

# Approximating Response Times of Static-Priority Tasks with Release Jitters

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## Abstract

We consider static-priority tasks with constrained-deadlines that are subjected to release jitters. We define an approximate worst-case response time analysis and we propose a polynomial time algorithm. For that purpose, we extend the Fully Polynomial Time Approximation Scheme (FPTAS) presented in [2] to take into account release jitters; this feasibility test is then used to define a polynomial time algorithm that computes approximate worst-case response times of tasks. Nevertheless, the approximate worst-case response time values have not been proved to have any bounded error in comparison with worst-case response times.

## 1 Introduction

Guaranteeing that tasks will always meet their deadlines is a major issue in the design of hard-real time systems. A real-time system is said *feasible* if no deadline miss can occur at run-time. We next consider periodic tasks scheduled by a preemptive static-priority scheduler upon a uniprocessor platform. We consider tasks having constrained-deadlines (i.e., deadlines are less than or equal to task periods) and subjected to release jitters. Such a task model allows to analyze hard real-time distributed systems [11].

The feasibility problem consists on proving that tasks will always meet their deadlines at run-time. For the considered real-time systems, the feasibility problem is not known to be NP-hard, but only pseudo-polynomial time tests are known [4, 6, 8]. Sufficient feasibility conditions are known and can be checked in polynomial time. But, when such a test returns "not feasible", this can be a rather pessimistic decision. Recently, approximate feasibility algorithms have been designed to reduce the gap between both approaches. According to an accuracy parameter  $\epsilon$ , they check, in polynomial time, if a task set is:

- feasible (upon a unit speed processor).
- infeasible upon a  $(1 - \epsilon)$ -speed processor. That is, "we must effectively ignore  $\epsilon$  of the processor capacity for the test to become exact" [2]. So, the pessimism in-

roduced by the feasibility test is kept bounded by a constant.

As far as we know, no approximation algorithm is known for approximating worst-case response times of tasks with a constant performance guarantee (i.e., upper bounds of worst-case response times). The aim of this paper is to introduce such an analysis and to try to show its relationship with approximate feasibility analysis. According to an accuracy parameter  $\epsilon$ , we define approximate worst-case response times as follow:

**Definition 1** Let  $\epsilon$  be a constant and  $R_i^*$  be the worst-case response time of a task  $\tau_i$ , then the approximate worst-case response time  $\widehat{R}_i^*$  satisfies:  $R_i^* \leq \widehat{R}_i^* \leq (1 + \epsilon)R_i^*$ .

We first define a preliminary result for computing worst-case response time while performing a processor demand analysis (e.g., [6]), then we extend the FPTAS presented in [2] with release jitters. These results are then combined to define for computing approximate worst-case response times. Nevertheless, the computed approximate worst-case response time values are not guaranteed to be closed to worst-case response times (i.e., with a bounded error).

## 2 Task model and exact analysis

### 2.1 Task model

A task  $\tau_i$ ,  $1 \leq i \leq n$ , is defined by a worst-case execution requirement  $C_i$ , a relative deadline  $D_i$  and a period between two successive releases  $T_i$ . Every task occurrence is called a job. We assume that deadlines are constrained:  $D_i \leq T_i$ . Such an assumption is realistic in many real-world applications and also leads to simpler algorithms for checking feasibility of task sets [5].

In order to model delay due to input data communications of tasks, we also consider that jobs are subjected to release jitters. A release jitter  $J_i$  of a task  $\tau_i$  is a interval of time after the release of a job in which it waits for its input data. When release jitters are considered in the task model, then dependencies among distributed tasks are modeled using the parameters  $J_i$ ,  $1 \leq i \leq n$ . Using such a model, a distributed system can be analyzed processor by processor,

separately using for instance an holistic based schedulability analysis [11].

For a given processor, we assume that all tasks are independent and synchronously released. All tasks have static priorities that are set before starting the application and are never changed at run-time. At any time, the highest priority task is selected for execution among ready tasks. Without loss of generality, we assume next that tasks are indexed according to priorities:  $\tau_1$  is the highest priority task and  $\tau_n$  is the lowest priority one.

