

Project Report of CSCE790i

# Performance of Algorithms for Scheduling Soft Real-Time System

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## Abstract

In this research project the performance of two well-known scheduling policies for Real Time Systems based on fixed priority, namely Enhanced (m,k)-firm Guarantee and Weakly hard scheduling, will be compared in terms of the overall schedulability of a number of randomly generated task sets . Both methods are based on the common assumptions. A new scheduling algorithm will then be derived by combining the advantages of both algorithm for the purpose of improving the overall schedulability of the real time system.

**Keywords**        Scheduling, task, job, Real-time system, Weakly hard, Constraints, Schedulability, Interference, WCIP, Independent Job .

## 1 Introduction

In some multimedia real time system like video conference system, the loss of some image frames due to their missing of deadlines are tolerable. This class of system are sometimes refer to as soft real-time system or firm real time system.

A real time system is said to have  $(m,k)$ -firm deadlines if at least  $m$  out of any  $k$  consecutive jobs from the task( or, say, a job chain) must meet their deadlines. Similarly, the concept of weakly hard real-time system refers to the real-time system that can tolerate a clearly specified number of deadlines

In[9], Ramanathan first proposed a scheduling technique that can provide the deterministic  $(m,k)$ -firm guarantee. In this technique, a simple algorithm is adopted to divide the jobs of each task into two categories: mandatory and optional. All mandatory jobs are required to meet their deadline. The advantages of this method is that it gives a simple formula to partition the jobs into mandatory or optional and the  $(m,k)$ -firm guarantee requirement is satisfied if all mandatory instances meet their deadlines. In[8], Quan developed the idea in[1] by adjusting the relative position of the mandatory jobs of each task. The beauty of this technique is that it reflects part of the real execution situation between jobs in terms of execution interference and by shifting the WCIPs of each task away, the execution interference from higher priority tasks with lower priority tasks was reduced. Thus the worst case response time of mandatory jobs is reduced. Thereafter the schedulability of the whole task set is improved. However, the problem of evenly distribution of mandatory jobs still exists and the relative positions of mandatory jobs of different values of  $m$  and  $k$  with the same ratio keeps the same. So some tasks might be designated as "optional" even when they would meet their deadlines if they were scheduled and conversely, some tasks might be designated as "mandatory" and invoked even when they would miss their deadlines. In[2], Bernat uses a method based on precise timing analysis to decide the schedulability of a given task set.

The difference between these two method is that for the former one, only the mandatory job are invoked and any missing deadline will cause dynamic failure of the whole task set. The analysis of the schedulability of task sets is based on the analysis of critical instance of each task. Similar work has been done by Lehoczky in analyzing hard schedulability of task sets with fixed priorities. The basic idea is that if the critical instance of each task can meet the deadline, then the whole task set is schedulable. For the latter one, all jobs of each task are invoked and runs until completion even though it misses the deadline. The schedulability of the task set is tested by offline computation on the response time of each job of each task throughout its hiperperiod and the constraint is checked based on the computation result. The advantage of this method is that precise timing ananlysis is done on the jobs of each of the task in respect to computing the values of  $F_i(k)$ ,  $S_i(k)$ ,  $R_i(k)$ . Thus, the exact running status of each job can be predicted and more information on the execution situation can be achieved by

an offline computation. Moreover, the precise quantification method in this algorithm can be used widely in the real-time system area with slightly modification. Although the computation of it may consume considerable time, it is not a major concern for offline computation. The disadvantage is that since it adopt delayed finalization to handle the missing of deadline. All jobs will be supposed to continue running until it finishes even though it may miss the deadline. This has two consequences: the first one is that part the processor time was occupied by those jobs missing the deadline(if they do) and this not only degrade the system efficiency but also imposed some limit on the utilization bound that algorithm can process. For the prevailing cases in overloaded system where the resource utilization is larger than 1, the algorithm will face the embarrassment that it can not guarantee the test has covered all the worst case response time of jobs and thus cannot tell whether the weakly hard constraint can be satisfied.; the second one is the so called "domino effect" which is also a typical phenomenon that may happen with EDF when the system is overloaded. Similar with EDF, this happens because the delayed finalization of the first job make the following jobs even harder to schedule and cause them to miss the deadline. However, if the two algorithms are combine together, better result can be expected in improving the overall schedulability of the system. That's a main work of this project.

