

# A New Approach for Optimal Implementation of Multi-core Reconfigurable Real-time Systems

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**Abstract:** This paper deals with a multi-core reconfigurable real-time system specified with a set of implementations, each of which is raised under a predefined condition and executes multiple functions which are in turns executed by threads. The implementation as threads generates a complex system code. This is due to the huge number of threads and the redundancy between the different implementations which may lead to an increase in the energy consumption. Thus we aim in this paper to optimize the system code by avoiding the redundancy between implementations and reducing the number of threads while meeting all related real-time constraints. The proposed approach adopts mixed integer linear programming (MILP) techniques in the exploration phase in order to provide a feasible task model. An optimal reconfigurable POSIX-based code of the system is manually generated as an output of this technique. An application to a case study and performance evaluation confirm and validate the expected results.

## 1 INTRODUCTION

A real-time system has to respond to externally generated input stimuli within a finite and specified delay (Burns and Wellings, 2009). Such system may have many implementation scenarios, the transition from an implementation to another is called reconfiguration. Reconfiguration refers to the architectural or behavioral modifications of a software system during its execution (Polakovic et al., 2007) to meet user requirements. Currently, some real-time systems such as automotive electronics, avionics, telecommunications, and consumer electronics become more complex and need more computational power. Thus, the necessity for multi-core architecture is a common answer. The multi-core technology allows increasing the processor clock frequency, which is limited by available instruction-level parallelism and leads to challenging power requirements (Geer, 2005). This paper deals with multi-core reconfigurable real-time systems.

One challenge during the development of multi-core reconfigurable real-time systems is to ensure an appropriate partitioning and scheduling of the ap-

plicative functions across the target platform such that the timing constraints are met. In that context, different scheduling policies have been proposed in the literature (Khan and Hafiz, 2014) (Lakshmanan, 2011). Existing multi-core scheduling policies can be classified into three different classes: the partitioned, the global and semi-partitioned approach. The partitioned scheduling allows to choose a core for all tasks and then runs a local scheduler on each core (i.e., off-line scheduling). However, the global scheduling allows to choose a task and to assign it to one of the cores (i.e., on-line scheduling). As opposed to the partitioned approach, different instances of the same task can execute on different cores. The semi-partitioned approach presents an improvement of the partitioning scheduling allowing the controlled tasks migration. It is a hybrid between partitioned and global scheduling (Lakshmanan, 2011). In this paper, we adopt a partitioned approach because it is easier to implement and to analyze. Also, it allows no task migration, thus has low runtime overheads (Funk and Baruah, 2005). For the three multi-core scheduling approaches, several scheduling algorithms have been proposed such as Rate Monotonic RM (Liu and Layland, 1973) which

will be adopted for tasks scheduling in this paper.

The multi-core introduces additional challenges that are still difficult to deal with in real world industrial domains. Indeed, the huge number of tasks exhibits a high complexity by increasing the energy consumption, invoking many redundancies between the different implementations, and producing a complex system code. Thus, we propose a multi-objective optimization approach that minimizes both the energy consumption and the number of tasks in order to reduce the redundancy between the implementation sets. Such optimization may reduce the time overhead and the complexity of the generated code. At the specification level, the developer defines the function sets, the condition sets, and the core sets. At the design level, this approach (i) generates the implementation sets, (ii) affects the functions to the tasks sets which are in turns assigned to the core sets, and (iii) generates a feasible and optimized task model by using the mixed integer linear programming (MILP) formulation. Finally, at the implementation level, each task is transformed to a thread to execute the application functions.

The originality of the proposed approach follows from the fact that (i) it deals with the multi-core (i.e., partitioning), reconfiguration, and real-time problems simultaneously, and (ii) it proposes a multi-objective optimization to minimize both the energy consumption and the number of tasks.

The rest of the paper is organized as follows. Section 2 presents the state of the art on multi-core reconfigurable real-time systems. In Section 3, we give the system formalization. We present the proposed approach in Section 4. In Section 5 we present the Global Positioning System (GPS) case study which is considered to evaluate the proposed approach. Finally, we summarize our work and discuss future directions in Section 6.

## 2 RELATED WORKS

In the area of real-time multi-core systems, some existing works focus on the synthesis problem (Wang et al., 2016), (Yehia et al., 2011), (Geismann et al., 2017). In (Wang et al., 2016), the authors propose a formalization of periodic tasks adapted to engine control applications in multi-core automotive systems. The work reported in (Yehia et al., 2011) presents a system level synthesis approach for multi-core system architectures from Task Precedence Graphs (TPG) models. In (Geismann et al., 2017), the authors present an approach for a semi-automatic synthesis of models into a deterministic scheduling that respects

real-time requirements for multi-core systems. Despite the importance of above related synthesis works, none of these solutions considers the reconfiguration property. In contrast, this paper focuses on how the feasible code may be generated for multi-core reconfigurable real-time system.