## 2.2 Known results

### 2.2.1 Request Bound and Workload Functions

The request bound function of a task  $\tau_i$  at time  $t$  (denoted  $\text{RBF}(\tau_i, t)$ ) and the cumulative processor demand (denoted  $W_i(t)$ ) of tasks at time  $t$  of tasks having priorities greater than or equal to  $i$  are respectively (see [11] for details):

$$\text{RBF}(\tau_i, t) \stackrel{\text{def}}{=} \left\lfloor \frac{t + J_i}{T_i} \right\rfloor C_i \quad (1)$$

$$W_i(t) \stackrel{\text{def}}{=} C_i + \sum_{j=1}^{i-1} \text{RBF}(\tau_j, t) \quad (2)$$

Notice that deadline failures of  $\tau_i$  (if any) occur necessarily in an interval of time where only tasks with a priority higher of equal to  $i$  are running. Such an interval of time is defined as a level- $i$  busy period [6]. Using these functions, two distinct (but linked) exact feasibility tests can be defined. We recall both results that will be reused in the remainder.

### 2.2.2 Processor Demand Analysis

The processor demand approach checks that the processor capacity is always less than or equal to the processor capacity required by task executions. In [6] is presented a processor demand schedulability test for constrained-deadline systems (but the test was extended for arbitrary deadline systems in [5]). It can be also easily extended to tasks subjected to release jitters as stated in the following result:

**Theorem 1** [6] *A static-priority system with release jitters is feasible iff  $\max_{i=1..n} \left\{ \min_{t \in S_i} \frac{W_i(t)}{t} \right\} \leq 1$ , where  $S_i$  is the set of scheduling points defined as follows:  $S_i \stackrel{\text{def}}{=} \{aT_j - J_j \mid j = 1..i, a = 1.. \lfloor \frac{D_i + J_i}{T_j} \rfloor\} \cup \{D_i\}$ .*

Note that schedulability points correspond to a set of time instants in the schedule where a task can start its execution, after the delay introduced by its release jitter.

### 2.2.3 Response Time Analysis

An alternative approach to check the feasibility of a static-priority task set is to compute the worst-case response time

$R_i^*$ . The worst-case response time of  $\tau_i$  is formally defined as:

**Definition 2** *The worst-case response time of a task  $\tau_i$  is:  $R_i^* \stackrel{\text{def}}{=} (\min\{t \in (0, D_i] \mid W_i(t) = t\}) + J_i$ .*

An exact algorithm is known [4] (e.g., for a recursive definition of the following method). Using successive approximations starting from a lower bound of  $R_i^*$ , we can compute to the smallest fixed-point of  $W_i(t) = t$  with the following iterative process:  $W_i^{(0)} = \sum_{j=1}^i C_j$ ,  $W_i^{(k+1)} = C_i + \sum_{j=1}^{i-1} \text{RBF}(\tau_j, W_i^{(k)})$ . Computations stop for the smallest integer  $k$  such that:  $W_i^{(k+1)} = W_i^{(k)}$ .

These approaches are all based on the analysis of the cumulative processor demand [9]. But, as far as we know, no direct link has been presented between these approaches. The initial value (e.g.,  $W_i^{(0)}$ ) plays an important role to limit the number of required iterations to reach the smallest fixed point of equation  $W_i(t) = t$ . Different approaches have been proposed in [10, 1] and are quite useful in practice to reduce computation time. Nevertheless, such improvements are not necessary to present our results and for that reason are not developed in the remainder.

## 2.3 A preliminary result

We show that worst-case response times of tasks can be easily computed using a Time Demand Analysis (i.e., Theorem 1), for every feasible task set (and only for them). For a feasible task  $\tau_i$ , it is sufficient to check the following testing set [6]:  $S_i = \{aT_j - J_j \mid j = 1..i, a = 1.. \lfloor \frac{D_i + J_i}{T_j} \rfloor\} \cup \{D_i\}$ .

