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Foreword

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I would like to welcome all readers to the first issue of this new not-for-profit open access journal: the Leibniz Transactions on Embedded Systems (LITES). Unless you have come across this journal by accident then you will already understand the key role that embedded systems have in modern life. One can hardly think of a single human activity that is not underpinned by such systems; transport, entertainment, supply lines for supermarkets, health care and drug production, energy production and transmission, robotic manufacturing, control systems and communication media of all kinds are now dependent on the fusion of embedded hardware and software. For researchers in this domain this provides great opportunities but also responsibilities. We need to make sure that society can justifiably rely on technology that is increasing beyond the understanding of most ordinary people. Computer-based technologies have been described as modern magic; it follows that we are therefore magicians. But the spells we cast must be based on sound principles, solid theory and demonstrable performance.

One of the influences that embedded and other IT technology has had in the last decade is in publishing itself. Online services are now the norm. And early and open access to publicly funded research is now rightly demanded by Government bodies and related funding councils. This new journal has been created to meet this challenge. All papers are open access, with copyright being retained by the authors. Moreover, only a small fee is charged to authors due to low operational overheads and the support of Google and the Klaus Tschira Stiftung. But the lack of a physical page limit in an online-only journal does not mean that quality is undermined. All papers are thoroughly reviewed, with only the best work, in terms of originality and rigour, being accepted. Our aim is to evolve an excellent and effective venue for publish scholarly articles. To help achieve this aim LITES benefits greatly from having the name and reputation of Schloss Dagstuhl behind it.

The volume of research material produced world-wide relating to embedded systems has lead to the spawning of many conferences and workshops, special issues and focused publications. In LITES we intend to cater for the broadest collection of relevant topics. We currently have subject editors to cover: the design, implementation, verification, and testing of embedded hardware and software systems; the theoretical foundations; single-core, multi-processor and networked architectures and their energy consumption and predictability properties; reliability and fault tolerance; security properties; applications in the avionics, automotive, telecommunication, medical and production domains; cyber-physical systems; high performance and real-time embedded systems; and hybrid systems. This is an impressive list, but it is not exhaustive. New areas will emerge and new editors will be appointed.

LITES obtains its governance from EDAA (European Design and Automation Association) and EMSIG (Embedded Systems Special Interest Group) as a joint endeavour with Schloss Dagstuhl. EDAA/EMSIG appoint the Editor-in-Chief (EiC) and the subject area editors. The terms for editors is four years, renewable once. All editorial work is done voluntarily.

The first few issues of the journal will contain standard papers that have been through the review process. Later, comments on previously published papers will be allowed and commentaries

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included that will help the reader trace forward the influence of each paper. Comments will be reviewed; commentaries will just need to be passed by the EiC.

I hope that as a reader you will find the papers in this journal of interest and often inspirational. As a researcher I hope you will consider it as a worthy place to entrust your work. All the editorial team will work towards building up the reputation of the journal. I hope the community at large will be part of that journey.

I am proud to be the founding EiC of this journal, but I promise not to include editorials in future issues. The papers are quite capable of introducing themselves.

Alan Burns
A Comparison between Fixed Priority and EDF Scheduling accounting for Cache Related Pre-emption Delays

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\textbf{Abstract}

In multitasking real-time systems, the choice of scheduling algorithm is an important factor to ensure that response time requirements are met while maximising limited system resources. Two popular scheduling algorithms include fixed priority (FP) and earliest deadline first (EDF). While they have been studied in great detail before, they have not been compared when taking into account cache related pre-emption delays (CRPD). Memory and cache are split into a number of blocks containing instructions and data. During a pre-emption, cache blocks from the pre-empting task can evict those of the pre-empted task. When the pre-empted task is resumed, if it then has to re-load the evicted blocks, CRPD are introduced which then affect the schedulability of the task.

In this paper we compare FP and EDF scheduling algorithms in the presence of CRPD using the state-of-the-art CRPD analysis. We find that when CRPD is accounted for, the performance gains offered by EDF over FP, while still notable, are diminished. Furthermore, we find that under scenarios that cause relatively high CRPD, task layout optimisation techniques can be applied to allow FP to schedule tasksets at a similar processor utilisation to EDF. Thus making the choice of the task layout in memory as important as the choice of scheduling algorithm. This is very relevant for industry, as it is much cheaper and simpler to adjust the task layout through the linker than it is to switch the scheduling algorithm.

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\section{Introduction}

Today’s real-time applications are complex systems built up of a large number of interacting tasks running on hardware with non-deterministic performance enhancing features such as caches, pipelines and out-of-order execution. To manage the available resources efficiently, scheduling algorithms are used to determine which task should run and at which time in order to fulfil the functional and temporal requirements of the system. The scheduling algorithms are often pre-empting, in that they allow important tasks to interrupt less important tasks before they have finished. Two popular scheduling algorithms for real-time systems are fixed priority (FP) and
A Comparison between Fixed Priority and EDF Scheduling accounting for CRPD

earliest deadline first (EDF). FP scheduling uses statically defined priorities to run the task with the highest priority first. In comparison, EDF is a dynamic scheduling algorithm that schedules the task with the earliest absolute deadline first. EDF is an optimal scheduling algorithm without pre-emption costs, whereas FP is not, and is therefore typically able to schedule tasks at a higher processor utilisation than FP [20]. However, despite the significant performance benefits over FP, EDF is not widely used in commercial real-time operating systems.

In real-time systems, and especially hard real-time systems, the schedulability of each task must be known in order to verify that the timing requirements will be met. The schedulability of a taskset is determined using information about the scheduling algorithm, the arrival pattern of tasks and the tasks’ worst-case execution times. Worst-case execution times are typically obtained assuming no pre-emption. However, in pre-emptive multi-tasking systems, caches introduce additional cache related pre-emption delays (CRPD) caused by the need to re-fetch blocks belonging to the pre-empted task which were evicted from the cache by the pre-empting task. These CRPD effectively increase the worst-case execution time of the tasks. It is therefore important to be able to calculate, and therefore account for, CRPD when determining if a system is schedulable or not.

In 2005, Buttazzo [13] performed a detailed study of FP and EDF scheduling. This work covered both schedulability under a variety of scenarios, in addition to practical implementation considerations. Results showed that the FP scheduling algorithm introduces more pre-emptions than EDF, especially at high processor utilisation levels. This leads to FP performing worse than EDF. Yet, FP has an advantage over EDF, in that it is generally simpler to implement in commercial kernels which do not provide explicit support for timing constraints. Despite being a very detailed study, these comparisons were done under the assumption that there were no pre-emption costs due to CRPD.

In this paper we build on the work by Buttazzo [13] and use state of the art CRPD analysis for FP [3] and EDF [22] to perform a comprehensive study of these two popular scheduling algorithms when accounting for CRPD.

1.1 Related Work on CRPD

Analysis of CRPD uses the concept of useful cache blocks (UCBs) and evicting cache blocks (ECBs) based on the work by Lee et al. [18]. Any memory block that is accessed by a task while executing is classified as an ECB, as accessing that block may evict a cache block of a pre-empted task. Out of the set of ECBs, some of them may also be UCBs. A memory block \( m \) is classified as a UCB at program point \( \mathcal{P} \), if (i) \( m \) may be cached at \( \mathcal{P} \) and (ii) \( m \) may be reused at program point \( \mathcal{Q} \) that may be reached from \( \mathcal{P} \) without eviction of \( m \) on this path. In the case of a pre-emption at program point \( \mathcal{P} \), only the memory blocks that are (i) in cache and (ii) will be reused, may cause additional reloads. The maximum possible pre-emption cost for a task is determined by the program point with the highest number of UCBs. For each subsequent pre-emption, the program point with the next smallest number of UCBs can be considered. Altmeyer and Burguérié [1] presented a tighter definition of UCBs however, we only need the basic concept for this paper.

Depending on the approach used, the CRPD analysis combines the UCBs belonging to the pre-empted task(s) with the ECBs of the pre-empting task(s). Using this information, the total number of blocks that are evicted, which must then be reloaded after the pre-emption, can be calculated and combined with the cost of reloading a block to then give the CRPD.

A number of approaches have been developed for calculating the CRPD when using FP pre-emptive scheduling. They include Lee et al. [18] UCB-Only approach, which considers just the pre-empted task(s), and Busquets et al. [12] ECB-Only approach which considers just the pre-empting task. Approaches that consider the pre-empted and pre-empting task(s) include Tan and Mooney [26] UCB-Union approach, Altmeyer et al. [2] ECB-Union approach, and an alternative
approach by Staschulat et al. [25]. Finally, there are advanced multiset based approaches that consider the pre-empted and pre-empting task(s) by Altmeyer et al. [3], ECB-Union Multiset, UCB-Union Multiset, and a combined multiset approach.

There has been less work towards developing CRPD analysis for EDF pre-emptive scheduling. In 2007, Ju et al. [17] considered the intersection of the pre-empted task’s UCBs with the pre-empting task’s ECBs. However, this method for handling nested pre-emptions can lead to significant pessimism as each pair of tasks is considered separately. In 2013, Lunniss et al. [22] adapted a number of approaches for calculating CRPD for FP to work with EDF. Including the ECB-Only, UCB-Only, UCB-Union, ECB-Union, ECB-Union Multiset, UCB-Union Multiset and combined multiset CRPD analysis for FP given by Busquets et al. [12], Lee et al. [18], Tan and Mooney [26], and Altmeyer et al. [2, 3].

A different methodology was used by Bastoni et al. [8]. Instead of focusing on how to calculate an upper bound on the CRPD, they used measurements on real hardware to estimate a lower bound on the CRPD and cache related migration delays for data caches in a multi-processor system.

CRPD can have a significant effect on schedulability, and can also vary dramatically depending on a number of factors. In particular, the CRPD is highly dependent on how tasks are placed in cache. As the layout of tasks in memory determines how they are positioned in cache, choosing a sensible layout can have a big impact on the CRPD caused due to pre-emptions. In 2012, Lunniss et al. [21] presented an approach that uses a Simulated Annealing algorithm to optimise the layout of tasks to increase system schedulability when using FP scheduling.

1.2 Organisation

The paper is organised as follows. Section 2 introduces the system model, terminology and notation used. Existing schedulability tests and CRPD analysis are outlined in Section 3 for FP scheduling, and in Section 4 for EDF scheduling. Section 5 briefly covers optimising task layout to reduce CRPD. Section 6 compares FP and EDF with CRPD analysis using a set of case studies. In Section 7, we investigate the effect of a variety of configuration parameters in a series of evaluations using synthetic tasksets. Finally, we conclude in Section 8.

2 System Model, Terminology and Notation

This section describes the system model, terminology, and notation used in the rest of the paper.

We assume a single processor system, running a statically defined taskset under either pre-emptive FP or pre-emptive EDF scheduling. The system comprises a taskset $\Gamma$ made up of a fixed number of tasks $\{\tau_1, \ldots, \tau_n\}$ where $n$ is a positive integer. In the case of FP, each task has a unique fixed priority and the priority of task $\tau_i$ is $i$, where a priority of 1 is the highest and $n$ is the lowest. Each task, $\tau_i$ may produce a potentially infinite stream of jobs that are separated by a minimum inter-arrival time or period $T_i$. Each task has a relative deadline $D_i$, and each job of a task has an absolute deadline $d_i$ which is $D_i$ after it is released. In the case of EDF, each task has a unique task index ordered by relative deadline from smallest to largest. In the case of a tie when assigning the unique task indices, an arbitrary choice is made. Each task also has a worst case execution time $C_i$ (determined for non-pre-emptive execution). In this paper, we consider tasks with constrained deadlines. (Task deadlines may be referred to as constrained deadlines, i.e. $D_i \leq T_i$ or implicit i.e. $D_i = T_i$). We assume a discrete time model. We define $T_{\max}$ as the largest period of any task in the taskset, and similarly $D_{\max}$ as the largest relative deadline of any task in the taskset. Each task has a utilisation $U_i$, where $U_i = C_i / T_i$, and each taskset has a utilisation $U$ which is equal to the sum of its tasks’ utilisations.
A taskset is said to be schedulable with respect to a scheduling algorithm if all valid sequences of jobs generated by the taskset can be scheduled by the algorithm without any missed deadlines. A taskset is feasible if there exists some scheduling algorithm that can schedule all possible sequences of jobs that may be generated by the taskset without any missed deadlines. A scheduling algorithm is said to be optimal with respect to a task model if it can schedule all of the feasible tasksets that comply with the task model.

Each task $\tau_i$ has a set of UCBs, $\text{UCB}_i$ and a set of ECBs, $\text{ECB}_i$ represented by a set of integers. If for example, task $\tau_1$ contains 4 ECBs, where the second and fourth ECBs are also UCBs, these can be represented using $\text{ECB}_1 = \{1, 2, 3, 4\}$ and $\text{UCB}_1 = \{2, 4\}$. The block reload time (BRT) is the time taken to load a block from memory into cache. This cost is incurred every time that a UCB has to be reloaded after a pre-emption. We assume that the remaining context switch costs, i.e., pipeline and scheduler related costs are subsumed in the execution time bound of each task. Furthermore, we assume that the OS resides in a different cache partition and therefore scheduler operations do not cause CRPD.

We use the term cache utilisation to describe the ratio of the total size of the tasks to the size of the cache. A cache utilisation of 1 means that the tasks fit exactly in the cache, whereas a cache utilisation of 5 means the total size of the tasks is 5 times the size of the cache.

We focus on instruction only caches. In the case of data caches, the analysis would either require a write-through cache or further extension in order to be applied to write-back caches. We assume that tasks do not share any code. We also assume a direct mapped cache, but the work extends to set-associative caches with the LRU replacement policy. In the case of set-associative LRU caches, a single cache-set may contain several UCBs. For example, $\text{UCB}_1 = \{2, 2, 4\}$ means that task $\tau_1$ has two UCBs in cache-set 2 and one UCB in cache set 4. As one ECB suffices to evict all UCBs of the same cache-set, multiple accesses to the same set by the pre-empting task do not appear in the set of ECBs. A bound on the CRPD in the case of LRU caches due to task $\tau_j$ directly pre-empting $\tau_i$ is thus given by the intersection $\text{UCB}_i \cap \text{ECB}_j = \{m | m \in \text{UCB}_i : m \in \text{ECB}_j\}$, where the result is a multiset that contains each element from $\text{UCB}_i$ if it is also in $\text{ECB}_j$. A precise computation of CRPD in the case of LRU caches is given in Altmeyer et al. [4]. The equations provided in this paper can be applied to set-associative LRU caches with the above adaptation to the set-intersection.

### 3 CRPD Analysis for FP Scheduling

In this section, we give an overview of FP scheduling and schedulability analysis for it. We then cover the state of the art CRPD analysis for FP scheduling, by Altmeyer et al. described in detail in [3].

Under FP scheduling, the sets of tasks that can pre-empt each other are based on the statically assigned fixed task priorities. Using the fixed priorities, we can define the following sets of tasks for determining which tasks can pre-empt each other. $\text{hp}(i)$ and $\text{lp}(i)$ are the sets of tasks with higher and lower priorities than task $\tau_i$, and $\text{hep}(i)$ and $\text{lep}(i)$ are the sets containing tasks with higher or equal and lower or equal priorities to task $\tau_i$. Additionally, $\text{aff}(i, j) = \text{hep}(i) \cap \text{lp}(j)$ is used to represent all tasks that can have CRPD caused by task $\tau_j$ pre-empting them, which affects the response time of task $\tau_i$. In other words, it is the set of tasks that may be pre-empted by task $\tau_j$ and have at least the priority of task $\tau_i$.

---

1 The concept of UCBs and ECBs cannot be applied to the FIFO or PLRU replacement policies as shown by Burguière [11].
Schedulability tests are used to determine if a taskset is schedulable, i.e. all the tasks will meet their deadlines given the worst-case pattern of arrivals and execution. For a given taskset, the response time \( R_i \) for each task \( \tau_i \), can be calculated and compared against the tasks’ deadline, \( D_i \). If every task in the taskset meets its deadline, then the taskset is schedulable. The equation used to calculate \( R_i \) is defined as [5]:

\[
R_{i}^{\alpha+1} = C_i + \sum_{\forall j \in hp(i)} \left[ \frac{R_i}{T_j} \right] (C_j).
\]

Equation (1) can be solved using fixed point iteration. Iteration continues until either \( R_{i}^{\alpha+1} > D_i \) in which case the task is unschedulable, or until \( R_{i}^{\alpha+1} = R_i^{\alpha} \) in which case the task is schedulable and has a worst-case response time of \( R_i^{\alpha} \).