Another related area is the safe deployment to multi-core real-time systems, where approaches focus on the mapping of tasks on multi-core platforms (Monot et al., 2012), (Saidi et al., 2015), (Yi et al., 2009), (Vulgarakis et al., 2014) (Faragardi et al., 2013). In (Monot et al., 2012), the authors develop a heuristic algorithm for the function mapping on a multi-core architecture. In this work, the functions are grouped and distributed across cores, then they are mapped to tasks. Similarly to (Saidi et al., 2015), which proposes a heuristic algorithm to create a task set according to the mapping of runnable entities on the cores. In (Yi et al., 2009) a linear program is developed for task partitioning, mapping, and scheduling on embedded multi-core systems. Some other related works propose an end-to-end approach to generate a full real-time system. In (Vulgarakis et al., 2014), the authors present a process for the automatic deployment of control applications on multi-core platforms and generate Java code. The work reported in (Faragardi et al., 2013) addresses the mapping problem of hard real-time systems composed of periodic AUTOSAR runnables in the context of the multi-core.

The proposed approach differs from the cited works in several points. First of all, it considers reconfigurable systems with real-time properties and multi-core architecture. Secondly, we address the synthesis, the partitioning, the scheduling problem, and the optimization simultaneously. Finally, the majority of them do not propose general solutions. They target a well defined system.

## 3 SYSTEM FORMALIZATION

In this section, we present a formal description of a multi-core reconfigurable processor system. We present in addition real-time prerequisites required to introduce the paper's contribution.

### 3.1 System and Architecture Modeling

It is assumed in this work that a reconfigurable real-time system  $Sys$  is defined as a set of  $m$  implementations:  $Sys = \{imp_1, imp_2 \dots imp_m\}$ . We denote by  $Sys(t)$  the implementation defining the system at particular time  $t$  (i.e.,  $Sys(t) = imp_i$ ). An

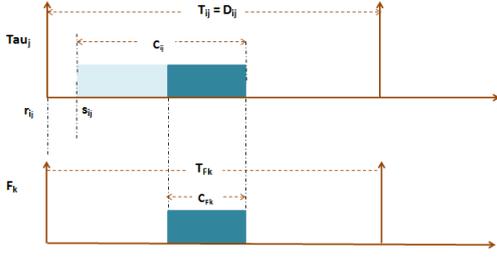


Figure 1: Task and function models.

implementation  $imp_i$  is composed of  $N_i$  tasks (i.e.,  $imp_i = \{\tau_1, \tau_2, \tau_3 \dots \tau_{N_i}\}$ ). Each task  $\tau_j$  in the implementation  $imp_i$  is characterized by (top of Fig 1):  $(r_{ij}, s_{ij}, T_{ij}, C_{ij}, D_{ij}, P_{ij}, E_{ij})$  where  $r_{ij}$  is its release time, we assume that  $r_{ij} = 0$ , its start time  $s_{ij}$  which denotes the effective starting time of a task  $\tau_j$ , its activation period  $T_{ij}$ ,  $C_{ij}$  denotes the capacity or worst case execution time, its deadline  $D_{ij}$  which is assumed to be equal to its period in this work  $D_{ij} = T_{ij}$ , the priority  $P_{ij}$  that is inversely proportional to the period  $T_{ij}$  as we use the RM policy,  $E_{ij}$  presents the energy consumption of task  $\tau_j$  which is computed as the sum of the energy consumed by the functions implemented by  $\tau_j$ .

The task  $\tau_j$  may implement a single or several functions which must have the same period  $\tau_j = \{F_1, F_2, F_3, \dots F_{p_j}\}$ . Each function  $F_k$  is characterized by static real-time parameters (bottom of Figure 1)  $(T_{F_k}, C_{F_k}, E_{F_k})$  where  $T_{F_k}$  is the activation period of the function  $F_k$ ,  $C_{F_k}$  is an estimation of its worst case execution time (WCET) and  $E_{F_k}$  is the energy consumed by the function  $F_k$  during its execution. Note that these parameters are considered as inputs for the proposed approach and must be specified by the user.

The system consists of one processor, containing a set of  $M$  identical cores  $\{\zeta_1, \zeta_2 \dots \zeta_M\}$  that share common memory. Each core runs a set of tasks. We assume that the tasks are independent and periodic.

