the number of processors. Here Theorem 4.4 performs compatible to Theorem 4.6, Theorem 4.7, and Baker’s test [3] for $M = 4$. When the number of processors increases, Theorem 4.4 performs better. The gap to Huang and Chen [31] is smaller for 8 processors and Theorem 4.4 has a higher acceptance rate when the utilization level is $80\% \times M$. For $M = 16$ processors
Theorem 4.4 accepts more task sets than Huang and Chen [31] when the utilization level is \( \geq 65\% \times M \). In addition, it is possible that the number of task sets that is accepted by at least one algorithm is not close to the number of task sets accepted by Huang and Chen [31] or Theorem 4.4 as can be seen for the utilization level \( 75\% \times M \) in the case where \( M = 8 \). Furthermore, we tracked if the test by Baker [3] accepted some task sets that were not accepted by Huang and Chen [31] or Theorem 4.4, which happened occasionally. Therefore, we conclude that there is no dominance relation between any of those three tests, i.e., Theorem 4.4, and the tests by Baker [3] and by Huang and Chen [31]. As these tests can all be implemented with polynomial-time complexity, all three should be applied.

7 Conclusion

We present a series of schedulability tests for multiprocessor systems under any given fixed-priority scheduling approach. Those schedulability tests have different tradeoffs between their accuracy and their time complexity. All those schedulability tests dominate the approach by Baruah and Fisher [9], both with respect to speedup bounds and schedulability analysis. Theorem 3.10 is the most powerful schedulability test in this paper. However, we do not reach any concrete implementation with affordable time complexity. In the future work, we will seek for efficient methods to implement the schedulability test in Theorem 3.10.
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Appendix: Additional Proofs

8.1 Proof of Theorem 5.4: Speedup lower bound of global DM for arbitrary-deadline task systems

We will specifically use the following task set $T^{ad}$ with $N = 2M + 1$ tasks. Let $\varepsilon$ be an arbitrarily small positive real number such that $1/\varepsilon$ is an integer. Let $\eta \ll \varepsilon$ be an arbitrarily small positive number, that is used to enforce the priority assignment under global DM:

$$C_i = \frac{\varepsilon}{3}, \quad T_i = \varepsilon, \quad D_i = 1, \text{ for } i = 1, 2, \ldots, M.$$  
$$C_i = \frac{1}{3}, \quad T_i = \infty, \quad D_i = 1 + \eta, \text{ for } i = M + 1, M + 2, \ldots, 2M.$$  
$$C_i = \frac{1 + \varepsilon}{3}, \quad T_i = \infty, \quad D_i = 1 + 2\eta, \text{ for } i = 2M + 1$$

As the setting of $\eta \ll \varepsilon$ is just to enforce the indexing, we will directly take $\eta \rightarrow 0$ here.

\begin{itemize}
  \item Lemma 8.1. $T^{ad}$ is not schedulable by global DM.
  \item Proof. This can be proved by showing that task $\tau_N$ misses its deadline in the following concrete arrival pattern: all tasks release their first jobs at time 0 and the subsequent jobs arrive as early as possible while respecting their minimum inter-arrival times. For this arrival pattern, the jobs of tasks $\tau_1, \tau_2, \ldots, \tau_M$ are executed from time $i\varepsilon$ to time $i\varepsilon + \frac{\varepsilon}{3}$ for $i = 0, 1, 2, \ldots, 1/\varepsilon$. Therefore, these $M$ tasks are executed for in total $1/3$ time units from time 0 to time 1. For tasks $\tau_{M+1}, \tau_{M+2}, \ldots, \tau_{2M}$, each of them is executed for $1/3$ time units from time 0 to time 1 when the processors do not execute $\tau_1, \tau_2, \ldots, \tau_M$. Task $\tau_{2M+1}$ is executed alone without any overlap with the executions of the higher-priority tasks. Therefore task $\tau_{2M+1}$ misses its deadline since it needs $1 + \varepsilon$ time units, but only $1/3$ time units are available before its deadline.  
  \item Lemma 8.2. There exists a feasible schedule for task set $T^{ad}$ at any speed no lower than $\frac{1 + \varepsilon}{3} + \frac{1 + \varepsilon}{3M}$. 
\end{itemize}
Proof. We apply multiprocessor semi-partitioned scheduling, in which tasks in \( \{\tau_m, \tau_{m+M}\} \) are assigned to processor \( m \) for \( m = 1, 2, \ldots, M \). In our designed semi-partitioned schedule, a job of task \( \tau_{2M+1} \), i.e., a part of \( \tau_N \), is executed partially on each of the \( M \) processors as follows: it runs on processor \( m \) for \( C_N/M \) amount of time, and then migrates to processor \( m+1 \) to continue its execution, for \( m = 1, 2, \ldots, M-1 \). To ensure that the migration can be served immediately, \( \tau_N \) is given the the highest-priority in this schedule. Therefore, a subtask of task \( \tau_N \) on processor \( m \), denoted as \( \tau_{N,m} \), has a relative deadline \( C_N/M \). As long as the speed of the processors is greater than or equal to \( 1+\varepsilon \), task \( \tau_N \) can meet its deadline.

Therefore, in our designed semi-partitioned schedule, each processor \( m \) has a task set \( T_m \) that consists of three tasks: \( \tau_m \) and \( \tau_{m+M} \) from \( T_{ad} \) and a subtask \( \tau_{N,m} \) of task \( \tau_N \) with execution time \( C_N/M \). We assign the second priority to task \( \tau_{m+M} \) and the lowest priority to task \( \tau_m \) on processor \( m \).

We utilize the worst-case response time analysis by Bini et al. [14]. They showed that if

\[
1 - \sum_{\tau_i \in hp(\tau_k,m)} U_i \leq 1,
\]

then the worst-case response time of a task \( \tau_k \) in a task set \( T_m \) under fixed-priority scheduling on a processor is at most

\[
\frac{C_k + \sum_{\tau_i \in hp(\tau_k,m)} C_i - \sum_{\tau_i \in hp(\tau_k,m)} U_i C_i}{1 - \sum_{\tau_i \in hp(\tau_k,m)} U_i},
\]

(38)

where \( hp(\tau_k,m) \) is the set of the tasks in \( T_m \) that have a higher priorities than task \( \tau_k \). Note that the precondition \( 1 - \sum_{\tau_i \in hp(\tau_k,m)} U_i \leq 1 \) for the test in Eq. (38) to be applicable always holds at any arbitrarily speed since we assign \( \tau_m \) as the lower-priority task on processor \( m \) and \( U_{m+M} \to 0 \), and \( U_{N,m} = C_{N,m}/T_N \to 0 \).

By Eq. (38), if the speed of processor \( m \) is greater than or equal to \( \frac{C_N}{M} + C_{m+M} = \frac{1+\varepsilon}{3M} + \frac{1}{3} \), task \( \tau_{m+M} \) can still meet its deadline in this schedule. By Eq. (38), task \( \tau_m \) can meet its deadline at speed \( s \) in this schedule if

\[
1 \geq \frac{C_m/s + \sum_{\tau_i \in hp(\tau_k,m)} C_i/s - \sum_{\tau_i \in hp(\tau_k,m)} U_i C_i/s}{1 - \sum_{\tau_i \in hp(\tau_k,m)} U_i/s} = \frac{\varepsilon}{3s} + \frac{1}{3s} + \frac{1 + \varepsilon}{3sM},
\]

(39)

Therefore, as long as \( s \geq \frac{1}{3} + \frac{1+\varepsilon}{3sM} \), task \( \tau_m \) meets its deadline under our designed schedule.