Note the convergence of (1) may be speeded up using the techniques described in [14].

### 3.1 CRPD Analysis

To account for the CRPD, a component \( \gamma_{i,j} \) is introduced into (1). There are a number of different methods that can be used to compute \( \gamma_{i,j} \) described by Altmeyer et al. in [3]. Depending on the method used, \( \gamma_{i,j} \) represents either a single pre-emption, or multiple pre-emptions and is calculated using the cost incurred when reloading a block, the block reload time (BRT), multiplied by the number of blocks which may need to be reloaded after each pre-emption.

We will now summarise the combined multiset approach, which has been shown to dominate all other approaches [3]. For worked examples of the analysis, see Section 4 ECB-Union and Multiset Approaches of Altmeyer et al. [3].

In the combined multiset approach, \( \gamma_{i,j} \) represents the total cost of all pre-emptions due to jobs of task \( \tau_j \) executing within the response time of task \( \tau_i \). Incorporating \( \gamma_{i,j} \) into (1) gives a revised equation for \( R_i \):

\[
R_{i}^{\alpha+1} = C_i + \sum_{\forall j \in hp(i)} \left( \left[ \frac{R_i}{T_j} \right] C_j + \gamma_{i,j} \right).
\]

\( \gamma_{i,j} \) is then calculated using two separate approaches, the UCB-Union multiset, and ECB-Union multiset which are described below. The key concept behind them is to calculate the cost of each individual pre-emption by jobs of task \( \tau_j \) that could occur within the response time of task \( \tau_i \). By calculating the cost of each pre-emption, the analysis is able to account for the fact that intermediate tasks in a nested pre-emption will often be pre-empted less than the lowest priority task. Consider the following example with three tasks shown in Figure 1.

In the example, the total cost of all jobs of task \( \tau_1 \) pre-empting task \( \tau_3 \) within the response time of task \( \tau_3 \) is equal to the cost of task \( \tau_1 \) pre-empting task \( \tau_2 \) and task \( \tau_3 \) once (nested pre-emption), and task \( \tau_3 \) on its own twice. It is therefore important to recognise that the cost of one task pre-empting another is highly dependent on any intermediate tasks that may be involved in a nested pre-emption. To calculate the number of pre-emptions, we use \( E_j(R_i) \) to denote the
maximum number of jobs of task $\tau_j$ that can execute during the response time, $R_i$, of task $\tau_i$. For our model, $E_j(R_i) = \lceil R_i/T_j \rceil$.

### 3.1.1 ECB-Union Multiset

The ECB-Union multiset approach computes the union of all ECBs that may affect a pre-empted task during a pre-emption by task $\tau_j$. Specifically, it accounts for nested pre-emptions by assuming that task $\tau_j$ has already been pre-empted by all tasks of a higher priority.

The first step is to calculate the number of UCBs that task $\tau_j$ could evict when pre-empting an intermediate task, $\tau_k$. This is given by calculating the intersection of the UCBs of the pre-empted task, task $\tau_k$, with the set of ECBs belonging to the pre-empting tasks $\bigcup_{h \in \text{hp}(j) \cup \{j\}} \text{ECB}_h$ to give:

$$\text{UCB}_k \cap \left( \bigcup_{h \in \text{hp}(j) \cup \{j\}} \text{ECB}_h \right).$$

(3)

Note $h \in \text{hp}(j) \cup \{j\}$ is used to account for the case when tasks can share priorities.

The ECB-Union multiset approach recognises that task $\tau_j$ cannot pre-empt each intermediate task $\tau_k$ more than $E_j(R_k)E_k(R_i)$ times during the response time of task $\tau_i$. Therefore, the next step is to form a multiset $M_{i,j}$ that contains the cost of task $\tau_j$ pre-empting task $\tau_k$ $(3)$ repeated $E_j(R_k)E_k(R_i)$ times, for each task $\tau_k \in \text{aff}(i,j)$ hence:

$$M_{i,j} = \bigcup_{k \in \text{aff}(i,j)} \left( \bigcup_{E_j(R_k)E_k(R_i)} \left( \text{UCB}_k \cap \left( \bigcup_{h \in \text{hp}(j) \cup \{j\}} \text{ECB}_h \right) \right) \right).$$

(4)

As only $E_j(R_i)$ jobs of task $\tau_j$ can execute during the response time of task $\tau_i$, the maximum CRPD is obtained by summing the $E_j(R_i)$ largest pre-emptions, i.e. the $E_j(R_i)$ largest values in $M_{i,j}$:

$$\gamma_{i,j}^{\text{ecb-m}} = \text{BRT} \cdot \sum_{l=1}^{E_j(R_i)} M_{i,j}^l,$$

(5)

where $M_{i,j}^l$ is the $l^{th}$ largest integer value from the multiset $M_{i,j}$.

### 3.1.2 UCB-Union Multiset

The UCB-Union multiset approach accounts for the effects of nested pre-emptions by assuming that the UCBs of any tasks that could be pre-empted, including nested pre-emptions, by task $\tau_j$ are evicted by the ECBs of task $\tau_j$. The first step is to form a multiset $M_{i,j}^{\text{ucb}}$ containing $E_j(R_k)E_k(R_i)$ copies of the UCBs of each task $\tau_k \in \text{aff}(i,j)$ that could be pre-empted by task $\tau_j$ and has at least the priority of task $\tau_i$. This multiset reflects the fact that jobs of task $\tau_j$ cannot evict the UCBs of jobs of task $\tau_k$ within the response time of task $\tau_i$ more than $E_j(R_k)E_k(R_i)$ times. Hence:

$$M_{i,j}^{\text{ucb}} = \bigcup_{k \in \text{aff}(i,j)} \left( \bigcup_{E_j(R_k)E_k(R_i)} \text{UCB}_k \right).$$

(6)

The second step is to form a separate multiset $M_{i,j}^{\text{ecb}}$ containing $E_j(R_i)$ copies of the ECBs of task $\tau_j$. This multiset reflects the fact that there can be no more than $E_j(R_i)$ jobs of task $\tau_j$
within the response time of task $\tau_i$, each of which can evict cache blocks in the set ECB$_j$:

$$M_{i,j}^{\text{ecb}} = \bigcup_{E_j(R_i)} (\text{ECB}_j).$$  \hspace{1cm} (7)

$\gamma_{i,j}^{\text{ucb-m}}$ is then given by the size of the multiset intersection between $M_{i,j}^{\text{ucb}}$ and $M_{i,j}^{\text{ecb}}$:

$$\gamma_{i,j}^{\text{ucb-m}} = BRT \cdot \lvert M_{i,j}^{\text{ucb}} \cap M_{i,j}^{\text{ecb}} \rvert.$$  \hspace{1cm} (8)

### 3.1.3 Combined Multiset

The ECB-Union multiset and UCB-Union multiset approaches are incomparable, meaning that each approach can find different sets of tasksets schedulable. Because of this property, they can be combined together to form a combined approach:

$$R_i = \min (R_i^{\text{ucb-m}}, R_i^{\text{ecb-m}}).$$  \hspace{1cm} (9)

The response time for every task is calculated using each approach and then the minimum is taken, because of this, the combined approach can deem some tasksets schedulable that are not schedulable by either approach on its own.

### 3.2 Comparison of Approaches

Figure 2 shows a Venn diagram that conveys the relationship between a number of different approaches for calculating CRPD under FP scheduling [3]. However, it does not include the method by Staschulat et al. [25] because it is incomparable to them. Specifically, while it typically deems a lower number of tasksets schedulable, it could potentially find a taskset schedulable that is not schedulable by any of the other approaches. Aside from the approach by Staschulat et al. [25], it can be seen that the combined multiset approach dominates all other approaches. See Altmeyer et al. [3] for a detailed comparison between each approach.

### 4 CRPD Analysis for EDF Scheduling

In this section, we give an overview of EDF scheduling and schedulability analysis for it. We then cover the state of the art CRPD analysis for EDF scheduling, by Lunniss et al. [22].
EDF is a dynamic scheduling algorithm which always schedules the job with the earliest absolute deadline first. In pre-emptive EDF, any time a job arrives with an earlier absolute deadline than the current running job, it will pre-empt the current job. When a job completes its execution, the EDF scheduler chooses the pending job with the earliest absolute deadline to execute next.

In 1973, Liu and Layland [20] gave a necessary and sufficient schedulability test that indicates whether a taskset is schedulable under EDF iff \( U \leq 1 \), under the assumption that all tasks have implicit deadlines \( (D_i = T_i) \). In the case where \( D_i \leq T_i \) this test is still necessary, but is no longer sufficient.

In 1974, Dertouzos [15] proved EDF to be optimal among all scheduling algorithms on a uniprocessor, in the sense that if a taskset cannot be scheduled by pre-emptive EDF, then this taskset cannot be scheduled by any algorithm.

In 1980, Leung and Merrill [19] showed that a set of periodic tasks is schedulable under EDF iff all absolute deadlines in the interval \([0, \max\{s_i\} + 2H]\) are met, where \( s_i \) is the start time of task \( \tau_i \), \( \min\{s_i\} = 0 \), and \( H \) is the hyperperiod (least common multiple) of all tasks’ periods.

In 1990 Baruah et al. [6, 7] extended Leung and Merrill’s work [19] to sporadic tasksets. They introduced \( h(t) \), the processor demand function, which denotes the maximum execution time requirement of all tasks’ jobs which have both their arrival times and their deadlines in a contiguous interval of length \( t \). Using this they showed that a taskset is schedulable iff \( \forall t > 0, h(t) \leq t \) where \( h(t) \) is defined as:

\[
h(t) = \sum_{i=1}^{n} \max \left\{ 0, 1 + \left[ \frac{t-D_i}{T_i} \right] \right\} C_i. \tag{10}
\]

Examining (10), it can be seen that \( h(t) \) can only change when \( t \) is equal to an absolute deadline, which restricts the number of values of \( t \) that need to be checked. In order to place an upper bound on \( t \), and therefore the number of calculations of \( h(t) \), the minimum interval in which it can be guaranteed that an unschedulable taskset will be shown to be unschedulable must be found. For a general taskset with arbitrary deadlines \( t \) can be bounded by \( L_a \) [16]:

\[
L_a = \max \left\{ D_1, \ldots, D_n, \frac{\sum_{i=1}^{n}(T_i-D_i)U_i}{1-U} \right\}. \tag{11}
\]

Spuri [24] and Ripoll et al. [23] showed that an alternative bound \( L_b \), given by the length of the synchronous busy period can be used. Where \( L_b \) is computed by solving the following equation using fixed point iteration:

\[
w_{\alpha+1} = \sum_{i=1}^{n} \left[ \frac{w_n}{T_i} \right] C_i. \tag{12}
\]

There is no direct relationship between \( L_a \) and \( L_b \), which enables \( t \) to be bounded by \( L = \min(L_a, L_b) \). Combined with the knowledge that \( h(t) \) can only change at an absolute deadline, a taskset is therefore schedulable under EDF iff \( U \leq 1 \) and:

\[
\forall t \in Q : h(t) \leq t, \tag{13}
\]

where \( Q \) is defined as:

\[
Q = \{ d_k \mid d_k = kT_i + D_i \land d_k < \min(L_a, L_b), k \in \mathbb{N} \}. \tag{14}
\]

In 2009, Zhang and Burns [27] presented their Quick convergence Processor-demand Analysis (QPA) algorithm which exploits the monotonicity of \( h(t) \). QPA determines schedulability by starting with a value of \( t \) that is close to \( L \), and then iterating back towards 0 checking a significantly smaller number of values of \( t \) than would otherwise be required.
Figure 3 Example schedule showing how the scheduler chooses which task should execute. Task $\tau_3$ is released at $t = 0$. At $t = 5$, task $\tau_2$ is released, pre-empting $\tau_3$ as although it has the same absolute deadline, it has a lower task index. At $t = 6$, task $\tau_1$ is released, pre-empting task $\tau_2$. At $t = 7$, $\tau_1$ completes, the scheduler then chooses to resume task $\tau_2$ as although it has the same absolute deadline as task $\tau_3$, it has the lower task index.

4.1 CRPD Analysis

Due to the undefined behaviour of EDF when two or more jobs have the same absolute deadline, an assumption needs to be made before we can tightly calculate CRPD for EDF. In the case where two or more jobs have the same absolute deadline, Lunniss et al. [22] assume the scheduler always picks the job belonging to the task with the lowest unique task index, see Figure 3. This has the benefit of minimising the number of pre-emptions. In the case where jobs of two tasks have the same absolute and relative deadlines, it ensures that they cannot pre-empt each other. Furthermore, it ensures that after a pre-emption, the task that was pre-empted last is resumed first.

Following the analysis of Lunniss et al. [22], we now define a number of terms with respect to EDF scheduling. Some of the terms are also present in the analysis for FP, but have slightly different meanings under EDF. Assuming any task $\tau_j$ with a relative deadline $D_j < D_i$ can pre-empt task $\tau_i$, the set of tasks that may have a higher priority, and can pre-empt task $\tau_i$, under EDF is:

$$hp(i) = \{ \tau_j \in \Gamma \mid D_j < D_i \}.$$  \hspace{1cm} (15)

The set of tasks that can be pre-empted by jobs of task $\tau_j$ in an interval of length $t$, $aff(t,j)$ is based on the relative deadlines of the tasks. It captures all of the tasks whose relative deadlines are greater than the relative deadline of task $\tau_j$ excluding tasks whose deadlines are larger than $t$ as they do not need to be included when calculating $h(t)$. This gives:

$$aff(t,i) = \{ \tau_j \in \Gamma \mid t \geq D_i > D_j \}.$$  \hspace{1cm} (16)

To determine how many pre-emptions can occur, we use $P_j(D_i)$ to denote the maximum number of jobs of task $\tau_j$ that can pre-empt a single job of task $\tau_i$:

$$P_j(D_i) = \max \left(0, \left\lceil \frac{D_i - D_j}{T_j} \right\rceil \right).$$  \hspace{1cm} (17)

Finally, we also use $E_j(t)$ to denote the maximum number of jobs of task $\tau_j$ that can have both their release times and their deadlines in an interval of length $t$:

$$E_j(t) = \max \left(0, \left\lfloor \frac{t - D_j}{T_j} \right\rfloor \right).$$  \hspace{1cm} (18)

We now summarise CRPD analysis for EDF by Lunniss et al. [22] using the combined multiset approach as it has been shown to dominate all other approaches for calculating CRPD for EDF. This approach is based on the combined multiset approach for FP as described in Section 3.1, and as such the equations and the intuition behind them are similar. The difference is to do with...
A Comparison between Fixed Priority and EDF Scheduling accounting for CRPD

which tasks pre-empt each other and the timeframe used to determine which jobs to include in the calculation.

CRPD analysis can be integrated into the EDF schedulability test by introducing an additional parameter, \( \gamma_{t,j} \) to represent the CRPD. In this case, \( \gamma_{t,j} \) represents the cost of the maximum number \( E_j(t) \) of pre-emptions by jobs of task \( \tau_j \) that have their release times and absolute deadlines in an interval of length \( t \). It is therefore included in (10) as follows:

\[
h(t) = \sum_{j=1}^{n} \left( \max \left\{ 0, \left\lfloor \frac{t - D_j}{T_j} \right\rfloor \right\} C_j + \gamma_{t,j} \right).
\] (19)

\( \gamma_{t,j} \) can then be calculated using two different methods and the lowest value of the two used to calculate the processor demand. These methods calculate the cost of each possible individual pre-emption by task \( \tau_j \) that could occur during an interval of length \( t \).

4.1.1 ECB-Union Multiset

The ECB-Union multiset approach computes the union of all ECBs that may affect a pre-empted task during a pre-emption by task \( \tau_j \). Specifically, it accounts for nested pre-emptions by assuming that task \( \tau_j \) has already been pre-empted by all other tasks that may pre-empt it. The first step is to form a multiset \( M_{t,j} \) that contains the cost:

\[
\left| \text{UCB}_k \cap \left( \bigcup_{h \in \text{hp}(j) \cup \{j\}} \text{ECB}_h \right) \right|
\] (20)

of task \( \tau_j \) pre-empting task \( \tau_k \) repeated \( P_j(D_k)E_k(t) \) times, for each task \( \tau_k \in \text{aff}(t,j) \). Hence:

\[
M_{t,j} = \bigcup_{k \in \text{aff}(t,j)} \left( \bigcup_{P_j(D_k)E_k(t)} \left| \text{UCB}_k \cap \left( \bigcup_{h \in \text{hp}(j) \cup \{j\}} \text{ECB}_h \right) \right| \right).
\] (21)

As there are only \( E_j(t) \) jobs of task \( \tau_j \) with release times and deadlines in an interval of length \( t \), the maximum CRPD is obtained by summing the \( E_j(t) \) largest values in \( M_{t,j} \):

\[
\gamma_{t,j}^{\text{ecb-m}} = E_j(t) \cdot \sum_{i=1}^{E_j(t)} M_{t,j}^i,
\] (22)

where \( M_{t,j}^i \) is the \( i \)th largest integer value from the multiset \( M_{t,j} \).