### 3.2 Energy Model

Each function  $F_k$  is described by two parameters: (i) the function's frequency  $f_{F_k}$ , and (ii) the function's voltage  $V_{F_k}$ . The energy consumption for the execution of function  $F_k$  that we denote by  $E_{F_k}$  is computed as  $E_{F_k} = f_{F_k} V_{F_k}^2 C_{F_k}$ . The energy consumption  $E_{ij}$  of task  $\tau_j$  is then equal to the sum of the energy consumed by the implemented functions  $E_{ij} = \sum_{k \in \{1 \dots p_j\}} E_{F_k} = \sum_{k \in \{1 \dots p_j\}} f_{F_k} V_{F_k}^2 C_{F_k} = f_{ij} V_{ij}^2 C_{ij}$  where  $(f_{ij}, V_{ij})$  are two parameters characterizing task  $\tau_j$  in implementation  $imp_i$ . We assume that  $f_{ij} = \sum_{k \in \{1 \dots p_j\}} f_{F_k}$  and  $V_{ij} = \sum_{k \in \{1 \dots p_j\}} V_{F_k}$ . Thus the total energy consumption (Lei et al., 2016) of im-

plementation  $imp_i$  is given by expression 1

$$E_i = \sum_{j \in \{0, N_i\}} E_{ij} = \sum_{j \in \{0, N_i\}} f_{ij} V_{ij}^2 C_{ij} \quad (1)$$

In expression 2, we denote by  $f_n$  and  $V_n$  the normalized frequency and voltage of the system. We denote by  $\eta_j$  the reduction factor of voltage when  $\tau_j$  is executed,  $V_{ij} = \frac{V_n}{\eta_j}$  and  $f_{ij} = \frac{f_n}{\eta_j}$ . In addition, we denote by  $C_n$  the computation time at the normalized processor frequency i.e.,  $C_{ij} = C_n \eta_j$ . Thus, the total energy consumption of the implementation  $imp_i$  according to (Chniter et al., 2014) is given by

$$E_i = \sum_{j \in \{0, N_i\}} \frac{f_n * V_n^2 * C_n}{\eta_j^2} = K \sum_{j \in \{0, N_i\}} \frac{C_n}{\eta_j^2} \quad (2)$$

where  $K = V_n^2 f_n$ .

### 3.3 Processor Utilization Factor

Let  $U_i$  be the processor utilization factor of the implementation  $imp_i$  is defined by:  $U_i = \sum_{j=1}^{N_i} \frac{C_{ij}}{T_{ij}}$  (Klein et al., 1993). As we perform Rate-Monotonic (RM) assignment and preemptive scheduling, the real-time system is feasible when the test given by expression 3 is verified.

$$\forall i \in \{1 \dots m\}, U_i \leq N_i (2^{\frac{1}{N_i}} - 1). \quad (3)$$

### 3.4 Reconfiguration Time

We define in addition the reconfiguration time  $T_{reconf}$  (Lakhdhar et al., 2016) as the sum of the time required to add/remove tasks (i.e., time spent by the system to jump from one implementation to another) and the time required for task's migration. The time for task's migration refers to the period of time required for a task to move from one core to another when the system load a new implementation. Thus, we define the reconfiguration time as follow:

$$T_{reconf} = (A + B) * T_{cost} + C * T_{migration} \quad (4)$$

Where  $A$  is the number of the deleted tasks,  $B$  is the number of created tasks,  $T_{cost}$  is the spent time to delete/add a task,  $C$  is the number of migrated tasks and  $T_{migration}$  is the time spent to migrate from a core to another. One objective of the present work is to reduce the reconfiguration time  $T_{reconf}$  of the multi-core reconfigurable real-time system with the aim to improve its reactivity.

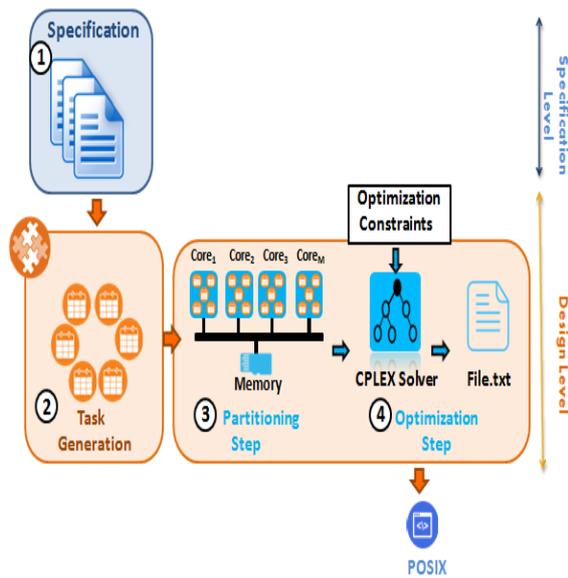


Figure 2: Process overview.