This rest part of this report is organized as followed: in section2, we will introduce the process model of our system; in section 3, we introduce the algorithms to be compared in specification; in section 4, we gave the experiment results and our analysis on it; in section 5, we consider real time systems with variable execution time and arbitrary deadlines; in section 6, we draw the conclusion based on what we've done in the previous sections.

## 2 System and Process Model

Before we describe the shedding algorithms we have considered in our performance study, we define the following notations to refer to the parameters of a task set with  $n$  independent periodic tasks,  $\tau = \{\tau_1, \tau_2, \dots, \tau_n\}$ , arranged in the decreasing order of their priorities. Each instance of task is called a *job*. The  $j$ th job of  $\tau_i$  is denoted as  $\tau_{ij}$ . The following notations are defined to refer to the parameters of task  $\tau_i$  :

- $O_i$  : the release time of the first job of  $\tau_i$ , refered to as initial time.
- $T_i$  : the interval between two consecutive job release times of  $\tau_i$ , refered to as *period*.

- $D_i$  : the maximum time allowed from the release to the completion of  $\tau_i$ 's job, referred to as deadline.
- $C_i$  : the maximum time needed to complete  $\tau_i$  without any interruption, referred to as execution time.
- $m_i$  and  $k_i$  : the (m,k) constraint for  $\tau_i$ , which mandates that at least m out of k consecutive jobs of  $\tau_i$  must meet their deadlines to avoid any dynamic failure.

Other assumptions associated with the notations are:

- Deadlines for hard tasks cannot be missed;
- Task periods are fixed and cannot be modified;
- Deadline equals period ( $D = T$ );
- Optional( or soft ) tasks are assigned the lowest priority

All tasks are assumed to be scheduled on a uniprocessor system and both of the methods adopt a fixed priority scheduler that always executes the runnable task with the highest priority . Static scheduling assumes that the complete knowledge regarding the task set and its constraints a priori and once a task's priority is assigned it does not have to be reevaluated as time progresses. This has several advantages: (i) scheduling can often be done with an offline analysis. (ii)it incurs lower overhead; (iii) the implementation is relatively simple; (iiii) it gives a designer control over task priorities.

### 3 Specification of Algorithms to Compare

#### 3.1 Algorithm 1 (Alg-WH) : Weakly Hard Scheduling Algorithm [2]

**Definition 1.** *The task's  $\mu$  – pattern is the worst case pattern of met and lost deadlines of task  $\tau_i$ . It is denoted by  $\mu_i$  and it is defined as:*

$$\mu_i(k) = \begin{cases} 1 & \text{if } R_i(k) \leq D_i \\ 0 & \text{Otherwise} \end{cases} \quad (1)$$

where  $R_i(k)$  is the worst case response time of the task at invocation  $k \geq 1$ , this value is given by  $R_i(k) = F_i(k) - S_i(k)$ , where  $F_i(k)$  is the worst case finishing time of the  $k$ -th invocation of the task and  $S_i(k)$  is the release time.

It has been shown in [2] that if the utilization factor  $U \leq 1$  then the worst case task  $\mu$ -pattern is repeated every hyperperiod at level  $i$ , given by:  $h_i = \text{lcm}\{T_j \mid 1 \leq j \leq i\}$ .

**Definition 2.** Bernat [2] A task  $\tau$  is said to satisfy the weekly hard temporal constraint "meets any  $n$  in  $m$  deadlines" ( $m \geq 1$  and  $0 \leq n \leq m$ ), denoted by  $\tau \vdash \binom{n}{m}$  if, in any window of  $m$  consecutive invocations of the task, there are at least  $n$  invocations in any order that meet the deadline. (i.e., in any window of size  $m$  in the corresponding  $u$ -pattern, there are at least  $n$  1s).

This definition corresponds to the definition of  $(m, k)$ -firm model in the Alg-EG algorithm to be compared.