4.1.2 UCB-Union Multiset Approach

The UCB-Union multiset approach accounts for the effects of nested pre-emptions by assuming that the UCBs of any tasks that could be pre-empted, including nested pre-emptions, by task \( \tau_j \) are evicted by the ECBs of task \( \tau_j \). The first step is to form a multiset \( M_{t,j}^{\text{ucb}} \) containing \( P_j(D_k)E_k(t) \) copies of the UCBs of each task \( \tau_k \in \text{aff}(t,j) \). This multiset reflects the fact that jobs of task \( \tau_j \) cannot evict the UCBs of jobs of task \( \tau_k \) that have both their release times and deadlines in an interval of length \( t \) more than \( P_j(D_k)E_k(t) \) times. Hence:

\[
M_{t,j}^{\text{ucb}} = \bigcup_{k \in \text{aff}(t,j)} \left( \bigcup_{P_j(D_k)E_k(t)} \text{UCB}_k \right).
\] (23)
The second step is to form a separate multiset $M_{\text{ecb}}$ containing $E_j(t)$ copies of the ECB$_j$ of task $\tau_j$. This multiset reflects the fact that there are at most $E_j(t)$ jobs of task $\tau_j$ that have their release times and deadlines in an interval of length $t$, each of which can evict cache blocks in the set ECB$_j$:

$$M_{\text{ecb}} = \bigcup_{E_j(t)} \text{(ECB$_j$)}.$$  \hspace{1cm} (24)

$\gamma_{ucb-m}$ is then given by the size of the multi-set intersection between $M_{\text{ucb}}$ and $M_{\text{ecb}}$:

$$\gamma_{ucb-m} = BRT \cdot \left| M_{\text{ucb}} \cap M_{\text{ecb}} \right|.$$  \hspace{1cm} (25)

### 4.1.3 Combined Multiset Approach

The ECB-Union Multiset and UCB-Union Multiset approaches provide upper bounds that are incomparable, therefore, $h(t)$ can be calculated at each stage of the QPA algorithm using both approaches and the minimum value taken to form a combined approach:

$$h(t) = \min \left( h(t)^{ucb-m}, h(t)^{ecb-m} \right).$$  \hspace{1cm} (26)

As the processor demand is calculated using each approach, for each interval $t$, the combined approach can deem some tasksets schedulable that are not schedulable by either approach on its own.

### 4.1.4 Effect on Task Utilisation and $h(t)$ Calculation

As the multiset approaches effectively inflate the execution time of task $\tau_j$ by the CRPD that it can cause in an interval of length $t$, the upper bound $L$, used for calculating the processor demand $h(t)$, must be adjusted. This is achieved by calculating an upper bound on the utilisation due to CRPD that is valid for all intervals of length greater than some value $L_c$. This CRPD utilisation value is then used to inflate the taskset utilisation and thus compute an upper bound on the maximum length of the synchronous busy period. This upper bound is valid provided that it is greater than $L_c$, otherwise the actual maximum length of the busy period may lie somewhere in the interval $[L_d, L_c]$, hence we can use $\max(L_c, L_d)$ as a bound.

The first step is to assign $t = L_c = 100T_{\max}$ which limits the overestimation of the CRPD utilisation $U^\gamma = \gamma_t/t$ to at most 1%. Next, $\gamma_t$ is calculated using (22) for ECB-Union Multiset and (25) for UCB-Union Multiset. However, in (21) and (23) & (24), $E_{\text{max}}(t)$ is substituted for $E_x(t)$ to ensure that the computed value of $U^\gamma$ is a valid upper bound for all intervals of length $t \geq L_c$:

$$E_{\text{max}}(t) = \max \left( 0, 1 + \left\lceil \frac{t - D_x}{T_x} \right\rceil \right).$$  \hspace{1cm} (27)

If $U + U^\gamma \geq 1$, then the taskset is deemed unschedulable, otherwise an upper bound on the length of the busy period can be computed via a modified version of (12):

$$w^{\alpha+1} \leq \sum_{\forall j} \left( \frac{w^{\alpha}}{T_j} + 1 \right) C_j + w^{\alpha}U^\gamma$$  \hspace{1cm} (28)

rearranged to give:

$$w \leq \frac{1}{(1 - (U + U^\gamma))} \sum_{\forall j} U_j T_j.$$  \hspace{1cm} (29)
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Figure 4 Venn diagram showing the relationship between different approaches for CPRD analysis under EDF scheduling.

Then, substituting in $T_{\text{max}}$ for each value of $T_j$ the upper bound is given by:

$$L_d = \frac{U \cdot T_{\text{max}}}{(1 - (U + U^r))}.$$  \hfill (30)

Finally, $L = \max(L_c, L_d)$ can then be used as the maximum value of $t$ to check in the EDF schedulability test.

4.2 Comparison of Approaches

Figure 4 shows a Venn diagram that conveys the relationship between the different approaches for calculating CRPD under EDF scheduling [22]. Note that JCR represents the approach of Ju et al. [17]. It can be seen that the combined multiset approach dominates all other approaches, see Lunniss et al. [22] for detailed comparisons of each approach.

5 Task Layout

The layout of tasks in memory determines how they are positioned in cache, which then affects the CRPD that occurs during pre-emption. Figure 5 shows an example layout of five tasks in cache. If scheduled under FP, task $\tau_1$ has the highest priority, so its UCBs can never be evicted as it cannot be pre-empted. Task $\tau_2$ and $\tau_3$’s UCBs are safe from eviction as they are not mapped to the same location in cache as higher priority task’s ECBs. However, task $\tau_4$’s UCBs could be evicted by task $\tau_1$, and $\tau_5$’s UCBs could be evicted by task $\tau_1$, $\tau_2$ or $\tau_4$. This layout could be improved by shifting task $\tau_5$ so that its UCBs can only be evicted by task $\tau_3$.

In 2012, Lunniss et al. [21] presented an approach that uses Simulated Annealing to optimise the layout of tasks to increase system schedulability. It does so by changing the order of tasks in memory, which can be implemented in practice by presenting the tasks to the linker in the desired order. The approach is driven by the schedulability of the taskset, favouring layouts that allow the taskset to be scheduled at a higher utilisation. While this approach was originally used for FP scheduling, it can also be applied to the EDF scheduling algorithm by switching the schedulability test used. We therefore use this approach to optimise the layout of the tasksets to make each scheduling algorithm as competitive as possible.
6 Case Studies

In this section we compare the different approaches for calculating CRPD using a set of case studies based on PapaBench\textsuperscript{2}, the Mälardalen\textsuperscript{3} benchmark suite and a set of SCADE\textsuperscript{4} tasks (partially provided by SCADE, partially from our own SCADE models). In all cases the system was set up to model an ARMv7\textsuperscript{5} processor clocked at 100 MHz with a 2 KB direct-mapped instruction cache and a line size of 8 Bytes, giving 256 cache sets, 4 Byte instructions, and a BRT of 8 $\mu$s.

6.1 Single Taskset Case Study

PapaBench is a real-time embedded benchmark based on the software of a GNU-license UAV, called Paparazzi. PapaBench contains two sets of tasks, fly-by-wire and autopilot. In this paper we used the autopilot tasks, for which the WCETs, periods, UCBs, and ECBs were collected using aiT\textsuperscript{6} – see Table 1. We made the following assumptions in our evaluation to handle the interrupt tasks:

- Interrupts have higher priority than the normal tasks, but they cannot pre-empt each other
- Interrupts can occur at any time
- All interrupts have the same deadline which must be greater than or equal to the sum of their execution times in order for them to be schedulable
- The cache is disabled whenever an interrupt is executing and enabled again after it completes

In the case of FP scheduling, the interrupts can be modelled as normal tasks with no UCBs or ECBs. Due to the interrupts having the same deadline which is large enough to accommodate the interrupts execution times, no other changes need to be made to the analysis. For EDF scheduling, a number of adjustments must be made to correctly account for the interrupts not being able to pre-empt each other. First we modify equation (19) to exclude interrupts when calculating the processor demand, $h(t)$. We then calculate the execution time of each interrupt in the interval $t$ using equation (2) of \cite{10}. The result of which is then added onto the result of the modified version

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\textsuperscript{2} http://www.irit.fr/recherches/ARCHI/MARCH/rubrique.php3?id_rubrique=97
\textsuperscript{3} http://www.mrtc.mdh.se/projects/wcet/benchmarks.html
\textsuperscript{4} Esterel SCADE http://www.esterel-technologies.com/
\textsuperscript{5} http://www.arm.com/products/processors/cortex-m/cortex-m3.php
\textsuperscript{6} AbsInt http://www.absint.com/ait/
of (19), giving the processor demand for both tasks and interrupts. We then adjust the upper bound $L$ used when checking $h(t)$. This is implemented by substituting $U = U_{\text{tasks}} + U_{\text{interrupts}}$ into equation (30). Note that we leave $U$ to represent the utilisation of the CRPD caused by just tasks as we assume that the cache is disabled while the interrupts are executing and as such they cannot cause any CRPD.

We assigned a deadline of 2 ms to all of the interrupt tasks, and implicit deadlines i.e. $D_i = T_i$, to the normal tasks. We then calculated the total utilisation for the system and then effectively scaled the clock speed in order to reduce the total utilisation to the target utilisation for the system. We used the number of UCBs and ECBs obtained via analysis, placing the UCBs in a group at a random location in each task.

In each evaluation, the taskset utilization not including pre-emption cost was varied from 0.025 to 1 in steps of 0.001. For each utilization value, the schedulability of the taskset was determined under both FP and EDF. Specifically, we compared each scheduling algorithm (i) assuming no pre-emption cost, (ii) using CRPD analysis using the standard task layout, and (iii) using CRPD analysis after optimising the task layout using Simulated Annealing as described in [21]. The standard task layout is obtained by ordering tasks sequentially in memory based on their unique task indices.

Table 2 shows the breakdown utilisation for the single taskset based on PapaBench. There are a few interesting points to note. Firstly the breakdown utilisation is very high for both FP and EDF, this is due to the nearly harmonic periods and small range of task periods, with EDF outperforming FP. Secondly, the CRPD is very low when scheduled using either FP or EDF due to the small number of UCBs. As the CRPD is very low, the layout optimisation makes little to no difference.

6.2 Multiple Taskset Case Studies

The single taskset case study provides one specific example based on the PapaBench tasks and their periods. The remaining case studies used tasksets generated by randomly selecting tasks from a set of benchmarks. In the case of the PapaBench tasks, we treated the interrupts as normal tasks. We obtained tasksets by randomly selecting 10 tasks from Table 1 (PapaBench benchmarks), or 10 tasks from Table 3 (Mälardalen and SCADe benchmarks) or 15 tasks from the two tables (mixed benchmarks). Using the UUnifast algorithm [9], we calculated the utilisation, $U_i$ of each task so that the utilisations added up to the desired utilisation level for the taskset. Based on the target utilisation and task execution times, $T_i$ was calculated such that $C_i = U_i T_i$. We used $D_i = y + x(T_i - y)$ to generate the constrained deadlines, where $x$ is a random number between 0 and 1, and $y = \max(T_i/2, 2C_i)$. This generates constrained deadlines that are no less than half the period of the tasks. Note, allowing deadlines to be as small as $C_i$ would result in tasks that were unschedulable once CRPD were introduced. We used the number of UCBs and ECBs obtained using aiT, placing the UCBs in a group at a random location in each task.

We generated 1000 tasksets for the multiple taskset case studies, and evaluated them using the same method as the single taskset case study, except that we varied the utilisation excluding pre-emption costs from 0.025 to 1 in steps of 0.0125.

6.2.1 PapaBench Benchmark

The tasks in the PapaBench benchmarks are simple, short control tasks with limited computations and data accesses. Figure 6 shows the percentage of tasksets that were deemed schedulable by each approach for the 1000 tasksets of cardinality 10 that we randomly selected from Table 1. The results are similar to those obtained using the single taskset PapaBench case study. Specifically,
Table 1 Execution times, periods and number of UCBs and ECBs for the tasks from PapaBench. (ms = milisecond)

<table>
<thead>
<tr>
<th>Task</th>
<th>Name</th>
<th>UCBs</th>
<th>ECBs</th>
<th>WCET</th>
<th>Period</th>
</tr>
</thead>
<tbody>
<tr>
<td>I4</td>
<td>interrupt_modem</td>
<td>2</td>
<td>10</td>
<td>0.303 ms</td>
<td>100 ms</td>
</tr>
<tr>
<td>I5</td>
<td>interrupt_spi_1</td>
<td>1</td>
<td>10</td>
<td>0.251 ms</td>
<td>50 ms</td>
</tr>
<tr>
<td>I6</td>
<td>interrupt_spi_2</td>
<td>1</td>
<td>4</td>
<td>0.151 ms</td>
<td>50 ms</td>
</tr>
<tr>
<td>I7</td>
<td>interrupt_gps</td>
<td>3</td>
<td>26</td>
<td>0.283 ms</td>
<td>250 ms</td>
</tr>
<tr>
<td>T5</td>
<td>altitude_control</td>
<td>20</td>
<td>66</td>
<td>1.478 ms</td>
<td>250 ms</td>
</tr>
<tr>
<td>T6</td>
<td>climb_control</td>
<td>1</td>
<td>210</td>
<td>5.429 ms</td>
<td>250 ms</td>
</tr>
<tr>
<td>T7</td>
<td>link_fbw_send</td>
<td>1</td>
<td>10</td>
<td>0.233 ms</td>
<td>50 ms</td>
</tr>
<tr>
<td>T8</td>
<td>navigation</td>
<td>10</td>
<td>256</td>
<td>4.432 ms</td>
<td>250 ms</td>
</tr>
<tr>
<td>T9</td>
<td>radio_control</td>
<td>0</td>
<td>256</td>
<td>15.681 ms</td>
<td>25 ms</td>
</tr>
<tr>
<td>T10</td>
<td>receive_gps_data</td>
<td>22</td>
<td>194</td>
<td>5.987 ms</td>
<td>250 ms</td>
</tr>
<tr>
<td>T11</td>
<td>reporting</td>
<td>2</td>
<td>256</td>
<td>12.222 ms</td>
<td>100 ms</td>
</tr>
<tr>
<td>T12</td>
<td>stabilization</td>
<td>11</td>
<td>194</td>
<td>5.681 ms</td>
<td>50 ms</td>
</tr>
</tbody>
</table>

Table 2 Breakdown utilisation under the different approaches for the single PapaBench taskset.