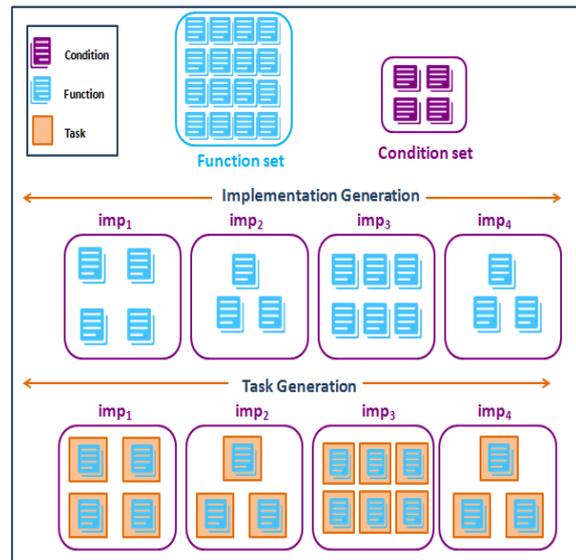


Figure 3: Example Of Initial Task Model.

## 4 PROPOSED APPROACH

In this section, we present the full working process used in this paper. As shown in Figure 2, the proposed process is composed of four main phases: (i) Specification, (ii) Task generation, (iii) Partitioning, and (iv) Optimization.

### 4.1 System Specification

In the specification level, the designer provides the specification model which defines (i) the reconfiguration conditions, (ii) the functions that must be executed under a considered condition. Each function is defined by a set of temporal parameters, and (iii) the set of cores. This model presents the input of the generation task step.

### 4.2 Task Generation

In this step, we aim to generate the initial task model from the specification model. As shown in Figure 3, this step is carried out in two sub-steps: (i) the generation of implementation sets from the condition sets and the assignment of functions to the appropriate implementation, and (ii) the generation of the task model from the function sets, such that each function is assigned to a task which will take the same parameters of the corresponding function (i.e., in this step, the task number is equal to the function number). Let us note that for the generation of this model, the real-time feasibility is not considered.

### 4.3 Task Partitioning

The partitioning of tasks into multiple cores must consider real-time feasibility. Under the hypotheses considered in this paper, the partitioning corresponds to an RM (Wang et al., 2014) scheduling problem related to periodic tasks. Each generated task must be affected to a specific core, then in each core, we run a local scheduler. This partitioning is characterized by (i) no migration at run time (i.e., in a given implementation a task must always run on a given core), (ii) the possibility of applying end-to-end worst case response time analysis, and (iii) off-line core assignment (Tindell and Clark, 1994). The basic principles of an RM-based partitioning heuristic are:

1. Order tasks according to RM policy,
2. Do task assignment according to their order,
3. For each task, look for an available core, by applying one of the following policies (Singhoff, 2014),
  - Best Fit policy: for each task  $i$ , we start with core  $j = 0$  and assign task  $i$  on the core on which the feasibility test is true and on which the processor utilization factor has the highest value
  - First Fit policy: for each task  $i$ , we start with core  $j = 0$  and assign task  $i$  on the first core on which the feasibility test (expression 3) is true. In this paper, we use the First Fit policy.
4. Stop when all tasks are assigned.

Algorithm 1 that illustrates this step considers as inputs: the implementations, tasks and the cores set. It generates as output the partitioning task model.

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**Algorithm 1:** Partitioning Task Model.

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**Input:**  
- Imp: Implementation set  
- Task: Task set  
- Core: Core set

**Output:**  
- PartTask: Partitioning Task Model

**1 Notations:**  
2 -  $m$ : implementation number  
3 -  $N_i$ : Number of tasks in the implementation  $i$ .  
4 -  $M$ : Core number

```

5 for  $i \leftarrow 0$  to  $m$  do
6     /*** Task Index ***/
7      $j \leftarrow 0$ 
8     /*** Core Index ***/
9      $k \leftarrow 0$ 
10    for  $j \leftarrow 0$  to  $N_i$  do
11        for  $k \leftarrow 0$  to  $M$  do
12            if Faisabilty is true then
13                AssignTask[j]toCore[k]
14                PartTask[j][k] = Core[k]
15            else
16                 $k++$ 
17 return PartTask
```

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## 4.4 Optimization Step

In order to ensure a reliable implementation of the multi-core reconfigurable real-time system from the initial task model and the partitioning task model taking into consideration the different constraints (Figure 2) (Real-time, no migration, energy), we propose in this section a MILP model which consists of a linear objective function to be optimized and a set of linear inequalities (constraints).