In order to check whether the tasks of a system satisfy their weakly hard specifications, we must compute the  $\mu$ -pattern of each task  $\tau_i$ . In [2], Bernat proved that if the utilization of the task set is not larger than 1, the  $\mu$ -patterns are closed, so their analysis can be reduced to a finite sequence. Thus, we can check whether the task set meets the weakly hard constraint by analyzing the  $\mu$ -pattern with a sliding window of length  $k_i$  in its *Hyperperiod* [2].

newline

**Theorem 1.** (Bernat [2]): Give a task set  $T = \{\tau_1, \tau_2, \dots, \tau_n\}$ , made up of independent periodic tasks scheduled under fixed priority, then all tasks (and the  $\mu$ -pattern correspondingly) are closed if and only if  $U \leq 1$ .

Now, the problem is how to compute the  $\mu$ -pattern. In [2], Bernat gave the specification of doing that. Formally, we need to compute  $F_i(k), R_i(k)$  of each job  $k$  within  $2 * h_i + \max\{O_j\}[1]$ , where  $F_i(k)$  is the minimum  $t$  that makes equation (2) hold;  $t$  is the time required to perform  $k$  computations of the task, plus the interference from higher priority tasks suffered during that period, plus the time the processor has not been used (or, say, the idle time at level  $i$ ) from  $t=0$  until the start time of the task by tasks with priority higher than or equal to  $\tau_i$ . Formally,

$$t = kC_i + \text{Int}f_i(t) + \text{Idle}_i(S_i(k)) \quad (2)$$

The Interference function  $\text{Int}f_i(t)$  is given by:

$$Intf_i(t) = \sum_{\tau_j \in hp(i)} \lceil \frac{t - O_j}{T_j} \rceil C_j \quad (3)$$

where  $hp(i)$  is the set of task of higher priority than task  $\tau_i$ .  $Idle_i(t)$  is the time the processor can be used by lower priority tasks between  $[0, t)$ . This can be computed by adding a virtual task  $\bar{\tau} = (\bar{T} = t, \bar{D} = t, \bar{C}, \bar{O} = 0)$  of lower priority than  $\tau_i$  and computing the maximum value  $\bar{C}$ , such that task  $\bar{\tau}$  meets its deadline. Formally,

$$Idle_i(t) = \max\{ \bar{C} \mid \bar{\tau} \text{ is schedulable } \} \quad (4)$$

The finish time for task  $\tau_i$  is given by:

$$t = \bar{C} + \sum_{\tau_j \in hep(\tau_i)} \lceil \frac{t - O_j}{T_j} \rceil C_j \quad (5)$$

And checking if  $t < \bar{D}$ .  $hep(\tau_i)$  is the set of tasks of higher or equal priority than  $\tau_i$ . After we compute the  $R_i(k)$  of each job  $k$  such that  $S_i(k) \leq 2 * h_i + \max\{O_j\}$ , we can compute the corresponding  $\mu - pattern$  according to (1) and check the result  $\mu - patterns$  to test the schedulability of the task set.

In our algorithm3, we developed this method to make it fit into the computation of (m,k)-firm guarantee and use the new method in deciding the local optimal pattern of task  $\tau_i$ .

### 3.2 Algorithm 2 (Alg-EG) : Enhanced (m,k)-firm Guarantee algorithm [8]

When scheduling a task set with (m,k) constraints according to a fixed-priority assignment, one critical step is to determine for each task whether its execution is mandatory or optional.

**Definition 3.** The  $(m,k)$ -pattern of task  $\tau_i$ , denoted by  $\Pi_i$ , is a binary string  $\Pi_i = \{\pi_{i1}\pi_{i2} \dots \pi_{ik_i}\}$  which satisfies the following: (i)  $\pi_{ij}$  is a mandatory job if  $\pi_{ij} = 1$  and optional if  $\pi_{ij} = 0$ , and (ii)  $\sum_{j=1}^{k_i} \pi_{ij} = m_i$ . By repeating the (m,k)-Pattern  $\Pi_i$ , we get a mandatory job pattern for  $\tau_i$ . It's not difficult to see that the (m,k) constraints for  $\tau_i$  can be satisfied if the mandatory jobs of  $\tau_i$  are selected accordingly.