<table>
<thead>
<tr>
<th>Breakdown</th>
<th>Utilisation</th>
</tr>
</thead>
<tbody>
<tr>
<td>EDF – No Pre-emption Cost</td>
<td>0.999</td>
</tr>
<tr>
<td>FP – No Pre-emption Cost</td>
<td>0.977</td>
</tr>
<tr>
<td>EDF – Optimised Layout</td>
<td>0.985</td>
</tr>
<tr>
<td>EDF – Standard Layout</td>
<td>0.985</td>
</tr>
<tr>
<td>FP – Optimised Layout</td>
<td>0.970</td>
</tr>
<tr>
<td>FP – Standard Layout</td>
<td>0.969</td>
</tr>
</tbody>
</table>

Table 3 Execution times and number of UCBs and ECBs for the largest benchmarks from the Mälardalen Benchmark Suite (M), and SCADE Benchmarks (S). (s = second, ms = milisecond)

<table>
<thead>
<tr>
<th>Source</th>
<th>Description</th>
<th>UCBs</th>
<th>ECBs</th>
<th>WCET</th>
</tr>
</thead>
<tbody>
<tr>
<td>M</td>
<td>adpcm</td>
<td>24</td>
<td>226</td>
<td>5.541 s</td>
</tr>
<tr>
<td>M</td>
<td>compress</td>
<td>25</td>
<td>114</td>
<td>3.664 s</td>
</tr>
<tr>
<td>M</td>
<td>edn</td>
<td>56</td>
<td>98</td>
<td>244.9 ms</td>
</tr>
<tr>
<td>M</td>
<td>fir</td>
<td>28</td>
<td>50</td>
<td>21.53 ms</td>
</tr>
<tr>
<td>M</td>
<td>jfdctinit</td>
<td>40</td>
<td>162</td>
<td>62.53 ms</td>
</tr>
<tr>
<td>M</td>
<td>ns</td>
<td>17</td>
<td>26</td>
<td>73.38 ms</td>
</tr>
<tr>
<td>M</td>
<td>nsichneu</td>
<td>53</td>
<td>256</td>
<td>149.6 ms</td>
</tr>
<tr>
<td>M</td>
<td>statemate</td>
<td>3</td>
<td>256</td>
<td>77.96 ms</td>
</tr>
<tr>
<td>S</td>
<td>cruise control system</td>
<td>25</td>
<td>107</td>
<td>1.959 s</td>
</tr>
<tr>
<td>S</td>
<td>flight control system</td>
<td>70</td>
<td>256</td>
<td>2.138 s</td>
</tr>
<tr>
<td>S</td>
<td>navigation system</td>
<td>45</td>
<td>82</td>
<td>1.409 s</td>
</tr>
<tr>
<td>S</td>
<td>stopwatch</td>
<td>58</td>
<td>130</td>
<td>3.786 s</td>
</tr>
<tr>
<td>S</td>
<td>elevator simulation</td>
<td>40</td>
<td>114</td>
<td>1.586 s</td>
</tr>
<tr>
<td>S</td>
<td>robotics systems</td>
<td>68</td>
<td>256</td>
<td>4.311 s</td>
</tr>
</tbody>
</table>
A Comparison between Fixed Priority and EDF Scheduling accounting for CRPD

Table 4 Weighted schedulability measures for the mixed case study shown in Figure 8. The higher the weighted schedulability measure, the more tasksets deemed schedulable by the approach.

<table>
<thead>
<tr>
<th></th>
<th>Weighted Schedulability</th>
</tr>
</thead>
<tbody>
<tr>
<td>EDF – No Pre-emption Cost</td>
<td>0.922</td>
</tr>
<tr>
<td>FP – No Pre-emption Cost</td>
<td>0.855</td>
</tr>
<tr>
<td>EDF – Optimised Layout</td>
<td>0.830</td>
</tr>
<tr>
<td>EDF – Standard Layout</td>
<td>0.771</td>
</tr>
<tr>
<td>FP – Optimised Layout</td>
<td>0.784</td>
</tr>
<tr>
<td>FP – Standard Layout</td>
<td>0.747</td>
</tr>
</tbody>
</table>

EDF outperformed FP as it deemed a higher number of tasksets schedulable at each utilisation level. Because the range of execution times is relatively small, so is the typical range of task periods for the generated tasksets, hence the number of pre-emption is also relatively small. Further, the number of UCBs is small, resulting in low CRPD. Therefore, the task layout optimisation was only able to make a small improvement, but did so for both FP and EDF.

6.2.2 Mälardalen and SCADE Benchmarks

The second multiple taskset case study was based on tasks from the Mälardalen and SCADE benchmarks, shown in Table 3. Compared to the tasks from PapaBench, these tasks have higher execution times, high amounts of computation, and a larger number of UCBs. Figure 7 shows the percentage of tasksets that were deemed schedulable by each approach for the 1000 tasksets of cardinality 10 that we randomly selected from Table 3. As with the PapaBench benchmarks, EDF outperformed FP scheduling as it has a higher percentage of schedulable tasksets at each utilisation level. Likewise, because the range of task periods was also relatively small, CRPD is minimised.

6.2.3 Mixed Benchmarks

The third multiple taskset case study was based on a mixture of the small and short PapaBench tasks, and the large and long Mälardalen and SCADE tasks. Here the tasksets had 15 tasks each, and represent systems with background tasks combined with short control tasks. As we mixed tasks from both tables, it also allowed us to generate tasksets with a higher number of tasks.

The results, shown in Figure 8, show that when a taskset contains tasks with a wide range of periods, CRPD can become a significant factor in the schedulability of the taskset. This is because short high priority tasks are able to pre-empt long running low priority tasks multiple times.

While EDF still outperformed FP, the gain in schedulability of using EDF over FP was diminished once CRPD was taken into account. Optimising the task layout resulted in a significant improvement for both FP and EDF, showing the task layout optimisation can be effectively applied to both EDF and FP scheduling. Furthermore, by optimising the task layout, FP was able to schedule a similar number of tasksets to EDF with the standard layout. In other words, in cases where the CRPD is relatively high, selecting an optimised task layout can be as effective as changing the scheduling algorithm. The results are summarised in Table 4 using weighted schedulability measures [8], see Section 7.2 for details. They show that for these tasksets, FP with an optimised layout achieved a weighted measure of 0.784, outperforming EDF with the standard layout as it achieved a weighted measure of 0.771.
Figure 6 Percentage of schedulable tasksets at each utilisation level for the PapaBench benchmark for tasksets of cardinality 10.

Figure 7 Percentage of schedulable tasksets at each utilisation level for the Mälardalen and SCADE benchmarks for tasksets of cardinality 10.

Figure 8 Percentage of schedulable tasksets at each utilisation level for the mixed case study with tasks randomly selected from both the PapaBench and Mälardalen and SCADE benchmarks (taskset cardinality 15).
7 Evaluation

In addition to the case studies based on the PapaBench, Mälardalen and SCADE benchmarks, we evaluated FP and EDF with CRPD analysis using synthetically generated tasksets. This enabled us to investigate the effect of varying key parameters under each scheduling algorithm.

The UUnifast algorithm [9] was again used to calculate the utilisation, $U_i$ of each task so that the utilisations added up to the desired utilisation level for the taskset. Task periods $T_i$, were generated at random between 5 ms and 500 ms according to a log-uniform distribution. $C_i$ was then calculated via $C_i = U_i T_i$.

We generated two sets of tasksets, one with implicit deadlines and one with constrained deadlines. In the following section, we present the results for constrained deadline tasksets. In general, the results for implicit deadline tasksets gave a higher number of schedulable tasksets for every approach compared to the constrained deadline tasksets. Additionally, the task layout had a similar or slightly larger effect on schedulability in relation to the chosen scheduling algorithm.

The UCB percentage for each task was based on a random number between 0 and a maximum UCB percentage specified for the evaluation. UCBs were split into $N$ groups (where $N$ was chosen at random between 1 and 5), and placed at a random starting point within the task’s ECBs.

7.1 Baseline Configuration

To investigate the effect of key cache and taskset configurations we varied the following parameters:

- Cache utilisation (default of 10)
- Maximum UCB percentage (default of 30%)
- Number of tasks (default of 15)
- Block Reload Time (BRT) (default of 8 $\mu$s)
- Period range (default of [5, 500] ms)

We used 1,000 randomly generated tasksets for the evaluation.

In addition to testing the different analyses as done for the case study, we also performed a simulation of the schedule for the tasksets. For FP, the simulation tested each task $\tau_i$ in turn by releasing it at time $t = 0$. It then released all of the other tasks that have a higher priority than task $\tau_i$, sorted by lowest to highest priority, at $t = 1$, $t = 2$, $t = 3$, etc. If all tasks were schedulable it also performed the same test but instead of staggering the other tasks, released them at random. For EDF, it is more complicated to generate the worst case arrival pattern. The first step is to determine the interval that needs to be checked, $L$, which can be achieved by using equation (30). Then for each task $\tau_i$ in turn, we scheduled a job of task $\tau_i$ so that it has a deadline at $t = L$. We then scheduled a job of all of the other tasks, sorted by longest to shortest deadline, so that they have their deadlines at $t = L - 1$, $t = L - 2$, $t = L - 3$ etc... Based on the final jobs’ deadlines, we then calculated when the first jobs for each task need to be released. If all tasks are schedulable, we repeated the process using $t = L - 1$ for all of the other tasks’ jobs, and also using a random schedule.

The results for the baseline configuration are shown in Figure 9 and are summarised in Table 5 using weighted schedulability measures. The results follow a similar pattern to the results from the case study. EDF outperformed FP finding a higher number of tasksets schedulable. The results for the simulations show that the CRPD affects both FP and EDF, with the CRPD being slightly lower for EDF. Specifically, the simulation shows that CRPD reduced the weighted measure by at least 0.129 for EDF (0.925 – 0.795) and 0.141 for FP (0.774 – 0.633) in this case. However, once

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Note that the simulation effectively provides a necessary, but not sufficient test of schedulability.
The percentage of schedulable tasksets at each utilisation level for the baseline configuration (taskset cardinality 15).

Table 5 Weighted schedulability measures for the baseline configuration study shown in Figure 9. The higher the weighted schedulability measure, the more tasksets deemed schedulable by the approach.

<table>
<thead>
<tr>
<th>Method</th>
<th>Weighted Schedulability</th>
</tr>
</thead>
<tbody>
<tr>
<td>EDF – No Pre-emption cost</td>
<td>0.925</td>
</tr>
<tr>
<td>EDF – Simulation</td>
<td>0.796</td>
</tr>
<tr>
<td>FP – No Pre-emption cost</td>
<td>0.774</td>
</tr>
<tr>
<td>FP – Simulation</td>
<td>0.633</td>
</tr>
<tr>
<td>EDF – Optimised layout</td>
<td>0.455</td>
</tr>
<tr>
<td>EDF – Standard layout</td>
<td>0.413</td>
</tr>
<tr>
<td>FP – Optimised layout</td>
<td>0.369</td>
</tr>
<tr>
<td>FP – Standard layout</td>
<td>0.336</td>
</tr>
</tbody>
</table>

the CRPD obtained via analysis is taken into account, the performance gains of using EDF over FP are diminished. This is most likely caused by increased pessimism in the CRPD analysis for EDF. The results for the layout optimisation showed that it was able to make improvements to the schedulability of tasksets scheduled under both FP and EDF.

7.2 Weighted Schedulability

Evaluating all combinations of different parameters is not possible. Therefore, the majority of our evaluations focused on varying one parameter at a time. To present the results, weighted schedulability measures [8] are used. This allows a graph to be drawn which shows the weighted schedulability, \( W_l(p) \), for each method used to obtain a layout \( l \) as a function of parameter \( p \). For each value of \( p \), this measure combines the data for all of the generated tasksets \( \tau \) for all of a set of equally spaced utilisation levels, where the utilisation is based on no pre-emption cost.

The schedulability test returns a binary result of 1 or 0 for each layout at each utilisation level. If this result is given by \( S_l(\tau, p) \), and \( u(\tau) \) is the utilisation of taskset \( \tau \), then:

\[
W_l(p) = \frac{\sum_{\tau} u(\tau) S_l(\tau, p)}{\sum_{\tau} u(\tau)}. \tag{31}
\]
The benefit of using a weighted schedulability measure is that it reduces a 3-dimensional plot to 2 dimensions. Individual results are weighted by taskset utilisation to reflect the higher value placed on a being able to schedule higher utilisation tasksets.

### 7.2.1 Cache Utilisation

As the cache utilisation increases the likelihood of tasks evicting each other from cache increases, this causes higher CRPD reducing the number of schedulable tasksets. It can be seen in Figure 10 that task layout optimisation is effective for FP and EDF across the same range of cache utilisations. In both cases it becomes less effective once the cache utilisation becomes high. We note that because the number of tasks is fixed, that the average size of the tasks is equal to the cache utilisation divided by the number of tasks. This means that as the cache utilisation increases, so does the size of the tasks and therefore, the number of UCBs. This in turn makes it harder to find an improved layout.

### 7.2.2 Maximum UCB Percentage

As the maximum UCB percentage increases, the CRPD increases resulting in a reduction in the number of tasksets that are deemed schedulable, as can be seen in Figure 11. With a low percentage of UCBs, the CRPD is low which means there is little benefit from layout optimisation. When the UCB percentage is very high, there are so many conflicts that there is very little that can be done to improve the layout. When the maximum UCB percentage is around 40–60%, there is a notable amount of CRPD, but there is also room for the task layout algorithm to optimise the layout. This allows FP using an optimised task layout to schedule a similar number of tasksets as EDF using the standard layout.

### 7.2.3 Number of Tasks

When varying the number of tasks, Figure 12, we scaled the cache utilisation to keep the average size of tasks constant based on a cache utilisation of 10 for 15 tasks. This is because it would be unrealistic for the size of tasks to decrease as more tasks are added to the system. Hence with 8 tasks the cache utilisation is equal to 5.33, whereas for 32 tasks, it is equal to 21.33. As the number of tasks increases, it becomes harder the schedule all tasks which leads to a decrease
in the overall weighted measure. The task layout optimisation performs best when there is a moderate number of tasks, as there are enough conflicts that optimising the layout can give an improvement, but not so many that there is nothing that can be done to avoid the conflicts.

### 7.2.4 Block Reload Time

As the block reload time is increased, it becomes more costly to reload a block, which causes an increase in CRPD. It can be seen in Figure 13 that as the block reload time is increased, the analysis that takes into account the pre-emption cost shows a decrease in the overall weighted measure. We note that as the cost of reloading a block increases, the potential gains of optimising the layout increase. Once the block reload time exceeds 14 µs, using an optimised layout under FP scheduling outperforms using a standard layout under EDF scheduling.

### 7.2.5 Period Range

We also investigated the effect of the scaling factor used to generate task periods to simulate tasksets with shorter to longer execution times. We varied the scaling factor, \( w \), from 0.5 to 10 and hence the range of task periods given by \( w[1, 100] \) ms. A lower scaling factor resembles tasks with shorter execution times, as seen in the PapaBench benchmark, and a higher scaling factor resembles tasks with higher execution times and commensurately longer periods, as seen in the Mälardalen and SCADE benchmarks. The results in Figure 14 show the layout optimisation performs best when task periods are relatively short, as that is when the pre-emption costs are highest. Once the period range is greater than \([10, 1000] \) ms, the relative pre-emption costs are low enough that performing the layout optimisation only makes a very small improvement on the schedulability of the tasksets.

### 8 Conclusion

The EDF scheduling algorithm is an optimal scheduling algorithm for single processors however, it has been largely disregarded by industry. Whereas FP, despite offering lower theoretical schedulable processor utilisation, is relatively popular with many commercial real-time operating systems supporting it.

Previous work by Buttazzo [13] has compared the two algorithms, but it did not take into account CRPD which can have a significant effect on the schedulability of a taskset.
A Comparison between Fixed Priority and EDF Scheduling accounting for CRPD

**Figure 12** Weighted measure for varying the number of tasks from $2^0$ to $2^6$ while maintaining a constant ratio of number of tasks to cache utilisation.

**Figure 13** Weighted measure for varying the block reload time from 0 to 20 µs in steps of 2.

**Figure 14** Weighted measure for varying the scaling factor used to generate periods, $w$, in $w[1, 100]$ ms, from 0.5 to 10.
The main contributions of this paper are:
- Performing a detailed comparison of FP and EDF taking into account CRPD using state-of-the-art CRPD analysis [3, 22].
- Showing the feasibility of simple, yet effective, task layout optimisation techniques for EDF.
- Finding that when CRPD is considered, the performance gains offered by EDF over FP, while still significant, are somewhat diminished. This is most likely due to greater pessimism in the CRPD analysis for EDF than FP.
- Discovering that in configurations that cause relatively high CRPD, optimising task layout can be just as effective as changing the scheduling algorithm from FP to EDF.

We investigated the effects of performing task layout optimisation based on Simulated Annealing under both FP and EDF scheduling algorithms. We found that in the scenarios that cause the pre-emption cost to be relatively high in relation to task execution times, applying task layout optimisation to a system scheduled using FP scheduling can allow it to be schedulable at a similar processor utilisation compared to using EDF scheduling with a standard layout. This is important in an industrial setting as it is considerably simpler and cheaper to control the task layout via the linker, than it is to change the scheduler. Nevertheless, our evaluations showed that changing to an EDF scheduler and optimising the task layout provides a gain over FP scheduling. Although this gain was not as pronounced as the advantage that EDF has over FP when pre-emption costs are not accounted for via analysis.

In the future we plan to further investigate techniques for CRPD analysis, and to apply them in an industrial context comparing the results of analysing a real system with those obtained via measurement.