We first define the critical scheduling point that allows to compute the worst-case response time of  $\tau_i$  (under the assumption that the task  $\tau_i$  will meet its deadline at execution time).

**Definition 3** *The critical scheduling point for a feasible task  $\tau_i$  is:  $t^* \stackrel{\text{def}}{=} \min\{t \in S_i \mid W_i(t) \leq t\}$ .*

We now prove if  $t^*$  exists, then  $W_i(t^*) + J_i$  defines the worst-case response time of  $\tau_i$ .

**Theorem 2** *The worst-case response time of a task  $\tau_i$ , such that  $W_i(t^*) \leq t^*$  is exactly  $R_i^* = W_i(t^*) + J_i$ .*

*Proof:* Since we assume that  $W_i(t^*) \leq t^*$ , then  $\tau_i$  is feasible. Let  $S_i = \{t_{i1}, t_{i2}, \dots, t_{i\ell}\}$  with  $t_{i1} < t_{i2} < \dots < t_{i\ell} < \dots < t_{i\ell} = D_i$ . By Definition 3, there exists  $t^* = t_{ij}$ , where  $1 \leq j \leq \ell$ , is the first scheduling point verifying  $W_i(t^*) \leq t^*$ :  $W_i(t) > t$  for all  $t \in \{t_{i1}, \dots, t_{ij-1}\}$  and  $W_i(t_{ij}) \leq t_{ij}$ .

Since  $W_i(t)$  is non-decreasing between subsequent scheduling points  $\{t_{ia}, t_{ia+1}\}$ ,  $1 \leq a \leq \ell - 1$ , then there exists a time  $t \in (t_{ij-1}, t_{ij})$  such that  $W_i(t) = t$ . Since scheduling points in  $S_i$  corresponds to task releases, then

no new task is released between  $t$  and  $t^*$  and as a consequence we have  $W_i(t) = W_i(t^*)$ . The worst-case response time of  $\tau_i$  is then defined as  $W_i(t^*) + J_i$ . ■

Thus, for all feasible tasks, it is quite easy to compute their worst-case response times. But, for an infeasible task  $\tau_i$  (e.g.,  $R_i^* > D_i$ ), there is not scheduling point  $t \in S_i$  such that  $W_i(t) \leq t$ . For this latter case, the presented method cannot be used to compute a worst-case response time (i.e., some scheduling points after the deadline must be considered).

Since the size of  $S_i$  depends on  $\sum_{j=1}^{i-1} \lfloor \frac{D_i + J_i}{T_j} \rfloor$ , then the algorithm runs in pseudo-polynomial time. Note that computing the smallest fixed-point  $W_i(t) = t$  using successive approximation is also performed in pseudo-polynomial time.

### 3 A FPTAS for feasibility analysis of task

#### 3.1 Approximating Request Bound Function

For synchronous task systems without release jitters, the worst-case activation scenario for the tasks occurs when they are simultaneously released [7]. When tasks are subjected to release jitters, then the worst-case processor workload occurs when tasks are simultaneously available after  $J_i$  units of time (i.e., when their input data are available). If we assume that tasks become simultaneously available by time 0, then the worst-case workload in a processor busy period is defined by the release at time  $-J_i$ . According to such a scenario, the total execution time requested at time  $t$  by a task  $\tau_i$  is defined by [11]:  $\text{RBF}(\tau_i, t) \stackrel{\text{def}}{=} \left\lceil \frac{t + J_i}{T_i} \right\rceil C_i$ .