### 4.4.1 Variable Definition

Let (i)  $Merge_{jq}$  be a boolean variable used to mention whether two tasks  $\tau_j$  and  $\tau_q$  are merged such that  $Merge_{jq}$  is equal to 1 if task  $\tau_j$  and task  $\tau_q$  are merged, the merge corresponds to the situation in which  $\tau_j$  absorbs  $\tau_q$ , to be deleted from the model, (ii)  $x_{js}$  be a boolean variable used to mention if  $\tau_j$  is executed in core  $s$ . Thus if the value of  $x_{js}$  is equal to 1, then the corresponding task  $\tau_j$  is running in core  $s$ , (iii)  $y_{ij}$  be a boolean variable used to mention if  $\tau_j$  is in the implementation  $i$ , (iv)  $T$  be the set of period of  $N$  task, (v)  $C_{newij}$  be the new WCET of the task  $\tau_j$  in  $imp_i$ , (vi)  $T_{newij}$  be the new period of the task  $\tau_j$  in  $imp_i$ , (vii)  $\mu_{jq}$  be a binary variable where  $\mu_{jq} = 1$  when  $\tau_q$  is executed before  $\tau_j$ .

### 4.4.2 Objective Function

$$\text{maximize } \sum_{j,q \in \{0..N\}} Merge_{jq} - \sum_{i \in \{1..m\}} \sum_{j \in \{0..N\}} E_{ij} \quad (5)$$

The expression 5 defines the objective function. It aims to maximize the number of merges while minimizing the total energy consumption.

### 4.4.3 Merging Situation Constraints

The constraints 6 and 7 introduce the merging condition such as the two tasks  $\tau_j \in \zeta_s$  and  $\tau_q \in \zeta_s$  will be merged if they have the same period.

$$\forall j, q \in \{1..N\} \quad s \in \{1..M\}$$

$$\text{if}(T_j * x_{js} - T_q * x_{qs} = 0) \text{ then } Merge_{jq} = 1; \quad (6)$$

$$\text{if}(T_j * x_{js} - T_q * x_{qs} <> 0) \text{ then } Merge_{jq} = 0; \quad (7)$$

The constraint in 8 is used to avoid a non-meaningful situations which corresponds to the merge of a task already merged i.e.,  $\forall j, q, r \in \{1..N\}$ ,

$$Merge_{jq} \leq 1, \quad , q, r \neq j, Merge_{jq} + Merge_{rj} \leq 1 \quad (8)$$

### 4.4.4 Real-time Constraints

In each implementation, every pair of tasks  $\forall j, q \in \{1..N\}$   $\tau_j$  and  $\tau_q$ , we should respect the constraints 9 and 10 to ensure that only one task will be executed at a single time.  $\forall i \in \{1..m\}$

$$s_{ij} - s_{iq} \geq C_{newiq} - M * \mu_{jq} \quad (9)$$

$$s_{iq} - s_{ij} \geq C_{newij} - M * (1 - \mu_{jq}) \quad (10)$$

where  $C_{newij}$  and  $C_{newiq}$  are the WCET of the tasks  $\tau_j$  and  $\tau_q$ . Constraints 4.4.4, 12, and 13 give the computation formula of  $C_{newij}$ .

If a task  $\tau_j \in imp_i$  does not be merged with any task in all implementations, the WCET of  $\tau_j$  does not change.  $\forall j \in \{1..N\} \quad \forall i \in \{1..m\}$

$$\text{if}(\sum_{q \in \{1..N\}} Merge_{qj} + \sum_{r \in \{1..N\}} Merge_{jr} = 0) \text{ then } C_{newij} = C_{ij}; \quad (11)$$

Else if a task  $\tau_j$  is merged with another task in the same implementation or not, the WCET of  $\tau_j$  is calculated in two cases: (i) the task  $\tau_j$  and  $\tau_q$  are in the same implementation, so the resulting WCET is the sum of  $C_{ij}$  and  $C_{iq}$ , and (ii) the task  $\tau_j$  and  $\tau_q$  are not in the same implementation, so the resulting task has two different WCETs in the two implementations. Constraints 12, and 13 allow to compute the resulting WCET in the two cases.  $\forall j, q \in \{1..N\} \quad i, l \in \{1..m\}$

$$\text{if}(Merge_{jq} + y_{ij} + y_{lp} = 3) \text{ then } C_{newij} = C_{ij} + C_{lq} \quad (12)$$

$$\text{and } C_{newlq} = 0;$$

$$\begin{aligned} \text{if}(Merge_{jq} + y_{ij} + y_{lp} = 2) \text{ then } C_{new_{ij}} &= C_{ij} \\ \text{and } C_{new_{lq}} &= C_{lq}; \end{aligned} \quad (13)$$