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A Comparison between Fixed Priority and EDF Scheduling accounting for CRPD


TLM.open: a SystemC/TLM Front-end for the CADP Verification Toolbox

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Abstract

SystemC/TLM models, which are C++ programs, allow the simulation of embedded software before hardware low-level descriptions are available and are used as golden models for hardware verification. The verification of the SystemC/TLM models is an important issue since an error in the model can mislead the system designers or reveal an error in the specifications. An open-source simulator for SystemC/TLM is provided but there are no tools for formal verification.

In order to apply model checking to a SystemC/TLM model, a semantics for standard C++ code and for specific SystemC/TLM features must be provided. The usual approach relies on the translation of the SystemC/TLM code into a formal language for which a model checker is available.

We propose another approach that suppresses the error-prone translation effort. Given a SystemC/TLM program, the transitions are obtained by executing the original code using g++ and an extended SystemC library, and we ask the user to provide additional functions to store the current model state. These additional functions generally represent less than 20% of the size of the original model, and allow it to apply all CADP verification tools to the SystemC/TLM model itself.

1 Introduction

The design of abstract models written in SystemC/TLM has become more common in the development of embedded systems. These models allow the simulation of the embedded software before the hardware RTL description is available, and are used as golden models for hardware verification. The verification of the SystemC/TLM models is an important issue since an error in the model can mislead the system designers or reveals an error in the specifications.

ASI (Accellera Systems Initiative, previously OSCI: Open SystemC Initiative) provides an open-source simulator for SystemC/TLM and a library SCV (SystemC Verification) to ease test generation. However, ASI does not provide tools for formal verification. Moreover, while the SystemC specification allows many schedules for a given test case, the ASI simulator always exhibits the same schedule. Thus, even if an execution leads to the expected result, another execution with a different schedule may be erroneous. To find these kind of bugs, many publications have experimented with the use of model checking.

The problem of verifying C++ and SystemC codes could be avoided by writing transactional models in a formal language directly. However, in order to model embedded systems at the...
transaction level, engineers of industrial companies prefer to use SystemC/TLM. One reason is that SystemC/TLM provides all the useful features directly, like shared memory and transactional communication channels. Another reason is that a SystemC/TLM program is mainly C++ code, so engineers can learn SystemC/TLM quickly, and existing C code can be reused easily.

In order to apply model checking to a SystemC/TLM program, the usual approach relies on the translation of the SystemC/TLM code into a formal language for which a model checker is available. A lot of languages and tools have been tested so far (see Subsection 3.1.2). Nonetheless, there have been few successes with industrial case studies.

We propose another approach that suppresses the translation effort. Basically, an explicit model checker must be able to execute transitions and store states. Given a SystemC/TLM program, we assume that the states are the places where processes yield back to the scheduler. Consequently, transitions correspond to pieces of C++ code delimited by yield points: either wait statements or return statements from the process main function. We obtain the transitions by executing the original code using g++ and a SystemC library, as in any simulator. Storing the state could be done by copying the whole memory used by the simulator, but would be inefficient. Therefore, we ask the user to provide additional functions to store the current state and restore a previous state. Part of the state, including the SystemC kernel, is stored automatically; so in general the user can only store the SystemC module data members.

Following this approach, we have developed a new front-end for the CADP tool suite. The CADP tool suite includes many tools useful for formal verification and bug finding; the main tool is an explicit model checker. This article does not introduce a new verification technique (we did not change anything in CADP) except a pragmatic and efficient way to use existing tools to verify programs written in a language that has not been designed to ease formal verification. The new front-end we have developed is not fully automatic since the user must provide some additional functions; these additional functions generally represent less than 20% of the size of the original model.

The model checking technique is known to be limited by the state space explosion. Because we rely on this technique and there are no changes in the core algorithm, we are limited in this area. Nevertheless, model checking has been applied to many real-life case studies (over 150 using CADP in many application fields\(^1\) most of the times, using model checking allowed to verify properties or discover bugs. We have written our new front-end in a way that avoids to make the state space explosion even worse by adding intermediate states and transitions, which was the case using a previous approach [16]. Experiments which were first made with benchmarks, then with a single SystemC module, and finally with a basic system, show that we can indeed find bugs and prove some properties on real-life TLM models.

The remainder of this article is organized as follows. We present briefly SystemC and TLM in Section 2. Section 3 gives an overview of the related work and presents the existing CADP toolbox. Section 4 describes our technique to connect SystemC/TLM with CADP. The performances of TLM.open are evaluated in Section 5 and Section 6 concludes this article.

## 2 SystemC and TLM

SystemC [1] is a C++ library published by the Accellera Systems Initiative (ASI) and defined by an IEEE standard which provides classes to describe heterogeneous systems composed of hardware and software. The architecture of a system is defined by a set of modules connected by synchronous or asynchronous ports and channels (sc_module, sc_port, ...). Each module contains zero, one, or

\(^1\) Case studies achieved using the CADP toolset: [http://www.inrialpes.fr/vasy/cadp/case-studies](http://www.inrialpes.fr/vasy/cadp/case-studies).
various processes (SC_THREAD or SC_METHOD) describing the system’s behavior. SystemC processes interact using shared memory or communication channels and are synchronized using SystemC events (sc_event e, e.notify(), wait(e)) with timing annotations (sc_time t, wait(t)).

Each SystemC process is a C++ method that is executed by the SystemC scheduler communicates with other processes using shared memory and may explicitly suspend itself by executing a `wait` statement. When the process is resumed by the scheduler, its execution continues from the `wait` statement. Each SystemC process is eligible or running or waiting for a SystemC event. There is, at most, one running process simultaneously. If the running process notifies an event, then all processes waiting for this event move from waiting to eligible.

The Transaction Level Modeling (TLM) library [10] built upon SystemC, provides a transaction mechanism that encapsulates communication protocols (data transfer and synchronization) between modules and accelerates both model design and simulation. Using a transaction, a process in an `initiator` module can directly call the methods exported by a `target` module. Thus, a process can read many values from a memory, or set many registers of a peripheral without any costly inter-process synchronization (no context switch is required). At the TLM level of abstraction, processes inside the same module communicate using SystemC events and shared variables. A TLM model can be timed or untimed: a timed model contains timing annotations (sc_time t, wait(t)) whereas an untimed model does not. An untimed model includes more possible behaviors than a timed model, increasing the coverage, but also the cost, of the verification.

Because SystemC and TLM are C++ libraries, simulating a SystemC/TLM model does not require a dedicated SystemC/TLM parser. A SystemC/TLM model is parsed and compiled as with any C++ program, using a regular C++ compiler, such as g++.

3 Related Works

3.1 Verification of SystemC/TLM models

In order to provide formal verification for SystemC/TLM programs, two approaches were investigated: stateless model checking of a SystemC/TLM program and a translation of a SystemC/TLM program into a language for which a stateful model checker is available.

3.1.1 Stateless model-checking

A stateless model-checker explores the set of all the possible executions of a given program without storing the states. Because the states are not stored, a stateless model-checker can execute the same transition many times. Moreover, if the program under verification has at least one possible execution that does not terminate then the stateless model-checker will not terminate either. However, stateless model-checkers have benefits: 1. naive stateless model-checkers are easy to implement because one just needs to modify the functions used for non-deterministic choices; 2. their memory consumption is limited (linear in terms of execution lengths).

Many stateless model checking tools have been implemented for SystemC/TLM programs [15, 21, 2]. In order to reduce the number of executions explored, these model-checkers select a subset for the possible executions; this subset is guaranteed to detect all the errors of a particular family; such as all the assertion failures or all the deadlocks. All these stateless model-checkers implement dynamic partial order reduction [5]; the selection of the executions explored is based on the analysis of detailed execution traces. The dynamic partial order algorithm was specifically adapted for the particularities of the SystemC scheduling policy.

In particular, [14, 15] show how to validate programs with loose timing annotations encoded by bounded intervals. This technique extracts a finite subset from the infinite set of the timings allowed
by the specification. Given a program that always terminates and without non-deterministic data choices, this technique detects all the assertion failures and the deadlocks.

These tools give interesting results for small and medium sized industrial examples. Using SCRV [13], a synchronization error was found in a model of a video decoder provided by STMicroelectronics. However, stateless model-checkers can only be applied to terminating programs without non-deterministic unbounded data inputs.

### 3.1.2 Translate then verify

For programs that do not terminate, a second approach was investigated. The idea was to translate the SystemC/TLM program to be verified into another language, and then verify the translated program using an existing stateful model checker. This approach has first been applied to the RTL level SystemC descriptions [4, 11].

Many translations and languages have been proposed for the validation of transactional models, as in [25], which translates SystemC/TLM programs into finite state machines (FSM), similarly [20], which describes abstraction techniques and a translation from SystemC/TLM to labeled Kripke structures. Most of these translations are manual, the first exception being the LusSy tool chain [24], which automatically translates TLM models into synchronous automata with variables; it provides some simple abstraction techniques (e.g., abstract address representation). The LusSy tool chain has been connected to many model checkers, including symbolic model checkers based on BDD or SAT. Some minor examples have been successfully verified, but industrial examples face the state space explosion problem. There are now other automatic translation tools starting from SystemC, including [17] that can translate SystemC/TLM models into UPPAAL models, and allow verification of liveness properties and timing constraints. [12] translates TLM models into sequential C programs, in order to use verification tools dedicated to software.

The state space explosion problem appears mainly because TLM models are mostly asynchronous. Thus, after each transition, there are many valid scheduling choices that should be explored. It is therefore suitable to use the model checkers for asynchronous programs as these model checkers have been specifically optimized to fight state space explosions arising from asynchrony. Translation of TLM programs into Promela [27] has allowed TLM models to be verified using the SPIN checker which uses partial orders to reduce state spaces. Since then other translations into Promela have been presented [22, 3], allowing the verification of larger models. Note that the translation defined in [3] was implemented in an automatic tool where other back-ends were available.

Furthermore, we firstly proposed a translation of TLM to LOTOS [16, 26] that enables verification of the benchmark of [27] for a slightly greater number of processes than using SPIN. The translation was fully manual, preventing the approach to scale up.

Next, [7] presented both an extension of our TLM to LOTOS translation and an application to an industrial case study. As shown in Figure 1, part of the SystemC/TLM code was translated into LOTOS while data-types and related operations were kept as C++ code, thus strongly reducing the translation effort. This paper showed that some properties can be checked on industrial code, but the amount of manual work still limited the efficiency of the approach.

### 3.2 The CADP verification toolbox

Our goal is to detect synchronization errors between asynchronous processes of SystemC/TLM programs or to guarantee that the communication protocol they use is correct. For this kind of task one of the best techniques is model checking and in particular explicit model checking.
A well-known explicit model-checker is SPIN. In this work, we investigated the use of another model-checker; namely CADP.

CADP (“Construction and Analysis of Distributed Processes”) \cite{CADP} is a toolbox for the validation of communication protocols and distributed systems.

The usual entry point for CADP is the language LOTOS. The ISO standard LOTOS \cite{LOTOS} (Language Of Temporal Ordering Specification) is a process algebra used to describe asynchronous concurrent processes communicating and synchronizing by rendezvous on gates. This language is well suited for designing communication protocols.

The semantics of a LOTOS specification is formally defined by a state graph, also called an LTS (labeled transition system) – i.e. a set of states and transitions labeled by gates and offers between states.

CADP \cite{CADP} includes a compiler from LOTOS to LTS with many tools exploiting the LTS for simulation as well as model checking of modal $\mu$-calculus formulae, equivalence checking, test generation and performance evaluation. The LOTOS to LTS compiler generates the LTS by executing all transitions of the system under verification; each visited state is recorded, so that each transition is executed once only. The algorithm is shown in Figure 2.

\section{Connecting SystemC/TLM with formal methods}

\subsection{The architecture of TLM.open}

The CADP toolbox architecture is similar to the GNU compiler tool suite, with many front-ends and back-ends. There is one front-end per input language; the front-end reads a program and implements some basic analysis (e.g., type checking). Then there is one back-end per CADP feature, such as simulation, LTS generation, or on-the-fly property checking. The most used front-ends are caesar.open which manages LOTOS programs and bcg_open which reads compressed
and explicit LTS (\texttt{bcg} stands for “binary coded graphs”). All CADP front-ends connect with CADP back-ends using the OPEN/CÆSAR interface [6].

In this section, we present \texttt{TLM.open} which is a new CADP front-end allowing the use of the same back-ends as \texttt{caesar.open} and \texttt{bcg_open}. \texttt{TLM.open} is a C and C++ library that implements two interfaces: the SystemC interface and the OPEN/CÆSAR interface. The architecture of \texttt{TLM.open} is shown in Figure 3.

A SystemC/TLM program communicates with \texttt{TLM.open} through the SystemC interface as a SystemC/TLM program communicates with a SystemC simulator. The library \texttt{TLM.open} provides the same classes as the ASI SystemC simulator, including the \texttt{sc_module}, \texttt{sc_port}, \texttt{sc_event}, \texttt{sc_signal}, etc.

The OPEN/CÆSAR interface provides the operators required by the CADP model-checker itself. In order to implement the algorithm described by Figure 2, the following operators are required and must be provided by the \texttt{TLM.open} front-end:

- generation of the initial state
- enumeration and simulation of the transitions starting from a given state
- efficient storage of a state (requires comparison and hash functions).

To simulate a transition, \texttt{TLM.open} executes the corresponding C++ code of the SystemC/TLM program. This C++ code is compiled with an unmodified C++ compiler such as \texttt{g++}. \texttt{TLM.open} does not parse the C++ code itself and does not produce LOTOS code.

The most difficult task is to store and restore the states of the SystemC/TLM program. The person who writes and verifies the SystemC/TLM program, called user in this paper, has to provide some additional functions that allow \texttt{TLM.open} to store the states of each SystemC module. To date, these additional functions have had to be written by hand. Thus, our approach is not fully automatic.

When \texttt{TLM.open} is used with the LTS generator of CADP, the result is an LTS with two kinds of transitions. Here, \textit{offers} are only used to add information to the transitions, and have no impact on synchronizations or communications.

- \texttt{TE} transitions indicate that time has elapsed; the offer gives the duration and the list of events triggered and processes awakened. For example, “\texttt{TE !o(+41ms, VGAC.compute)}” means that the SystemC clock has advanced 41 ms and the process “VGAC.compute” is now awake.

- \texttt{EXEC} transitions represent the execution of a SystemC process; the offers name the executed SystemC process, the inputs of this process if the special \texttt{rand()} function of \texttt{TLM.open} was called (cf. Section 4.3.2) and the outputs generated using the overloaded \texttt{puts()} function. For example, “\texttt{EXEC !VGAC.compute !o(image updated, IRQ sent)}” means that the SystemC scheduler has executed the process “VGAC.compute” and this process has printed two messages “image updated” and “IRQ sent”.

![Figure 3](image-url) Overview of the verification framework. The model is linked with the SystemC \texttt{TLM.open} library instead of the ASI library.
4.2 Storing and restoring program states

The TLM.open library includes a SystemC simulator. The state of this simulator consists of the state (i.e., the current value) of each object that has been instantiated. Some objects are described by SystemC classes (such as: `sc_event`, `sc_signal`, ...) and others are described by user classes. SystemC modules are hybrid: some class members are inherited from the base class `sc_module` but other members are defined by the user.

A stored state contains a copy of each value of the simulator state that may change during the simulation. A stored state must be as small as possible and does not use the same types as the simulator states: constant values are not stored, Boolean values can be grouped in one byte using bit fields.

All objects which are defined by a SystemC class are stored automatically by the TLM.open library. The other objects are stored using callback functions implemented by the user. Each SystemC module must provide the following functions:

- `size_t size() const`: number of bytes needed to store a copy of the SystemC module.
- `size_t alignment() const`: specify whether padding bytes are needed.
- `void store(char *dest) const`: store the current state of the module in dest.
- `void restore(const char *src)`: restore the state of the module according to the copy stored in src.

The `store()` function must generate a canonical representation, so that state comparison can be done using `memcmp()` and hash functions can be generated automatically.

Implementing the functions `size()` and `alignment()` generally requires only one line of code for each. The `store()` function implementation contains two lines of code per module member on average; similarly the `restore()` function. Implementing these functions requires some manual work, but less than translating the whole model into another language.