The RBF function is a discontinuous function with a “step” of height  $C_i$  every  $T_i$  units of time. In order to approximate the request bound function according to an error bound  $\epsilon$  (accuracy parameter,  $0 < \epsilon < 1$ ), we use the same principle as in [2, 3]: we consider the first  $(k - 1)$  steps of  $\text{RBF}(\tau_i, t)$ , where  $k$  is defined as  $k = \lceil 1/\epsilon \rceil - 1$  and a linear approximation, thereafter. The approximate request bound function is defined as follow:

$$\delta(\tau_i, t) = \begin{cases} \text{RBF}(\tau_i, t) & \text{for } t \leq (k - 1)T_i - J_i, \\ C_i + (t + J_i) \frac{C_i}{T_i} & \text{otherwise.} \end{cases} \quad (3)$$

Thus, up to  $(k - 1)T_i$  no approximation is performed to evaluate the total execution requirement of  $\tau_i$ , and after that it is approximated by a linear function with a slope equal to the utilization factor of  $\tau_i$ .

#### 3.2 Approximation scheme

In [11] is shown that a static-priority task system with release jitters is feasible, iff, worst-case response times of tasks are not greater than their relative deadlines. This

problem is known as the *release jitter problem*. An alternative way is to define a time demand approach using the principles of the well-known exact feasibility test presented for the rate monotonic scheduling algorithm in [6].

The cumulative request bound function at time  $t$  is defined by:  $W_i(t) \stackrel{\text{def}}{=} C_i + \sum_{j=1}^{i-1} \text{RBF}(\tau_j, t)$ . A task  $\tau_i$  is feasible (with a constrained relative deadline) iff, there exists a time  $t$ ,  $0 \leq t \leq D_i$ , such that  $W_i(t) \leq t$ . Since request bound functions are step functions, then  $W_i(t)$  is also a step function that increases its value of  $C_i$  for every scheduling point in the following set  $S_i = \{t = bT_a - J_a; a = 1 \dots i, b = 1 \dots \lfloor \frac{J_i + D_i}{T_a} \rfloor\} \cup \{D_i\}$ . The feasibility test can then be formulated as follows: if there exists a scheduling point  $t \in S_i$ , such that  $W_i(t)/t \leq 1$  then the task is feasible.

To define an approximate feasibility test, we define an approximate cumulative request bound function as:  $\widehat{W}_i(t) \stackrel{\text{def}}{=} C_i + \sum_{j=1}^{i-1} \delta(\tau_j, t)$ . According to the error bound  $\epsilon$  leading to  $k = \lceil 1/\epsilon \rceil - 1$ , we define the following testing set  $\widehat{S}_i \subseteq S_i$ :  $\widehat{S}_i \stackrel{\text{def}}{=} \{t = bT_a - J_a; a = 1 \dots i, b = 1 \dots k\} \cup \{D_i\}$ .

A simple implementation of this approximate feasibility test leads to a  $O(n^2/\epsilon)$  algorithm. This is a FPTAS since the algorithm is polynomial according the input size and the input parameter  $1/\epsilon$ . We now prove the correctness of this approximate feasibility test.

#### 3.3 Correctness of Approximation

The key point to ensure the correctness is:  $\delta(\tau_i, t)/\text{RBF}(\tau_i, t) \leq (1 + \epsilon)$ . This result will then be used to prove that if a task set is stated infeasible by the FPTAS, then it is infeasible under a  $(1 - \epsilon)$  speed processor.

**Theorem 3**  $\forall t \geq 0$ , we verify that:  $\text{RBF}(\tau_i, t) \leq \delta(\tau_i, t) \leq (1 + \frac{1}{k})\text{RBF}(\tau_i, t)$  where  $k = \lceil \frac{1}{\epsilon} \rceil - 1$ .

*Proof:* We first prove the first inequality: for all  $t \in [0, (k - 1)T_i - J_i]$ ,  $\delta(\tau_i, t) = \text{RBF}(\tau_i, t)$ . For  $t > (k - 1)T_i - J_i$ ,  $\delta(\tau_i, t) = C_i + (t + J_i) \frac{C_i}{T_i} = C_i \left(1 + \frac{t + J_i}{T_i}\right)$ .

As a consequence:  $\delta(\tau_i, t) \geq \left\lceil \frac{t + J_i}{T_i} \right\rceil C_i = \text{RBF}(\tau_i, t)$ .