Constraint 14 ensures the feasibility of the system,  $\forall i \in \{1..m\}$ :

$$U_i \leq N(2^{\frac{1}{N}} - 1) \quad (14)$$

where  $U_i$  is given by

$$U_i = \sum_{j=1}^N \frac{C_{new_{ij}}}{T_{new_{ij}}} \quad (15)$$

Where  $T_{new_{ij}}$  is computed as follow:  $\forall j, q \in \{1..N\} i, l \in \{1..m\}$

$$\begin{aligned} \text{if}(Merge_{jq} + y_{ij} + y_{lp} = 3) \text{ then } T_{new_{ij}} &= T_{ij} \\ \text{and } T_{new_{lq}} &= 0; \end{aligned} \quad (16)$$

$$\begin{aligned} \text{if}(Merge_{jq} + y_{ij} + y_{lp} = 2) \text{ then } T_{new_{ij}} &= T_{ij} \\ \text{and } T_{new_{lq}} &= T_{lq}; \end{aligned} \quad (17)$$

The start time should respect Constraint 18 i.e.,

$$\forall j \in \{1..N\}, \forall i \in \{1..m\}, s_{ij} \geq r_{ij} \quad (18)$$

#### 4.4.5 Energy Constraints

$$E_{ij} = K \sum_{j \in \{1..N\}} \frac{C_n}{\eta_j^2} \quad (19)$$

The energy consumption's equation is fractional, thus we simplify this program in order to be interpretable by using the CPLEX solver that maximizes the reduction factor  $\eta_j$  which is inversely proportional to the energy consumption as defined in Section 3. The objective function becomes

$$\text{Maximise } \sum_{j,q \in \{1..N\}} Merge_{jq} + \sum_{i \in \{1..m\}} \sum_{j,q \in \{0..N\}} x \quad (20)$$

To ensure the no simultaneous execution of tasks the constraints 9 and 10 become respectively 21 and 22:  $\forall j, q \in \{1..N\} \tau_j$  and  $\tau_q \forall i \in \{1..m\}$

$$s_{ij} - s_{iq} \geq C_n * \eta_q - M * \mu_{jq} \quad (21)$$

$$s_{iq} - s_{ij} \geq C_n * \eta_j - M * (1 - \mu_{jq}) \quad (22)$$

$\forall i \in \{1..m\}, j \in \{1..N\}$

To ensure that the start time is always greater than the release time, Constraint 23 is considered:

$$s_{ij} \geq r_{ij} \quad (23)$$

We limit the value of  $x$  by Constraint 24

$$x \leq \eta_j \quad (24)$$

The proposed approach allows to manually generate the code from the optimized task model. The optimized task model has properties providing information on (i) the core set, (ii) the implementation set, (iii) the task set, (iv) the function set, (v) the task assignment, and (vi) the scheduling order. For each task in the optimized task model, we implement a POSIX thread by using the POSIX *pthread*. In the POSIX code, the assignment of the tasks to the core is giving by *stick\_this\_thread\_system*, which allow moving from implementation to another, following well-defined conditions (i.e., user requirements).

## 5 CASE STUDY

In this section, we illustrate the proposed approach through a classical case study: a Global Positioning System (GPS) (Lakhdhar et al., 2016).

### 5.1 Specification Level

The GPS is used to define the position of an object on a plan or a map using the information provided via radio signals by the associated satellites. In the GPS, the satellite sends to the terminal an encrypted signal containing various information relevant to the location and timing. The terminal collects and converts radio signals received into information about the position, speed and time (Lakhdhar et al., 2016). In order to illustrate the proposed approach, we have enriched and extended this case study by introducing two modes: (i) default mode which consists of a default use of GPS, and (ii) secure mode which represents a restricted use of GPS with safety requirements.

The software architecture of the studied application is composed of seven functions in default mode and of eight functions in the secure mode such that every function is characterized by a period, a WCET and an energy consumption (Table 1). It is mapped to a preemptive execution platform composed of one processor which contains two cores  $C_1$  and  $C_2$ . A tabular description of the specification model is given in Table 1.

Table 1 depicts two execution modes *Default Mode* and *Secure Mode*. Each mode is characterized by a set of functions defined by a set of timing parameters.

### 5.2 Initial Task Model

The second step consists in generating the implementations and their tasks. Proceeding from the specification model, for each condition we generate an im-

Table 1: Specification Model.

Execution mode	Condition	Function Name	Period ms	WCET ms	Energy mW
Default	Sec=Disabled	$F_1$ : ControlBase	100	20	1687.5
		$F_2$ : GpsSatellite	200	20	30
		$F_3$ : Position	300	20	270
		$F_4$ : Receiver	300	50	2031.25
		$F_5$ : Decoder	400	40	5760
		$F_6$ : TreatmentUnit	400	50	15000
		$F_7$ : Encoder	500	30	15360
Secure	Sec=Enabled	$F_1$ : ControlBase	100	20	1687.5
		$F_2$ : GpsSatellite	200	20	30
		$F_3$ : Position Secure	300	20	270
		$F_4$ : Receiver	300	50	2031.25
		$F_5$ : Decoder	400	40	5760
		$F_6$ : TreatmentUnit	400	50	15000
		$F_7$ : Encoder	400	50	15360
		$F_8$ : AccessController	400	50	10800

plementation so we have two implementations. Then, we assign each function to a task. The resulting task model is given in Table 2 that shows two implementations which are composed of fifteen tasks. Each task is characterized by the same real-time parameters of the executed function.