Theoretically, generating automatically the `store()` and `restore()` functions should be simpler than translating the whole code, because it is not necessary to manage the code describing the behavior. However, such a generator would have to parse and manage a large part of C++, and the generated functions would likely be less efficient than those hand-written.
4.2.1 Storing modules using flat state

The user has many possibilities to implement the store and restore functions. The basic solution is to define a new struct type with one field per member of the SystemC module that is not constant and not managed directly by TLM.open. To store the state, the user filled this new type by copying values from the C++ class, and inversely, the user filled the C++ class by copying values from the struct type when the state must be restored. This is shown in the example below.

SC_MODULE(Foo) {
    sc_event e; // state stored by the tlm.open library
    bool flag; // dynamic data
    uint32_t data; // dynamic data
    const sc_time period; // static data, not stored
    ... // module implementation

    // code below is used only by TLM.open
    struct State { // container type
        bool flag; uint32_t data;
        void set(Foo *f) const {f->flag=flag; f->data=data;}
        void set(const Foo *f) {flag=f->flag; data=f->data;}
    };
    size_t size() const {return sizeof(State);}
    size_t alignment() const {return 4 /*alignmentof(State)*/;}
    void store(char *dst) const {
        reinterpret_cast<State*>(dst)->set(this);
    }
    void restore(const char *src) {
        reinterpret_cast<const State*>(src)->set(this);
    }
}; // Foo

In this example, storing the state of an instance of Foo requires 8 bytes (i.e., sizeof(Foo::State)). If a program contains \( n \) modules \( M_1, \ldots, M_n \), each module being stored using a type \( M_i::\text{State} \), then each state stored consumes at least \( \sum_{i=1}^{n} \text{sizeof}(M_i::\text{State}) \) bytes.

4.2.2 Storing module using hierarchical state

Most of the time, a transition modifies the state of only one or two modules. If storing a module consumes a lot of memory, it is then mostly better to use a hierarchical state. Using hierarchical states, the main state contains a pointer to the module state instead of the module state itself. When a transition is executed and the module has not been modified, then the new stored state contains only a pointer to the previously stored value.

Moreover, checking whether the module has been modified by the last transition is not enough. Even if the module has been modified, it is possible that we already have a copy of its current state. At the end of a transition, we search all the previous states of this module, which are stored in a container (hash table or binary tree). If this module state is encountered for the first time, then it is added to this container, or else we reuse the existing module state.

Here is how the Foo state could be recorded using a hierarchical state:

typedef std::set<const State*, StateCmp> state_set;
static state_set foo_states;
size_t Foo::size() const {return sizeof(State);};
void Foo::store(char *dst) const {
    State *s = new State(); s->set(this);
    std::pair<state_set::iterator,bool> p = foo_states.insert(s);
    if (!p.second) delete s; //This Foo state already exists,
    // so we reuse the previous version.
    *reinterpret_cast<const State**>(dst) = *p.first;}

To compare two states of the whole program, we just need to compare the pointers because identical module states are never stored in distinct memory locations.

In some cases, hierarchical states can significantly reduce the memory consumption. Moreover, whereas the OPEN/CÆSAR interface requires the main state to have a fixed size, hierarchical states allow a module to be stored whose size is not statically bound.

Internally, for all objects that are stored automatically, the TLM.open library uses a flat state for all objects except SystemC threads (SC_THREAD). Moreover, storing the state of a thread is done by copying its execution stack. Note that when yielding, the QuickThreads library used by SystemC pushes the register contents and the program counter on top of the thread stack. As thread stack sizes vary during simulation because stacks may become large, and because at most one thread stack is modified during a transition, the hierarchical state technique here is more efficient than flat states.

4.3 Implementation of the OPEN/CÆSAR interface

4.3.1 Generation of the initial state

The generation of the initial state faces a technical problem. Moreover, SystemC and CADP do not use the same control flow:

- A SystemC simulator creates the initial state by calling the function *sc_main*, which is implemented by the user, and the simulation starts when the user calls back the function *sc_start* from the *sc_main* function.

- A CADP back-end creates the initial state by calling the function CAESAR_START_STATE, which is implemented by the front-end and the verification starts after the function CAESAR_START_STATE returned.

Thus, the CADP back-end calls the function CAESAR_START_STATE of TLM.open, and this function calls sc_main. The function CAESAR_START_STATE must return when sc_start is called, before sc_main returns. If one returns in advance of sc_start using a return statement or a C long jump or a C++ exception then the modules allocated on the stack are destroyed before they are used. The solution is to execute the sc_main function in a separated thread, which has its own stack and suspends this thread when sc_start is called. As we do not need real concurrency, this thread is implemented using the collaborative QuickThreads library, which is used to implement the SystemC threads too.

4.3.2 Enumerating the transitions

The key function of the OPEN/CÆSAR interface is CAESAR_ITERATE_STATE. This function must enumerate the transitions starting from a given stored state \( x \). A transition is defined by a label \( s \) (a C string) and the successor state \( y \).

There is at least one transition per eligible process. Assuming all transitions are deterministic, the TLM.open library behaves as follows:

1. A SystemC process is selected.
2. The simulator is set according to the stored state $x$, by calling the \texttt{restore} function of each stored object (either a user function for modules, or a \texttt{TLM.open} library function for other SystemC objects).
3. The transition is executed, until the elected process yields back to the scheduler.
4. The new simulator state is stored in $y$, by calling the \texttt{store} function of each stored objects.
   The label $s$ is created using the name of the elected process, and the outputs generated by the user using the \texttt{puts()} function.
5. This transition is sent to the back-end.
6. If there is another eligible thread, then go back to step 1.

If no processes are eligible, then \texttt{TLM.open} can let the time elapse until a process is awoken, just like a regular SystemC simulator. In this case, a specific transition is generated with the label “TE”. If no process can be awoken by a time elapse, then this means that $x$ is in a deadlock state. In order to simulate inputs or a non determinism, the \texttt{TLM.open} library provides a \texttt{rand(int MAX)} function. From the user point of view, this function returns a number between 0 and MAX. In case of a simulation, an implementation would choose a number randomly. On the contrary, model checking requires an enumeration of all values. In order to generate the full LTS, each time \texttt{TLM.open} encounters a call to \texttt{rand()}, it records that another transition exists for the same process for which values have already been tried. Thus, the same code will be executed $MAX+1$ times, generating as many LTS transitions. Because a transition may call \texttt{rand()} many times, \texttt{TLM.open} uses a stack to remember its position in the transition tree. Thus, all input combinations are finally generated (e.g., “$x=\texttt{rand}(2); y=\texttt{rand}(3); \texttt{wait}();$” generates $3 \times 4 = 12$ transitions).

### 4.4 Features and limitations

Most SystemC, TLM and C++ features can be used normally. However, some features require special care. As aforementioned, the functions \texttt{puts()} and \texttt{rand()} have a special meaning when used with \texttt{TLM.open}.

#### 4.4.1 The \texttt{sc_stop} function

SystemC provides a function \texttt{sc_stop()} to stop the simulation. Because all states that can be reached using a simulator must be reached using \texttt{TLM.open}, calling \texttt{sc_stop()} may not stop the generation of the LTS. With respect to the SystemC specifications, the effect of executing \texttt{sc_stop()} in a transition $x \rightarrow y$ is to eliminate all the successors of $y$. If other transitions are pending, then they are explored normally.

#### 4.4.2 Recording the current time

A SystemC simulator, such as the ASI simulator, records the current date. The user can read this data using the function \texttt{sc_timestamp()}. Because this value is stored in the state, the state space becomes infinite for all programs containing a timed instruction in an unbounded loop. An example of such program is:

```c
SC_THREAD(compute); ...
void compute() {
   while (true) {wait(1,SC_SEC);}}
```

To allow the verification of this program, \texttt{TLM.open} provides an option to record only relative durations. This option disables the function \texttt{sc_timestamp()}. Using this option, the LTS of the program above has only two states and two transitions: a transition with gate “EXEC” leads from the initial state to the second state and another transition with gate “TE” leads back to the initial state.
4.4.3 Pointers and dynamic allocations

It is perfectly safe to use pointers when verifying a SystemC/TLM program with TLM.open. Both the pointer and the pointed value must be stored (respectively restored) when the module is stored (resp. restored).

However, dynamic allocations should not be used because a transition can be executed many times, calling new creates a memory leak (a second object will be created if the transition is executed again), and calling delete corrupts the memory (memory can be freed twice). There is one exception: a new statement can be used safely if the corresponding delete statement is found in the same transition.

If using dynamic allocation is necessary, then the user must define its own memory allocator. Next, the user must provide store and restore functions to manage the state of the memory allocator itself. Thus, when a state is restored, the memory allocator knows which objects are allocated and which memory locations are available.

5 Examples

5.1 The chain benchmark

We evaluate our new front-end on the benchmark proposed in [27] and reused in [16]. This benchmark consists of a chain of interrupt transmitter modules, whose length is parametrised by \( n \). Modules communicate through transactions, and processes synchronize with events.

Figure 5 presents the SystemC original benchmark for \( n = 1 \). To increase \( n \), one adds a transmitter module between the last transmitter and the sink module. There are always \( n + 2 \) threads (functions named compute and process) and \( n + 1 \) events (private attribute \( e \) of each module).

It is very easy to use TLM.open to verify this benchmark because the modules do not contain any dynamic members. Thus, the store and restore functions can be left empty. The state of the SystemC events and of the SystemC threads (possibly including local variables) are stored automatically.

We have also tried TLM.open on a modified version of this benchmark. The modified version uses the SC_METHOD instead of the SC_THREAD. Using SC_METHOD makes the code more difficult to read, but accelerates the simulation and reduces the memory consumption. When replacing a
SC_THREAD by a SC_METHOD, local variables have generally to be replaced by module members, and thus must be stored and restored by the user callback methods.

Among the CADP tools that can be used, there is OCIS, an interactive simulator with backtracking. Figure 6 provides a screen-shot of this tool. Also, for small values of \( n \), the LTS can be fully generated and displayed (cf. Figure 7).

Table 1 presents the results for the generation of the full LTS, using a Macbook machine with 4 GB of memory. For comparison, [27] verifies this benchmark up to \( n = 15 \) (47 seconds), and [16] verifies this benchmark up to \( n = 19 \) (8293 seconds for \( n = 19 \), 60.2 seconds for \( n = 15 \)). These results show a significant improvement compared to the previous approach based on the translation into Promela or LOTOS. The efficiency of TLM.open can be explained by two points:

1. One transition in the LTS corresponds exactly to one SystemC transition (i.e., the execution of a process between two \texttt{wait} statements.) There are no additional transitions used to mimic the behavior of the SystemC scheduler.
2. The memory size of a state is kept as small as possible, allowing the model checker to store more states.

The modified benchmark in which \texttt{SC_THREAD}s have been replaced by \texttt{SC_METHOD}s gives identical LTSes. However, the generation is three times faster, and the memory consumption is reduced: for \( n = 19 \), generating the LTS for the original benchmark requires 650 MB whereas the modified version requires only 387 MB.

In this experiment, \texttt{SC_THREAD}s are stored using hierarchical states. We have tried another
implementation using only flat states. Using the original benchmark with SC_THREADS, the flat state technique leads to an explosion of the memory consumption: 700 MB instead of 3.5 MB for $n = 12$ and the LTS generation is about 1.5 times slower.

5.2 The LusSy benchmark

The thesis [23] describes another SystemC/TLM benchmark, which is similar to the chain benchmark. The main difference is that the LusSy benchmark uses real transactions which are routed by a bus model.

It is worth noting that LusSy has a special interpretation of the timing annotations [23]. TLM.open provides an option to mimic the semantics of LusSy. This allows a greater number of schedules than the official specification, because it considers that all durations are equal.

Using TLM.open, instrumenting this benchmark with store() and restore() functions is trivial for all modules but the bus model because the whole state is contained in SC_THREAD stacks, which are automatically stored and restored. The bus model requires about 40 additional lines of code, used for storing and restoring the list of pending transactions. When using LusSy, no additional code is needed. However, LusSy does not use the bus model code. Indeed, LusSy is currently restricted to a few bus models for which a corresponding automaton model has been manually provided. Modeling a bus using automata requires more work and knowledge than adding store() and restore() functions. Therefore using LusSy is not easier than using TLM.open.

Table 2 provides the results obtained with TLM.open. It appears that TLM.open uses less memory than LusSy combined with SMV. Thus, TLM.open can verify this benchmark up to $n = 18$, whereas LusSy does not work over $n = 13$ (with a common memory limit fixed at 512 MB). For $n = 12$, TLM.open needs only two seconds where LusSy+SMV spends over one hour.
Table 2 Results for the LusSy benchmark verification using TLM.open.

<table>
<thead>
<tr>
<th>n</th>
<th>memory consumption</th>
<th>time</th>
</tr>
</thead>
<tbody>
<tr>
<td>15</td>
<td>30.8 MB</td>
<td>11.3 sec</td>
</tr>
<tr>
<td>16</td>
<td>64.6 MB</td>
<td>24.1 sec</td>
</tr>
<tr>
<td>17</td>
<td>136.1 MB</td>
<td>52.6 sec</td>
</tr>
<tr>
<td>18</td>
<td>289.8 MB</td>
<td>116.3 sec</td>
</tr>
</tbody>
</table>

5.3 Application to a timer

This subsection illustrates the features provided by TLM.open by showing how it can be applied to a simple but realistic example. We consider a timer with two registers; PERIOD and ACK.

- Writing a non-null value to PERIOD starts the timer.
- When enabled, the timer generates an interrupt periodically.
- Writing to ACK acknowledges the interrupt.
- Writing 0 to PERIOD stops the timer.

We have 4 TLM models for this timer. The first comes from the SimSoC project [19]; the second is identical to the first with a bug fix; the third and the fourth were provided respectively by an engineer and a PhD student.

In order to verify the first TLM model, which contains 80 lines of code, we had to write 17 additional lines of code to implement the `store()` and `restore()` functions. The timer verification requires the design of an environment modeling the commands generated by the embedded software. For this example, we decided to describe the environment using SystemC code. Here, the code of the environment process:

```c
void compute () {
    switch (rand(5)) {
        case 0: puts(''stop''); port->write(Timer::PERIOD_REG_OFFSET,0); break;
        case 1: puts(''start''); port->write(Timer::PERIOD_REG_OFFSET,5); break;
        case 2: puts(''ack''); port->write(Timer::ACK_REG_OFFSET,1); break;
        case 3: {
            std::ostringstream oss;
            oss <<''read_period:' <<port->read(Timer::PERIOD_REG_OFFSET);
            puts(oss.str().c_str());
            break;}
        case 4: {
            std::ostringstream oss;
            oss <<''read_ack:' <<port->read(Timer::ACK_REG_OFFSET);
            puts(oss.str().c_str());
            break;}
        case 5: puts(''wait''); next_trigger(5,sc_core::SC_MS); return;
    }
    next_trigger(sc_core::IMMEDIATE_WAKE_UP);
}
```

Note that we trigger the timer with only one specific period. Explicit model-checking does not permit the verification of this model for all values of the period. Thus, we have to assume that the presence of bugs does not depend on this particular period.
The last line uses a special feature of TLM.open: the process yields back to the scheduler but remains eligible. This statement is similar to the yield() statement introduced in [15]. The rationale of this statement is to break critical sections that would exist in the model but not in the real system.

Firstly, we applied on-the-fly property checking. The property checker of CADP revealed an error in the first version: for some particular schedules, the timer could generate an interrupt after it was stopped. A counter-example was automatically shown, allowing us to fix the bug. Another minor bug was found in the third version.

Secondly, we applied equivalence checking. We generated the LTS of each timer TLM model, we hid the internal transitions and we minimized them according to the branching equivalence. We got the proof that the second and the forth version are bisimilar modulo branching equivalence. It means that if one contains an error, the other contains the same error. Obviously, the first and third versions are not bisimilar, since they contain distinct errors.

5.4 Application to a basic system

In order to evaluate the behavior of TLM.open on a system made of many modules, we studied a basic system that was originally used for practical work. This system was implemented on FPGA. It contains a MicroBlaze processor, a VGA controller, plus the usual and mandatory peripherals: bus, memory, timers, interrupt controller. In the SystemC/TLM model, the user can model the processor, by either using a native wrapper or an instruction set simulator (ISS). The embedded software compute images and manage the configuration of the peripherals.

For the validation of the embedded software, we decided to use the native wrapper instead of the ISS. On the one hand, there is nothing that prevents us using the native wrapper for this software (i.e., no inline assembly code and no dynamic code loading). On the other hand, using the ISS would multiply the number of states: 1 state per binary instruction with the ISS instead of one state per explicit synchronization point with the native wrapper.