After given the (m,k) constraint, how to find the proper (m,k)-pattern to schedule the task set is a difficult problem. In [8], Quan has shown that selecting the "optimal" (m,k)-pattern for the given (m,k) constraint is NP-hard in the strong sense. In [9], Ramanathan proposed a simple algorithm to select the mandatory jobs and proved that so long

as the (m,k)-pattern is determined by the following formular, the (m,k)constraint is guaranteed.

$$\pi_{ij} = \begin{cases} 1 & \text{if } j = \lfloor \lceil \frac{(j-1) \times m_i}{k_i} \rceil \times \frac{k_i}{m_i} \rfloor + 1 \\ 0 & \text{otherwise} \end{cases} \quad j = 1, 2, \dots, k_i \quad (6)$$

In [8], Quan developed this idea by shifting away the WCIPs between tasks away and proposed a new algorithm which greatly improved the schedulibility of the task set. The way that Quan determines the (m,k)-pattern of the task set is to modify the formular(2) like this:

$$\pi_{ij} = \begin{cases} 1 & \text{if } j = \lfloor \lceil \frac{((j-1)+s_i) \times m_i}{k_i} \rceil \times \frac{k_i}{m_i} \rfloor + 1 \\ 0 & \text{otherwise} \end{cases} \quad j = 1, 2, \dots, k_i \quad (7)$$

where  $s_i \leq 0$  and  $s_i \in Z$ . Note that the new (m,k)-pattern can be viewed as rotating the (m,k)-pattern in (2) right by  $s_i$  bits. It's easy to see the new (m,k)-pattern satisfies the (m,k)constraints as well. Quan also present the algorithm to compute the proper value of  $S_i$  in [8].

When it comes to checking the schedulability of a task set for a given set of (m,k)- patterns, Quan also presented a sufficient condition [8] to predict the schedulability of mandatory jobs in polynomial time. It's shown by the theorm below.

**Theorem 2.** (Quan[8]): *Given two task set  $T$  and  $T'$  with  $T'_i \leq T_i$ ,  $C'_i = C_i$ ,  $m'_i = m_i$ ,  $k'_i = k_i$ , and  $T'_j$  divides  $T'_i$  if  $T'_j \leq T'_i$ . With the given (m, k) – patterns, if  $\sum_{j \leq i} (l_{ij} \times C_j) / T'_i \leq 1$ ,  $l_{ij}$  is the maximum number of mandatory jobs during any time interval of length  $T'_i$ , then  $T$  is schedulable.*

The advantage of this sufficient condition is that it takes only  $O(n^k)$  time to implement so that it can tell a task set is schedulable quickly if it is. However, it also has the disadvantage that for some task sets with (m,k)-pattern given by (3), it can not tell they are schedulable while actually they are. This behaviour is somewhat similar to that of the Liu and layland's upper bound (the sufficient condition in[7]). Due to this obvious reason in [8] Quan used simulation to achieve the goal. The advantage of using simulation is that it's straightforward and easy to implement. The disadvantage is that it may take very long time. This is due to the fact that simulation has to run all tasks until the Hyperperiod[2](or twice the Hyperperiod if offsets are present);for non-trivial systems this can be very large. In [2],Bernat introduce a method based on offline computation to check the schedulability of weakly hard system and showed that it's more efficient. However, this method cannot be directly used on (m,k)-firm guarantee. Here

we developed the method in [2] to make it fit into the (m,k)-firm guarantee and use it to check the schedulability of the task set offline.

We now show how to check whether each mandatory job can meet its' deadline. Before doing that, we need to show how to compute  $F_i(k), R_i(k)$  of each job with the given (m,k)-pattern. To achieve this goal, we shall develop the method in Algorithm 1 by incorporating the (m,k)-pattern to it. Formally, given the (m,k)-pattern of a task set, we need to compute  $F_i(k), S_i(k), R_i(k)$  of the mandatory jobs within the first hyperperiod at level i, where  $F_i(k)$  is the minimum t that makes equation(8) hold; t is the time required to perform k computations of the task, plus the interference from higher priority tasks suffered during that period, plus the time the processor has not been used (or, say, the idle time at level i) from t=0 until the start time of the task by tasks with priority higher than or equal to  $\tau_i$ . Formally,

$$t = \sum_{1 \leq l \leq k} \pi_{il} C_i + Intf_i(t) + Idle_i(S_i(k)) \quad (8)$$