Table 2: Initial Task Model.

Implementation	Task	Period ms	WCET ms	Energy mW
Default	$\tau_1$	100	20	1687.5
	$\tau_2$	200	20	30
	$\tau_3$	300	20	270
	$\tau_4$	300	50	2031.25
	$\tau_5$	400	40	5760
	$\tau_6$	400	50	15000
	$\tau_7$	500	30	15360
Secure	$\tau_8$	100	20	1687.5
	$\tau_9$	200	20	30
	$\tau_{10}$	300	20	270
	$\tau_{11}$	300	50	2031.25
	$\tau_{12}$	400	40	5760
	$\tau_{13}$	400	50	15000
	$\tau_{14}$	400	50	15360
	$\tau_{15}$	400	50	10800

### 5.3 Partitioning Task Model

The next step consists in distributing the task model into a specific multi-core architecture. The targeted multi-core architecture contains 2 cores and a shared memory. In order to assign the tasks to the cores, we apply Algorithm 1. Table 3 presents the partitioning task model.

Table 3: Partitioning Task Model.

Implementation	Core	Task	Period/Deadline ms	WCET ms
Default	$\zeta_1$	$\tau_1$	100	20
		$\tau_2$	200	20
		$\tau_3$	300	20
		$\tau_4$	300	50
	$\zeta_2$	$\tau_5$	400	40
		$\tau_6$	400	50
		$\tau_7$	500	30
Secure	$\zeta_1$	$\tau_8$	100	20
		$\tau_9$	200	20
		$\tau_{10}$	300	20
		$\tau_{11}$	300	50
	$\zeta_2$	$\tau_{12}$	400	40
		$\tau_{13}$	400	50
		$\tau_{14}$	500	30
		$\tau_{15}$	600	50

For each implementation, we assign the tasks to the appropriate core based on the feasibility tests. The resulting partitioning task model represents the input of the optimization step.

### 5.4 Optimized Task Model

Once the initial solution is defined, we process the optimization step. This step involves the execution using CPLEX solver of the proposed linear program. We model the input of this step as two matrices: “tasks to implementation mapping matrix” (i.e.,  $y$ ) and “tasks to core assignment matrix” (i.e.,  $x$ ). These two matrices are defined as follows:

$$y = \begin{pmatrix} \tau_1 & imp_1 & imp_2 \\ \tau_2 & 1 & 0 \\ \tau_3 & 1 & 0 \\ \tau_4 & 1 & 0 \\ \tau_5 & 1 & 0 \\ \tau_6 & 1 & 0 \\ \tau_7 & 1 & 0 \\ \tau_8 & 0 & 1 \\ \tau_9 & 0 & 1 \\ \tau_{10} & 0 & 1 \\ \tau_{11} & 0 & 1 \\ \tau_{12} & 0 & 1 \\ \tau_{13} & 0 & 1 \\ \tau_{14} & 0 & 1 \\ \tau_{15} & 0 & 1 \end{pmatrix} \begin{pmatrix} \zeta_1 & \zeta_2 \\ \tau_1 & 1 & 0 \\ \tau_2 & 1 & 0 \\ \tau_3 & 1 & 0 \\ \tau_4 & 1 & 0 \\ \tau_5 & 0 & 1 \\ \tau_6 & 0 & 1 \\ \tau_7 & 0 & 1 \\ \tau_8 & 1 & 0 \\ \tau_9 & 1 & 0 \\ \tau_{10} & 1 & 0 \\ \tau_{11} & 1 & 0 \\ \tau_{12} & 0 & 1 \\ \tau_{13} & 0 & 1 \\ \tau_{14} & 0 & 1 \\ \tau_{15} & 0 & 1 \end{pmatrix} = x$$