Thus, we have instrumented all modules with store() and restore() methods. Then, we changed the output functions so that traces are added to LTS labels instead of sent to the terminal. Finally, we made some simplifications: 1. we have disconnected the graphical library used by the VGA module, which means that during model checking we do not display the simulated VGA screen. 2. We have simplified the TLM protocol so that it no longer uses a transaction pool because the transaction pool mechanism is only a trick to make simulations a little faster.

Using TLM.open, we generated the LTS corresponding to this basic SystemC model for one embedded software. Using the bcg_min tool of CADP, the LTS can be reduced to a minimal LTS. This minimal LTS is small enough to be read by human. By observing this LTS, we notice that in some cases the processor was receiving an interruption before any other module raised one. The rationale was a missing dont_initialize() in the SystemC code. Because the occurrence of this error depends on the scheduling, this bug had not been noticed before we used TLM.open. After fixing this bug, we generated the minimized LTS again. This second LTS is represented by Figure 8.

Finally, we tried to add some errors in the embedded software, such as changing an initial value or disabling a register write in order to verify that all errors can be discovered during model checking. For each error, we got either a SystemC error message (such as assertion failure coming from the TLM code) during LTS generation, or an LTS that was not equivalent to the reference once (equivalence checked using the CADP tool bisimulator).
6 Conclusion

We presented a new framework for the verification of SystemC/TLM programs. Our new SystemC/TLM front-end avoids the need to translate the whole SystemC/TLM program into another language. Compared to approaches based on manual translation, the verification using TLM.open is much simpler: there are less lines of code to write and the engineers do not need to learn a new modeling language. Moreover, TLM.open allows better scaling than previous works. Thanks to the numerous tools of CADP, it is now possible to check complex properties and to test the equivalence of two SystemC/TLM programs.

Note that TLM.open can be used with pure SystemC programs also (i.e., programs not using TLM). The rationale of calling our tool TLM.open instead of SystemC.open is related to the abstraction level: the CADP verification toolbox is optimized for asynchronous processes. SystemC/TLM models use asynchronous processes, but SystemC programs modeling a system at a lower level of abstraction use synchronous processes. In order to verify synchronous processes, symbolic modal-checker based on BDD or SAT, are in general more efficient than CADP. Thus, TLM.open can be used for pure SystemC programs, but is not likely to be the most efficient tool.

As explained in [7], the most difficult task when verifying a SystemC/TLM program is to extract an abstract model that is simple enough to be formally verified. Thus, the main further work is to integrate TLM.open in the design flow in such a way that this task becomes simple and safe. Additionally, it would help to automatize the generation of the store() and restore methods.

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Randomized Caches Considered Harmful in Hard Real-Time Systems

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Abstract
We investigate the suitability of caches with randomized placement and replacement in the context of hard real-time systems. Such caches have been claimed to drastically reduce the amount of information required by static worst-case execution time (WCET) analysis, and to be an enabler for measurement-based probabilistic timing analysis. We refute these claims and conclude that with prevailing static and measurement-based analysis techniques caches with deterministic placement and least-recently-used replacement are preferable over randomized ones.

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1 Introduction
Recent work has promoted the use of randomized caches in hard real-time systems [4, 20, 22, 23, 21, 7, 5, 25]. Along with randomized microarchitectures, this line of work proposes static probabilistic timing analysis (SPTA) and measurement-based probabilistic timing analysis (MBPTA). Caches are a major challenge in the timing analysis of traditional, deterministic microarchitectures. A key feature of randomized microarchitectures are caches with randomized placement and replacement. Such caches have been claimed to drastically reduce the amount of information required by WCET analyses. To quote Kosmidis et al. [23]: “The key benefit of embracing PTA (probabilistic timing analysis) is that execution timing becomes dramatically less dependent on execution history, with drastic reduction in the amount of information required to obtain tight WCET estimates in comparison to other timing analysis approaches.”

In this paper, we critically assess these claims both in the context of static and measurement-based analysis. Specifically, we compare the precision of static cache analyses for caches with least-recently-used (LRU) replacement and with randomized replacement provided the same amount of information, i.e. the information stated to be sufficient for the analysis of randomized caches. Among deterministic caches we restrict our attention to those with LRU replacement, as it is widely considered to be the most predictable replacement policy, and it has been demonstrated to be efficiently implementable [1, 8]. Our analysis demonstrates that, with simple, state-of-the-art analyses, deterministic LRU replacement is preferable over random replacement. We also observe that, with its current restrictions, MBPTA is equally applicable to LRU caches as it is to randomized ones.

Regarding random placement, we show that it is impossible to assign non-zero hit probabilities to individual memory accesses that are independent of the outcome of other accesses. This means that caches with random placement are not amenable to the prevailing SPTA approach that relies on independence, as execution time profiles (ETPs) of individual instructions are convolved.
Finally, we provide a class of memory access sequences that is problematic for MBPTA under random placement. On these sequences, which may occur in practice due to loops, MBPTA either fails, is incorrect, or highly imprecise.

We provide the necessary background about probabilistic timing analysis in Section 2. Then, we introduce deterministic and randomized caches in Section 3. We assess the suitability of random placement and replacement for use in hard real-time systems in Sections 4 and 5. Finally, we briefly summarize our findings in Section 6.

2 Static and Measurement-Based Timing Analysis

The goal of static and measurement-based timing analysis for deterministic microarchitectures is to compute tight upper bounds on the worst-case execution times (WCET) of programs. The goal of timing analyses for randomized microarchitectures is slightly different: in such microarchitectures, very high execution times are possible, but—hopefully—only with a very low probability. Thus, timing analyses for such microarchitectures compute exceedance functions. These exceedance functions determine upper bounds on the probability of exceeding any given execution time. From such a function, and a probability threshold $p$, an execution time can be obtained that is exceeded only with a probability of e.g. $p = 10^{-12}$.

2.1 Static Probabilistic Timing Analysis

The de facto standard approach to static timing analysis (STA) for deterministic microarchitectures divides analysis into two main parts [31]:

1. **Low-level analysis**, which determines execution-time bounds for basic blocks (or other small contiguous program fragments) based on an accurate model of the underlying microarchitecture.
2. **Path-level analysis**, which determines an upper bound on the execution time of the program as a whole based on constraints on the control flow, e.g. loop bounds, and the execution-time bounds for basic blocks determined by low-level analysis.

A critical assumption of this approach is that the bounds obtained for a basic block during low-level analysis hold for all possible execution histories leading to the respective basic block. As execution times may depend heavily on the execution history, low-level analysis is often made context sensitive, e.g. by distinguishing the first iteration of a loop from the following ones.

While so far less studied and thus less developed, static probabilistic timing analysis (SPTA) follows a similar approach [4]:

1. For each instruction in the program, an execution time profile (ETP), i.e., a discrete probability distribution over the instruction’s possible execution times, is derived. This step corresponds to the low-level analysis in STA.
2. To arrive at an ETP for a sequence of instructions the ETPs of all instructions in the sequence are combined by convolution. If multiple different execution paths are possible, their ETPs can be merged conservatively [4]. This roughly corresponds to path-level analysis in STA. From an ETP, a corresponding exceedance function can then be determined easily.

A critical assumption for SPTA to be sound is that the ETPs derived in step one are independent of each other. Only if they are independent, can they be soundly combined by convolution to arrive at an ETP for a sequence of instructions.

2.2 Measurement-based Probabilistic Timing Analysis

Measurement-based probabilistic timing analysis (MBPTA) derives exceedance functions for the execution time of a program from measurements. MBPTA as described by Cucu et al. [5] is
based on Extreme Value Theory (EVT). In this approach, a series of end-to-end execution-time measurements is performed. The measurement results are used to estimate the parameters of an extreme value distribution, the Gumbel distribution. Measurements and estimation of the Gumbel distribution are interleaved until the distribution is considered to have converged [5]. The thus obtained Gumbel distribution then immediately induces an exceedance function.

Applicability of this approach relies on two assumptions:

1. The execution-time measurements can be modeled by independent and identically-distributed (i.i.d.) random variables.
2. The maximum of a sample of these i.i.d. random variables converges in distribution to the Gumbel distribution.

To satisfy the first assumption, Cucu et al. [5] propose a number of changes to the microarchitecture to eliminate the dependence of execution times on input data. For instance, input-dependent memory accesses must bypass the cache. They also initially limit their approach to single-path programs, which they later [5] show how to relax.

The satisfaction of the second assumption is validated during the analysis of a particular program by statistical tests.

3 Deterministic and Randomized Caches

Caches are fast but small memories that store a subset of the main memory’s contents to bridge the latency gap between the CPU and main memory. To reduce management overhead and to profit from spatial locality, data is not cached at the granularity of words, but at the granularity of so-called memory blocks. To this end, main memory is logically partitioned into the set of equally-sized memory blocks \( B = \{0, \ldots, n\} \). Blocks are cached as a whole in cache lines of the same size. The size of a memory block varies from one processor to another, but is usually between 32 and 128 bytes.

When accessing a memory block, the cache logic has to determine whether the block is stored in the cache (a cache hit) or not (a cache miss). To enable an efficient look-up, each memory block can only be stored in a small number of cache lines. For this purpose, caches are partitioned into equally-sized cache sets. The size of a cache set is called the associativity \( k \) of the cache.

The placement policy determines the cache set a memory blocks maps to. In Section 3.2 we describe common deterministic and randomized placement policies.

Since caches are usually much smaller than main memory, a replacement policy must decide which memory block to replace upon a cache miss. In Section 3.1 we describe common deterministic and randomized replacement policies.

The performance of a cache depends on the temporal and spatial locality of the memory accesses. In Section 3.3, we describe two notions of locality that are approximated by state-of-the-art static (probabilistic) cache analyses.

3.1 Replacement Policies

Usually, replacement policies treat each cache set separately, so that accesses to a particular cache set do not influence replacement decisions in other cache sets. While exceptions to this rule exist, they have been identified as particularly unsuitable for real-time systems [17]. Thus, in the following, we only consider replacement policies treating each cache set separately.

Well-known deterministic replacement policies in this class are least-recently used (LRU), used in various Freescale processors such as the MPC603E and the TriCore17xx, as well as the recent Kalray MPPA 256; pseudo-LRU (PLRU), a cost-efficient variant of LRU, used in the
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Freescale MPC750 family and multiple Intel microarchitectures; most-recently used (MRU), also known as not most-recently used (NMRU), another cost-efficient variant of LRU, used in the Intel Nehalem; first-in first-out (FIFO), also known as ROUND ROBIN, used in several ARM and Freescale processors such as the ARM922 and the Freescale MPC55xx family.

Logically, LRU orders cached memory blocks by the recency of their last use, from most- to least-recently-used. Upon a miss, the least-recently-used block is evicted. Among deterministic policies, LRU is generally accepted as the most predictable policy [28]. Thus, in the following, among deterministic policies, we restrict our attention to LRU, which has been shown to be efficiently implementable [1], and which is used in the Kalray MPPA 256 [8] for predictability.

Quiñones et al. [25, 20, 21, 4] have promoted the use of randomized caches in real-time systems. They have focused on a policy, which we will call Random in the following, that was introduced by Belady [2]. Upon a miss, Random chooses the block to evict randomly and uniformly among the \( k \) cache lines of the cache set. Thus, upon a miss, a cached block—in the cache set that the accessed block maps to—is evicted with probability \( \frac{1}{k} \). This policy is also referred to as evict-on-miss in the literature [7].

Several commercial processors are claimed to employ random replacement, e.g. the ARM720T, the ARM940T, the ARM11xx, and the Freescale MPC7450. However, most processor documents are inconclusive about the exact meaning of “random”. Such caches could be based on hardware random number generators that generate random numbers from a physical process, such as thermal noise, or they could employ deterministic pseudo-random number generators.

The well-documented MPC7450 [18] allows to choose between two random policies for its second-level caches [10]: “The simpler one uses a modulo counter that is incremented on each clock cycle and whose value determines the cache line to replace.”

To achieve independence between cache-miss probabilities, Cazorla et al. [4] have also proposed the evict-on-access policy, which evicts a block uniformly at random upon each memory access, rather than upon each cache miss. In the following, we limit our attention to Random, as it provably dominates evict-on-access in terms of the induced exceedance function on any workload.

Randomized policies have been studied extensively in the context of competitive analysis [3]. Policies such as Mark and Equitable have been shown to have smaller competitive ratios than any deterministic policy. These results concern the expected performance of a policy, rather than the performance achieved with high probability, which would be of greater interest in the hard real-time setting. However, recent results by Komm et al. [19] suggest that randomized policies can also be shown to be competitive “with high probability”.

### 3.2 Placement Policies

A placement policy can be formalized as a mapping from memory blocks to cache sets: \( \text{place} : \mathcal{B} \to \{0, \ldots, s - 1\} \), where \( s \) is the number of cache sets.

The most common deterministic placement policy for set-associative caches is modulo placement: \( \text{place}_\text{modulo}(b) = b \mod s \). The number of sets \( s \) is usually a power of two, so that the cache set of a block is simply determined by its \( \log s \) least significant bits.

In random placement the mapping \( \text{place}_\text{random} \) from memory blocks to cache sets is chosen randomly from the set of all mappings \( \mathcal{B} \to \{0, \ldots, s - 1\} \). For static-analysis purposes it is convenient if the mapping is chosen uniformly at random from this set. Then, the probability of block \( b \) mapping to cache set \( t \), \( P(\text{place}_\text{random}(b) = t) \), is \( \frac{1}{s} \). Note, that \( \text{place}_\text{random} \) needs to be fixed for the entire execution of a program. Otherwise, it would not be possible to locate memory blocks that were cached earlier under a different mapping.\(^1\)

\(^1\) Changing the mapping at runtime requires either flushing of cache contents or their migration.
Kosmidis et al. present an approximation of random placement in hardware \cite{20, 21} based on a parametric hash function and in software \cite{22} on top of conventional caches with deterministic placement. The crucial difference between the hardware and the software solution is that the software solution can only randomize mapping at the granularity of “memory objects”, i.e., memory entities normally stored in consecutive memory addresses such as functions, basic blocks, or arrays.

### 3.3 Notions of Locality

Caches rely on locality in the memory access sequences generated by programs. Different ways of capturing locality have been proposed over time. Two notions of locality particularly relevant to LRU and Random replacement are the reuse distance and the stack distance of a memory access.

The reuse distance of a memory access to block $b$ is the number of memory accesses between the current and the previous access to block $b$. The first access to a block has reuse distance $\infty$. As an example, we have annotated each memory access in the following sequence with its reuse distance:

$$a^{\infty}, b^{\infty}, b^0, a^2, c^{\infty}, d^0, c^3, b^5, a^5.$$  

In contrast to the reuse distance, the stack distance of an access to block $b$ is defined as the number of distinct memory blocks accessed between the current and the previous access to block $b$. The stack distance of a block is sometimes also referred to as the age of the block. The first access to a block has stack distance $\infty$. In the sequence from above, the stack distances are as follows:

$$a^{\infty}, b^{\infty}, b^0, a^1, c^{\infty}, d^0, c^1, b^3, a^3.$$  

Note that the stack distance of any access is less than or equal to its reuse distance.

We will see later how hit and miss probabilities of a memory access can be given based on its reuse and stack distance for both randomized and deterministic caches.

### 3.4 Cache Analysis in Static (Probabilistic) Timing Analysis

Cache analysis is an important part of low-level analysis. In STA, its purpose is to classify memory accesses in the program as either definite hits or definite misses. Sometimes, an access may result in a hit or a miss depending on the execution history leading to the access. As a consequence of such inherent uncertainty or uncertainty due to analysis imprecision, cache analysis may also classify an access as “unknown”. Due to timing anomalies \cite{24, 29} it is not always safe to simply assume a cache miss in case of uncertainty.