The Interference function  $Intf_i(t)$  is given by:

$$Intf_i(t) = \sum_{\tau_j \in hp(i)} \sum_{1 \leq l \leq \lceil \frac{t-O_j}{T_j} \rceil} \pi_{jl} C_j \quad (9)$$

where  $hp(i)$  is the set of task of higher priority than task  $\tau_i$ .  $Idle_i(t)$  is the time the processor can be used by lower priority tasks between  $[0, t)$ . This can be computed by adding a virtual task  $\bar{\tau} = (\bar{T} = t, \bar{D} = t, \bar{C}, \bar{O} = 0)$  of lower priority than  $\tau_i$  and computing the maximum value  $\bar{C}$ , such that task  $\bar{\tau}$  meets its deadline. Formally,

$$Idle_i(t) = \max \{ \bar{C} \mid \bar{\tau} \text{ is schedulable } \} \quad (10)$$

The finish time for task  $\tau_i$  is given by:

$$t = \bar{C} + \sum_{\tau_j \in hp(i)} \sum_{1 \leq l \leq \lceil \frac{t-O_j}{T_j} \rceil} \pi_{jl} C_j \quad (11)$$

And checking if  $t < \bar{D}$ .  $hep(\tau_i)$  is the set of tasks of higher or equal priority than  $\tau_i$ .

Observe that  $\tau_i$ 's mandatory jobs corresponding to bit  $\pi_{ij} = 1$  can be viewed as a periodic event  $E_i$  with period  $k_i T_i$ , the concept of hyperperiod at level i in algorithm 1 should also be modified correspondingly. It's given by:  $h_i = lcm\{k_j T_j \mid 1 \leq j \leq i\}$ .

In [1], Bernat showed that it's only necessary to compute  $R_i(k)$  for values of  $k$  such that  $S_i(k) \leq 2 * h_i + \max\{O_j\}$ .

### 3.3 Algorithm 3 (Alg-IG) : Improved (m,k)-firm Guarantee algorithm

This algorithm is a development of Alg-EG algorithm. As we mentioned before, although in most cases Alg-EG algorithm works better than Alg-WH algorithm (this will be shown by the experiment result), it has some limitation in partitioning the jobs, for example designating the some job that will miss the deadline as *mandatory*. Sometime maybe just "one job" is wrongly selected as *mandatory* and it causes the whole task set unschedulable because it breaks the guarantee that *each* of the *mandatory* job meet the deadline. For example, suppose task  $\tau_i$  has the following  $\mu$ -pattern for each *independent job* in its hyperperiod (the concept of independent job is introduced in section 3.4) :

"11001110111101111001"

Suppose it has the (m,k)-constraint as (3,5). Then the evenly distributed pattern (e.g. "10101") does not work for this job while evenly distributed pattern like "11001" works.

In this Algorithm, we will develop the timing analysis method in Alg-WH and then applied it to Alg-EG to achieve a better pattern. The origin of this idea is that in algorithm Alg-EG, the problem of implicit even distribution of pattern still exists because it still depends on the formula introduced in [9] to determine the pattern, which makes the pattern still not flexible. Although the schedulability of the whole system is improved by shifting the WCIPs[8] away, there are still a lot of task sets that's are schedulable (e.g. the task the above example) but cannot be scheduled by Alg-EG. The reason is because only part of execution situation of the whole system is selected in Alg-EG. In order to improve the schedulability of the whole task sets, specifically, to schedule those schedulable task sets that cannot be scheduled by Alg-EG, more of the real execution situation need to be reflected in choosing the pattern. Fortunately, the precise timing analysis method in Alg-WH provides us a powerful tool in doing that. In [8], Quan proved that the problem of "selecting" the "optimal" (m,k) pattern for each task is *NP-hard*. And we conjecture the problem of finding the "optimal" *loop style (m,k)-pattern* for each task is also *NP-hard*. However, with our new method, the locally optimal *loop style (m,k)-pattern* of task  $\tau_i$  can be decided easily. Before doing that, we need prove the theorem below first:

**Theorem 3.** (*loop pattern theorem*) *Given an (m,k)-firm constraint for a task  $\tau_i$ , any loop*

pattern with  $m$  1's out of  $k$  positions( $k$  is the length of the piece of pattern to be looped) can meet the  $(m,k)$ -firm constraint.