We now prove the second inequality of the statement: If  $\delta(\tau_i, t) > \text{RBF}(\tau_i, t)$  then since  $t > (k - 1)T_i - J_i$  then  $k - 1$  steps before approximating the request bound function, we verify:

$$\text{RBF}(\tau_i, t) \geq kC_i \quad (4)$$

Furthermore,  $\delta(\tau_i, t) - \text{RBF}(\tau_i, t) \leq C_i$ : this is obvious if  $t \in [0, (k - 1)T_i - J_i]$  since  $\delta(\tau_i, t) = \text{RBF}(\tau_i, t)$ , and if  $t > (k - 1)T_i - J_i$ , then:  $\delta(\tau_i, t) - \text{RBF}(\tau_i, t) = C_i + (t + J_i) \frac{C_i}{T_i} - \left\lceil \frac{t + J_i}{T_i} \right\rceil C_i \leq C_i$ .

As a consequence:  $\delta(\tau_i, t) \leq \text{RBF}(\tau_i, t) + C_i$  and using inequality (4), we obtain the result:  $\delta(\tau_i, t) \leq (1 + \frac{1}{k})\text{RBF}(\tau_i, t)$ . As a consequence, both inequalities are verified. ■

Using the same approach presented in [2, 3], we can establish the correctness of approximation.

**Theorem 4** *If there exists a time instant  $t \in (0, D_i]$ , such that  $\widehat{W}_i(t) \leq t$ , then  $\tau_i$  is feasible (i.e.,  $W_i(t) \leq t$ ).*

*Proof:* Directly follows from Theorem 3 ■

**Theorem 5** *If  $\forall t \in (0, D_i]$ ,  $\widehat{W}_i(t) > t$ , then  $\tau_i$  is infeasible on a processor of  $(1 - \epsilon)$  capacity.*

*Proof:* Assume that  $\forall t \in (0, D_i]$ ,  $\widehat{W}_i(t) > t$ , but  $\tau_i$  is still feasible on a  $(1 - \epsilon)$  speed processor. Since assuming  $\tau_i$  to be feasible upon a  $(1 - \epsilon)$  speed processor, then there must exist a time  $t_0$  such that  $\tau_i: W_i(t_0) \leq (1 - \epsilon)t_0$ . But, using Theorem 3 we verify that  $\widehat{W}_i(t) \leq (1 + \frac{1}{k})W_i(t)$ , where  $k = \lceil \frac{1}{\epsilon} \rceil - 1$ , then for all  $t \in (0, D_i]$ , the condition  $\widehat{W}_i(t) > t$  implies that:  $W_i(t) > \frac{t}{1 + \frac{1}{k}} > \frac{k}{k+1}t \geq (1 - \epsilon)t \quad \forall t \in (0, D_i]$ .

As a consequence, a time  $t_0$  such that  $W_i(t_0) \leq (1 - \epsilon)t_0$  cannot exist and  $\tau_i$  is infeasible. ■

To conclude the correctness, we must prove that scheduling points are sufficient.

**Theorem 6** *For all  $t \in \widehat{S}_i$  such that  $\widehat{W}_i(t) > t$ , then we also verify that:  $\forall t \in (0, D_i]$ ,  $\widehat{W}_i(t) > t$*

*Proof:* Let  $t_1$  and  $t_2$  be two adjacent points in  $\widehat{S}_i$  (i.e.,  $\nexists t \in \widehat{S}_i$  such that  $t_1 < t < t_2$ ). Since  $\widehat{W}_i(t_1) > t_1$ ,  $\widehat{W}_i(t_2) > t_2$  and the fact that  $\widehat{W}_i(t)$  is a non-decreasing step left-continuous function we conclude that  $\forall t \in (t_1, t_2)$   $\widehat{W}_i(t) > t$ . The property follows. ■

## 4 Approximate Response Time Analysis with release jitters

We shall combine results presented in Sections 2 and 3, in order to define approximate worst-case response times. Using the FPTAS presented in Section 3, we can check that a task is feasible or not. If it is feasible, then we are able to compute an upper bound of the worst-case response time of a task as presented in Section 2.