Similarly, in SPTA \cite{4}, a probability needs to be attached to the hit and the miss case for each memory access. Current SPTAs assume microarchitectures in which hits and misses have a fixed, context-independent cost and thus cache-related timing anomalies may not occur. As a consequence, it is sufficient to determine a lower bound $h$ on the hit probability of an access, which induces an upper bound of $1 - h$ on its miss probability. Together with hit and miss latencies $\text{hit latency}$ and $\text{miss latency}$ we get the following ETP for a memory instruction with hit probability $h$:

$$\begin{pmatrix} \text{hit latency} & \text{miss latency} \\ h & 1 - h \end{pmatrix}.$$  

The issue of timing anomalies in the pipeline is orthogonal to that of using deterministic or randomized caches. In order not to mix the two issues, we compare deterministic\footnote{Caches with LRU replacement do not exhibit timing anomalies.} and randomized caches in the context of SPTA, i.e., in terms of deriving lower bounds on the hit probability of a memory access.
4 Deterministic versus Random Replacement in Fully-Associative Caches

4.1 In Static Probabilistic Timing Analysis

In a cache with LRU replacement, an access \( b \) with stack distance \( sd(b) \) less than the associativity \( k \) is a hit, otherwise it is a miss:

\[
P(hit_{LRU}(b)) = \begin{cases} 
1 & : sd(b) < k \\
0 & : sd(b) \geq k 
\end{cases}
\]  
(1)

Note, that the hit probabilities of different memory accesses in a sequence are independent. Static cache analyses thus determine upper bounds on the stack distance of each memory access to guarantee cache hits. Such analyses are called must analyses [9]. Analogously, may analyses [9] determine lower bounds on stack distances to guarantee cache misses.

In the case of a fully-associative cache there are two challenges for may and must analyses:

1. The initial state of the cache is unknown. Thus, must analyses conservatively assume an upper bound \( > k \) on the stack distance of any block at program start. Similarly, may analyses assume a lower bound of \( 0 \).
2. Logically, memory accesses are at the granularity of words, not memory blocks. Thus a value analysis needs to determine for pairs of memory accesses whether they refer to the same memory block or not.\(^3\) This is trivial for instruction caches, but may be very hard for input-dependent data accesses.

In a cache with Random replacement the situation is different. In contrast to LRU, the hit probability of an access cannot be given purely in terms of its stack or reuse distance. Zhou [32] observes that the hit probability of an access \( b \) to a block that has been accessed before is

\[
P(hit_{Random}(b)) = \left(1 - \frac{1}{k}\right)^m
\]  
(2)

where \( m \) is the number of cache misses between access \( b \) and the previous access to the same memory block. Clearly \( m \) is bounded from above by \( b \)'s reuse distance \( rd(b) \), so

\[
P(hit_{Random}(b)) \geq \left(1 - \frac{1}{k}\right)^{rd(b)}
\]  
(3)

This formula correctly underestimates the hit probability of an individual access. Unfortunately, however, hit probabilities computed with the formula above are not independent of each other. Thus, the convolution of corresponding ETPs may underestimate the probability of observing a given number of misses. Consider, e.g., the access sequence \( a, b, c, a, b, c \) and a cache with associativity 2. Clearly, at least one miss must occur on the final three accesses of the sequence, as the first access to \( c \) will evict either \( a \) or \( b \). Yet, the convolution of the ETPs obtained from Equation (3) yields a non-zero probability of having no misses on those three accesses, because the hit probability of each individual access is greater than zero\(^4\). As a consequence, the probability of observing four or more cache misses on the entire sequence, which is 1, would be underestimated.

---

\(^3\) Note, that it is not necessary to determine which memory block is referred to by a memory access. It is sufficient to determine for pairs of accesses whether they refer to the same block or not. This is exploited by relational cache analysis [16].

\(^4\) According to Equation (3), it is \( \frac{1}{4} \).
Davis et al. [7] provide the following formula, which bounds the hit probability of an access $b$, independently of whether preceding accesses hit or miss, in terms of the reuse distance of $b$:

$$P(\text{hit}_{\text{Random}}(b)) \geq \begin{cases} (1 - \frac{1}{k})^{rd(b)} : rd(b) < k \\ 0 : rd(b) \geq k \end{cases}$$  \hspace{1cm} (4)

The more optimistic formula for hit probabilities given in [20] has been refuted in [6]. Based on the reuse distance, the above formula is the most precise hit probability that holds independently of the outcome of previous memory accesses.

For associativity 8, Figure 1 illustrates the hit probabilities of LRU and Random in terms of reuse distances. Remember that by definition the stack distance $sd(b)$ of an access is less than or equal to its reuse distance $rd(b)$. With this in mind, comparing the hit probabilities for LRU and for Random from Equations 1 and 4, we make the following two observations:

- **Observation 1.** With the same information about an access sequence, i.e. upper bounds on the reuse distances of accesses, the hit probabilities for LRU are always greater than or equal to the hit probabilities for Random.
- **Observation 2.** In case of LRU, and in contrast to Random, current cache analyses can profit from bounding stack distances, which can be arbitrarily lower than reuse distances.
- **Conclusion 1.** With simple, state-of-the-art analysis methods, LRU replacement is preferable over Random replacement in static (probabilistic) timing analysis.

In general, we note that the state (or the state distribution in case of randomized caches) of any cache is a function of the history of memory accesses. This holds independently of whether the cache is deterministic or randomized. More precisely, the state of any fully-associative cache depends, at least, on the suffix of the history of memory accesses containing $k$ distinct blocks, where $k$ is the associativity of the cache. Among all policies, randomized or deterministic, the state of an LRU-controlled cache depends on the shortest suffix of the access history. Thus, LRU requires the least information about the access history to fully determine its state [28].

**Other Deterministic Policies: FIFO, PLRU, and MRU.** Common deterministic policies such as FIFO, PLRU, and MRU have been found to be less predictable than LRU [28]. In contrast to LRU, the exact hit probability of an access cannot be determined in terms of stack or reuse
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distances for these policies. The best lower bounds on hit probabilities that can be given based on stack distances for FIFO, PLRU, and MRU are:

\[
P(\text{hit}_{\text{FIFO}}(b)) \geq \begin{cases} 1 & : sd(b) < 1 \\ 0 & : sd(b) \geq 1 \end{cases} 
\]

(5)

\[
P(\text{hit}_{\text{PLRU}}(b)) \geq \begin{cases} 1 & : sd(b) < \log_2 k + 1 \\ 0 & : sd(b) \geq \log_2 k + 1 \end{cases} 
\]

(6)

\[
P(\text{hit}_{\text{MRU}}(b)) \geq \begin{cases} 1 & : sd(b) < 2 \\ 0 & : sd(b) \geq 2 \end{cases} 
\]

(7)

Static (probabilistic) timing analyses that are based purely on stack (or reuse) distances thus yield worse results under FIFO than under Random, MRU, PLRU, or LRU. MRU and PLRU are incomparable with Random. Depending on the benchmark, and the resulting distribution of reuse distances, MRU and PLRU may yield better results than Random and vice versa.

It should be noted that there are access sequences (and programs that generate such sequences) on which Random outperforms LRU and other deterministic policies with a very high probability. A prime example for such sequences are so-called “payroll sequences”, i.e. sequences of the form \((a_1, \ldots, a_{k+1})^*\), where \(k\) is the associativity of the cache. On such sequences LRU incurs cache misses only. Similar sequences can be constructed for every deterministic policy.

To profit from Random replacement in such cases, more sophisticated analysis techniques are required. Such analyses will likely have to derive conditional hit probabilities and combine results for individual memory accesses in a different way than convolution, which is not required in case of LRU. Similarly, sophisticated static analyses have recently been proposed for FIFO [11, 12, 15], PLRU [13], and MRU [14]. Yet, while being more complex than Ferdinand’s analysis for LRU [9], they still do not quite achieve the same level of precision.

**Stack Distance versus Reuse Distance.** As an example where reuse and stack distances may differ a lot, consider instruction accesses following the execution of a small, nested loop:

```python
x = 0
y = 0
for i in [1, 1000]:
    for j in [1, i]:
        x = x+1
    y = y+1
```

Assume for simplicity that the instructions implementing each line of the program occupy exactly one memory block. Thus, the program’s instructions occupy exactly six memory blocks. Then, except for the first access, the stack distance of each instruction access for \(y = y+1\) is three. Yet, the corresponding reuse distance depends on the value of \(i\). In the last iteration of the outer loop the reuse distance of the instruction access for \(y = y+1\) is 2001.

In practice, the relation between reuse and stack distances likely varies strongly within benchmarks and from one benchmark to another. Unfortunately, we were not able to find empirical data concerning their relation, with one exception: Sen and Wood [30] plot the distribution of reuse distances for accesses of a given stack distance (Figure 3 in [30]) for an online transaction processing application.

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5 This follows immediately from the competitiveness of the respective policies relative to LRU [26].
4.2 In Measurement-Based Probabilistic Timing Analysis

MBPTA derives WCET estimates from a series of execution-time measurements. Cache performance depends on the initial state of the cache and on the sequence of memory accesses provided to the cache. Cucu et al. [5] flush the cache upon program start and eliminate all input-dependent memory accesses from the program. Thus, on a given path through the program, the sequence of memory accesses is the same for different program inputs. Then, a number of end-to-end execution-time measurements is performed on each program path. By nature of the approach, the analysis results apply only to those program paths for which measurements have been performed.

In such a scenario, i.e., flushed cache and no input-dependent memory accesses, the requirements of MBPTA, namely independence and identical distribution of execution times on a given program path, are also met by a conventional cache with LRU replacement: the cache behavior will be identical on each measurement; it follows a degenerate probability distribution. Independence is thus trivial. In fact, the same argument applies to any deterministic cache replacement policy.

Thus, under the conditions described above, a single measurement will reveal the worst case—in terms of cache performance. This compares with having to perform hundreds of measurements in case of a randomized cache [5].

> Conclusion 2. Deterministic replacement yields more efficient MBPTA than Random replacement.

For LRU, the empty state is the worst-case initial state for any memory access sequence [27]. Therefore, measurements obtained starting with a flushed cache yield upper bounds on the number of cache misses under any initial cache state for LRU. In this case, flushing the cache would only be required during the measurement-based analysis and could in principle be disabled during normal operation, assuming the microarchitecture features no timing anomalies [24, 29].

5 Deterministic versus Random Placement in Set-Associative Caches

5.1 In Static Probabilistic Timing Analysis

In a set-associative cache with $s$ cache sets, the placement policy partitions the stream of memory accesses into $s$ substreams, each of which is processed by one of the $s$ cache sets. For set-associative caches, it is convenient to define the reuse and the stack distance of an access based on the subsequence the access belongs to. In other words, the reuse distance of an access to block $b$ is the number of memory accesses between the current and the previous access to block $b$ within the same cache set. Let the stack distance be defined analogously for set-associative caches.

Then we get the same hit probabilities in terms of stack and reuse distance for LRU and Random as in case of a fully-associative cache.

The additional difficulty in static cache analysis with deterministic placement is thus to determine which memory accesses map to the same cache set. This is again trivial for instruction accesses, but may be very difficult for data accesses. Note, however, that it is not required to determine the absolute cache set an access maps to, as demonstrated by relational cache analysis [16].

Randomized placement [20, 21] promises to reduce the analysis effort for set-associative caches, as two memory blocks will only collide in the cache with a certain probability. If the placement function is chosen randomly from a uniform distribution over all possible placement functions, then the probability of any two blocks to map to the same cache set is $\frac{1}{s}$. Based on this assumption, Kosmidis et al. [20] derive the following hit probability for a direct-mapped cache in terms of the
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stack distance\(^6\) of an access:

\[ P(\text{hit}_{\text{Random}}(b)) = \left( 1 - \frac{1}{s} \right)^{sd(b)} \]  

(8)

This formula is correct. However, as in the case of Zhou’s formula for random replacement, hit probabilities determined in this way are not independent. Consider the following access sequence:

\[ a, b, a, b, a, b, a, b, a, b \]

As the placement is chosen randomly at program start and does not change during program execution, there are only two possibilities: either \(a\) and \(b\) systematically collide in the cache or they do not. They collide with probability \(\frac{1}{s}\). Thus with a probability of \(\frac{1}{s}\) all ten memory accesses will be cache misses, and with a probability of \(1 - \frac{1}{s}\) only two (compulsory) misses will occur. In the example, each access’s miss probability is \(1 - (1 - \frac{1}{s})^1 = \frac{1}{s}\). Assuming independence of these miss probabilities would incorrectly yield a probability of \(\frac{1}{s^8} \ll \frac{1}{s}\) of observing ten misses.

Unfortunately, no non-zero hit probability can be assigned to an access based on its stack distance that is independent of other whether previous accesses resulted in hits or misses. Yet, independence is required by the current SPTA approach. To see this, consider arbitrarily long sequences of the form \(a, b, a, b, a, b, \ldots\) No matter how long the sequence, with a probability of \(\frac{1}{s}\) all accesses will miss in a direct-mapped cache. For any non-zero hit probability \(p\) assigned to each individual access there is a length \(n\) of the sequence, such that the convolution of the ETPs based on \(p\) will underestimate the probability of incurring \(n\) misses.

Observation 3. In case of random placement, no mutually independent hit probabilities greater than zero can be assigned to individual memory accesses with stack distances greater than zero.

Observation 3 immediately implies the following conclusion:

Conclusion 3. Random placement requires complex static analyses that take into account conditional hit probabilities. Random placement is thus not amenable to current analysis approaches.

5.2 In Measurement-Based Probabilistic Timing Analysis

As we have seen in the previous section, with random placement there are cases in which we observe very few misses with a high probability \(p\) and very many misses with a low probability \(1 - p\), with no cases in between the two extremes.

If \(p\) is sufficiently close to 1, MBPTA is unlikely to ever observe the case of very many misses. Then, its observations are indistinguishable from a case in which many misses are in fact impossible. Consider as an example\(^7\) the same sequence \(\sigma_{\text{slow}} = a, b, a, b, a, b, \ldots\) as above, which may be generated by a loop, and a very large direct-mapped cache with \(s = 10^6\) cache sets. The probabilities of the two possible execution times are depicted in Figure 2.

Even \(10000 = 10^4\) measurements will only reveal the worst case with a probability of \(1 - (1 - 1/10^6)10^4 < 1\%\). In other words, with a probability greater than 99\%, all measurements will yield exactly two cache misses, and thus the sequence would be indistinguishable from the sequence \(\sigma_{\text{fast}} = a, b, b, b, b, \ldots\), which will yield exactly two misses independently of the placement.

---

\(^6\) Here, \(sd(b)\) refers to \(b\)’s stack distance among all memory accesses, i.e., not to its stack distance among blocks mapping to the same cache set.

\(^7\) While this example is slightly constrained, due to the unrealistically high number of cache sets, the sequence \(a, b, c, d, a, b, c, d, \ldots\) in a 3-way set-associative cache with a more realistic \(s = 10^2\) leads to similar results, yet is more difficult to analyze precisely.
MBPTA is used to estimate execution times that are only exceeded with a very low probability, such as $10^{-12}$, ideally without performing $10^{12}$ measurements. In our example, with a probability of $10^{-6} \gg 10^{-12}$, the execution time for sequence $\sigma_{\text{slow}}$ will be very high, as all memory accesses will result in cache misses.

MBPTA bases its estimates solely on measurement results. It would thus generate the same execution-time distribution for programs that generate the sequences $\sigma_{\text{slow}}$ and $\sigma_{\text{fast}}$ with a high probability. In such a situation there are three possibilities:

1. MBPTA correctly estimates the execution-time distribution for the sequence $\sigma_{\text{slow}}$.
2. MBPTA incorrectly underestimates the execution-time distribution for the sequence $\sigma_{\text{slow}}$.
3. Based on the measurement results, the statistical tests in MBPTA reject such programs.

The two latter cases are clearly undesirable. In the first case, MBPTA’s estimate would have to be the same for the sequence $\sigma_{\text{fast}}$. However, a correct estimate for $\sigma_{\text{slow}}$ is necessarily extremely pessimistic for $\sigma_{\text{fast}}$, which never exhibits more than two cache misses. This leads us to our final conclusion:

> Conclusion 4. Random placement is not suitable for MBPTA.

6 Summary

We have critically assessed the suitability of randomized caches for use in hard real-time systems. We observe that when used in SPTA, state-of-the-art cache analyses deliver better hit probabilities for LRU than for Random replacement with the same amount of information. With the restrictions currently imposed upon the use of randomized caches in MBPTA, i.e., no input-dependent memory accesses, deterministic caches with LRU replacement can also be safely employed in MBPTA. This comes with the additional benefit of requiring only a single measurement to identify the worst-case cache performance.

Further, we have shown that non-trivial hit probabilities under random placement are not independent and can thus not be safely combined by convolution. We have also identified simple access sequences, which may be generated by simple loops, on which MBPTA must—by construction—be either unsound, extremely pessimistic, or fail to produce an estimate at all.

Despite the negative results obtained in this paper, we believe that randomization might have a place in microarchitectures for real-time systems. A benefit over deterministic microarchitectures may be an increase in robustness. It is future work to rigorously analyze the benefits of randomization in this direction.

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