*proof:* the proof of this theorem can be done with mathematical induction or simply by exhaustive enumeration for any given values of  $m$  and  $k$ . For example,give  $(m_i, k_i)$ -firm constraint as  $(3,5)$  for task  $\tau_i$ , then any loop pattern based on loop piece "11100", "11001","10101",or "11010" ... can meet the requirement." We will leave this part to interested readers.

This theorem forms the basis of our new algorithm Alg-IG.

In[8], Quan introduced the concept of execution interference and uses it as the basis of Alg-EG. Now we generalize this concept here to make it fit into our new approach.

**Definition 4.** Given a task set and the  $(m,k)$ -pattern for each task, the execution interference suffered by  $\tau_{ik}$  from all higher priority tasks, denoted by  $I_{ik}$ , equals the total portions of the execution time of all mandatory jobs of higer priority tasks that fall inside  $[(k-1)T_i + O_i, kT_i + O_i]$ .

The concept of *excecution interference* of  $\tau_{ik}$  also forms the basis of our new algorithm to be discussed next. and the precise computation of  $I_{ik}$  is outlined in Algorithm 3.1 in the appendix section.

Next let's see how to determine the reasonable pattern with our new algorithm. Before doing that,we should point out the fact that for any loop pattern for task  $\tau_i$ , the pattern of all jobs at invocation  $j + nk_i$  ( $kn \geq 0$ ) is the same and equal to the pattern at  $j$  ( $j$  is a given value,  $1 \leq j \leq k_i$ ), so all invocation of the jobs can be categorized into  $k_i$  positions,namely: $1 + nk_i, 2 + nk_i, \dots, k_i + nk_i$ . Since all job invocations at the same position has the same attribute(*mandatory* or *optinonal*), we call those position with mandatory jobs as *mandatory positions* and those with optional jobs as *optional position*. If any job at the mandatory positions miss the deadline we will say the task set miss the constraint. Since the partition of positions is equal to the partition of  $(m,k)$ -patterns, we should try to avoid partitioning those positions with at least one job miss the deadline as *mandatory positions*.

As we know, whether a job will miss the deadline is determined by how much execution interference  $I_{ik}$  it suffers from higher priority tasks.

We define  $\hat{I}_{ij}$ , for  $1 \leq j \leq k_i$ , as *maximum interference* that each job at invocation  $j + nk_i$  suffers from higher priority tasks. next we will show how to precisely compute the *maximum interference* that job at invocation  $j + nk_i$  suffers from higher priority tasks.

Then the mandatory jobs is selected according to the values of *maximum interference* of jobs at each invocation of  $j + nk_i (1 \leq j \leq k_i)$ . The idea is that for a given  $j (1 \leq j \leq k_i)$ , the less the *maximum interference* of job at invocations  $j + nk_i$ , the less the possibility that the jobs at invocations  $j + nk_i$  will miss the deadline. So we should select those jobs with less *maximum interference* at invocation  $j + nk_i$  as mandatory while designate those with larger *maximum interference* at invocation  $j + nk_i$  as optional. This idea is implemented with the algorithm 1. The algorithm uses Alg-EG as a start point. The reason is that Alg-EG produces evenly distributed patterns with their WCIPs shifted away. This kind of pattern has both advantages and disadvantages. Since we usually schedule higher priority tasks first. For tasks that can be scheduled with evenly distributed pattern, we should try to do so because each mandatory instance of high priority tasks can be regarded as a *interference block* to low priority tasks. By adopting evenly distributed pattern with their WCIPs shifted away, from the perspective of low priority jobs, the *interference blocks* are made to *sparingly* distributed to the most extend and it improves the chance for low priority jobs to be schedulable. So evenly distributed pattern with WCIPs shifted away, if schedulable, should try to be used. However, for the cases that evenly distributed pattern doesn't work, we should adjust the pattern correspondingly.