**Definition 4** *Consider a task  $\tau_i$  such that there exists a time  $t$  satisfying  $\widehat{W}_i(t) \leq t$ , then an approximate worst-case response time is defined by:*

$$t^* \stackrel{\text{def}}{=} \min \left( t \in \widehat{S}_i \mid \widehat{W}_i(t) \leq t \right) \text{ and } \widehat{R}_i^* \stackrel{\text{def}}{=} \widehat{W}_i(t^*) + J_i.$$

We now prove that such a method defines an upper bound of the worst-case response time of task  $\tau_i$ .

**Theorem 7** *For every task  $\tau_i$  such that there exists a time  $t$  satisfying  $\widehat{W}_i(t) \leq t$ , then:  $R_i^* \leq \widehat{R}_i^*$*

*Proof:* Let  $t$  be a scheduling point such that  $\widehat{W}_i(t) \leq t$ . From the approximate feasibility test, we verify that  $\tau_i$  is feasible: there exists a time  $t^*$  such that  $W_i(t^*) \leq t^*$  and  $t^* \leq t$ . Since  $R_i^* = W_i(t^*) + J_i$  and  $\widehat{R}_i^* = \widehat{W}_i(t) + J_i$  then, it follows from properties of the approximate feasibility test that  $R_i^* \leq \widehat{R}_i^*$ . ■

It can be shown that this method does not lead to an approximation algorithm (i.e., with the expected bounded error presented in Definition 1).

## 5 Conclusion

The existence of an approximation scheme (or weakly an approximation algorithm) to solve that problem is still an interesting open issue.

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## References

- [1] R. Bril, W. Verhaege, and E. Pol. Initial values for on-line response time calculations. *proc. Int Euromicro Conf. on Real-Time Systems (ECRTS'03), Porto, 2003.*
- [2] N. Fisher and S. Baruah. A polynomial-time approximation scheme for feasibility analysis in static-priority systems with bounded relative deadlines. *Proceedings of the 13th International Conference on Real-Time Systems, Paris, France, pages 233–249, 2005.*
- [3] N. Fisher and S. Baruah. A polynomial-time approximation scheme for feasibility analysis in static-priority systems with arbitrary relative deadlines. In I. C. Society, editor, *Proceedings of the EuroMicro Conference on Real-Time Systems*, pages 117–126, 2005.
- [4] M. Joseph and P. Pandya. Finding response times in a real-time systems. *The Computer Journal*, 29(5):390–395, 1986.
- [5] J. Lehoczky. Fixed priority scheduling of periodic tasks with arbitrary deadlines. *proc. Real-Time System Symposium (RTSS'90)*, pages 201–209, 1990.
- [6] J. Lehoczky, L. Sha, and Y. Ding. The rate monotonic scheduling algorithm: exact characterization and average case behavior. *proc. Real-Time System Symposium (RTSS'89)*, pages 166–171, 1989.
- [7] J. C. Liu and J. W. Layland. Scheduling algorithms for multiprogramming in hard real-time environment. *Journal of the ACM*, 20(1):46–61, 1973.
- [8] Y. Manabe and S. Aoyagi. A feasible decision algorithm for rate monotonic and deadline monotonic scheduling. *Real-Time Systems Journal*, pages 171–181, 1998.
- [9] L. Sha, T. Abdelzaher, K.-E. arzen, A. Cervin, T. Baker, A. Burns, G. Buttazzo, M. Caccamo, J. Lehoczky, and A. K. Mok. Real time scheduling theory: A historical perspective. *Journal of Real-Time Systems*, pages 101–155, 2005.
- [10] M. Sjodin and H. Hansson. Improved response time analysis calculations. *proc. IEEE Int Symposium on Real-Time Systems (RTSS'98)*, 1998.
- [11] K. Tindell. *Fixed Priority Scheduling of Hard Real-Time Systems*. PhD thesis, University of York, 1994.