The advantage of this method is that after determining the (m,k)-pattern, we already know whether it's schedulable or not. For example, if  $\max\{\hat{I}_{ik}|k$  is the job invocations chosen as mandatory  $\} > (D_i - C_i)$ , then we know this task is not schedulable by our algorithm, otherwise it is schedulable.

Algorithm Alg-IG is quite effective in improving the schedulability of task sets with (m,k) constraints. We will give experimental results later to illustrate this.

## 4 Performance Evaluation

In this section we will compare the performance of the three algorithms described above by experiment results. In our experiments, we randomly generate some task sets with 4 or 5 tasks. The period of each task is randomly generated from a uniform distribution between 10 to 50, and the deadline of each task is assumed to equal to its period. The  $m_i$  and  $k_i$  values are also randomly selected, where  $k_i$  is uniformly distributed between 2 and 10, and  $m_i$  is uniformly distributed between 1 and  $k_i$ . We partition the utilization factor values into intervals of length 0.2. Then, the execution time of each task is randomly selected such that the utilization values of the resulting task sets are uniformly distributed within each interval. To reduce statistical

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**Algorithm 1** Algorithm 3.3.1 : Determine the proper regular (m,k)-pattern of given task set

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**Input:** Task set  $T = \{\tau_1, \tau_2, \dots, \tau_n\}$ , where  $\tau_i = \{O_i, T_i, D_i, C_i, m_i, k_i\}$

**Output:** whether the task set is schedulable

TRUE: schedulable with  $\pi_{i1}, \dots, \pi_{ij}$  //(m,k)-pattern of T

FALSE: not schedulable

**for**  $i$  from 1 to  $n$  **do**

    determine (m,k)-pattern of  $\tau_i$  according to (7) ;

**if**  $\tau_i$  is not schedulable **then**

        MandatoryJobs = 0;

**for**  $j$  from  $k_i$  to 1 **do**

$\hat{I}_{ij}$ =0;

**while**  $S_i(j + nk_i) \leq 2 * h_i + \max\{O_i\}$  **do**

$I_{ij}$  is calculated according to Algorithm 7.1;

**if**  $\hat{I}_{ij} < I_{ij}$  **then**

$\hat{I}_{ij} = I_{ij}$ ;

**end if**

**end while**

**if**  $\hat{I}_{ij} < (D_j - C_j)$  AND MandatoryJobs  $< m_i$  **then**

$\pi_{ij} = 1$ ;

**else**

$\pi_{ij} = 0$ ;

**end if**

**end for**

**end if**

**end for**

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errors, the number of task sets schedulable by at least one of the approaches is no less than 50 within each interval, or at least 5000 different task sets have been generated for the interval.

Utilization	No. of Schedulable Task Sets			Improvement(%)	
	Alg_WH	Alg_EG	Alg_IG	IG - WH	IG - EG
0.6 - 0.8	11	11	11	0	0
0.8 - 1.0	17	29	29	70.5	0
1.0 - 1.2	4	65	82	1500	26.1
1.2 - 1.4	0	64	112	NaN	75
1.4 - 1.6	0	3	13	NaN	333.3
1.6 - 1.8	0	0	2	NaN	NaN
1.8 - 2.0	0	0	0	NaN	NaN

Table 1: Experimental results comparing the five approaches

The experiment results are collected in Table 1. In our experiments, task sets with utilization values less than 0.6 are all hard schedulable, thus we simply discard them. None of the task set with utilization greater than 2.0 is schedulable with any of the approaches. From Table 1, one can conclude that Alg-EG performs better than Alg-WH whilst Alg-IG outperforms the other two. The improvement become more significant as the task set utilization factor values increase(especially between 1.0-1.6 ). In the experiments, as we expect, a task set is schedulable with Alg-IG as long as it is schedulable with Alg-EG.

## 5 Conclusion

In this project we compared the performance of two well known algorithms in scheduling soft real time systems with fixed priority. And by developing the methodologies in both algorithms, we derived a new off-line algorithms which combines the advantages of both algorithms and achieve better performance than both of them. All of the algorithms above are easy to implement and introduce low scheduler overhead. Simulation results indicate that the proposed method does significantly improved the overall shedulability of a real time system.

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