The linear program generates the following Merge matrix:

$$Merge = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

We note that the MILP program allows to merge the tasks: (i)  $\tau_1$  with  $\tau_8$ , (ii)  $\tau_2$  with  $\tau_9$ , (iii)  $\tau_3$  with  $\tau_4$ ,  $\tau_{10}$  and  $\tau_{11}$ , (vi) the task  $\tau_5$  with  $\tau_6$ ,  $\tau_{12}$  and  $\tau_{13}$ , and (v)  $\tau_7$  with  $\tau_{14}$ . The input matrices become:

$$y = \begin{pmatrix} \tau_1 & imp_1 & imp_2 \\ \tau_1 & 1 & 1 \\ \tau_2 & 1 & 1 \\ \tau_3 & 1 & 1 \\ \tau_5 & 1 & 1 \\ \tau_7 & 1 & 1 \\ \tau_{15} & 0 & 1 \end{pmatrix} \begin{pmatrix} \zeta_1 & \zeta_2 \\ \tau_1 & 1 & 0 \\ \tau_2 & 1 & 0 \\ \tau_3 & 1 & 0 \\ \tau_5 & 0 & 1 \\ \tau_7 & 0 & 1 \\ \tau_{15} & 0 & 1 \end{pmatrix} = x$$

The optimized task model is presented in Table 4.

Table 4: Optimized Task Model.

Implementation	Core	Task	Period ms	WCET ms	$E_{old}$ mW	$E_{new}$ mW
Default	$\zeta_1$	$\tau_1$	100	20	1687.5	1181.25
		$\tau_2$	200	20	30	21
		$\tau_3$	300	20	2301.25	1610.87
	$\zeta_2$	$\tau_5$	300	50	20760	14532
		$\tau_7$	400	40	15360	10752
		$\tau_{15}$	400	50	10800	7560
Secure	$\zeta_1$	$\tau_1$	100	20	1687.5	1181.25
		$\tau_2$	200	20	30	21
		$\tau_3$	300	20	2301	1610.87
	$\zeta_2$	$\tau_5$	300	50	20760	14532
		$\tau_7$	400	40	15360	10752
		$\tau_{15}$	400	50	10800	7560

We note that the task number is reduced as well as the energy consumption. Finally, we generate a POSIX code from the optimized task model describing the GPS. Listing 1 gives an excerpt of the GPS code. The role of the controller which corresponds to the main function in Listing 1, is to switch from one implementation to another under a considered condition.

```

1 #include <pthread.h>
2 void* F1 (void* arg);
3 void* F2 (void* arg);
4 ...
5 /****** Controller POSIX code *****/
6 int main (void){
7 pthread_t tau_prime_1;
8 pthread_t tau_prime_2;
9 pthread_t tau_prime_3;
10 // Default mode
11 pthread_create (&tau_prime_1, NULL, F1, (void *) 100);
12 // Craation of tau_prime_1 thread
13 pthread_create (&tau_prime_2, NULL, F2, (void *) 200);
14 // Craation of tau_prime_2 thread
15 pthread_create (&tau_prime_3, NULL, F3, (void *) 300);
16 // Craation of tau_prime_3 thread
17 ...
18 stick_this_thread_to_core(1); //assign thread 1 to core 1 //
19 //
20 pthread_join (tau_prime_1, NULL);
21 stick_this_thread_to_core(1);
22 pthread_join (tau_prime_2, NULL);
23 // Secure Mode
24 if (cnd=="Secure"){
25 pthread_create (&tau_prime_1, NULL, F8, (void *) 100);
26 // Craation of tau_prime_1 thread
27 pthread_create (&tau_prime_2, NULL, F9, (void *) 200);
28 // Craation of tau_prime_2 thread
29 ...
30 return 0;}
31 ...
32 void* F1 (void* arg)
33 {/**/}
34 // core_id = 0, 1, ... n-1, where n is the system's
35 // number of cores
36 int stick_this_thread_to_core(int core_id) {
37 int num_cores = sysconf(_SC_NPROCESSORS_ONLN);
38 if (core_id < 0 || core_id >= num_cores)
39 return EINVAL;
40 cpu_set_t cpuset;
41 CPU_ZERO(&cpuset);
42 CPU_SET(core_id, &cpuset);
43 pthread_t current_thread = pthread_self();
44 return pthread_setaffinity_np(current_thread, sizeof(
45 cpu_set_t), &cpuset);}
    
```

Listing 1: GPS POSIX code.

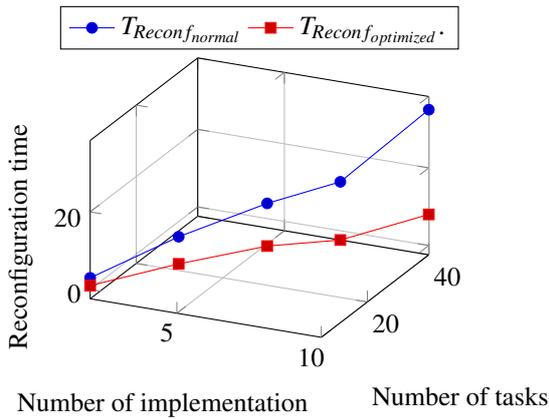


Figure 4: Evaluation of the Reconfiguration Time.

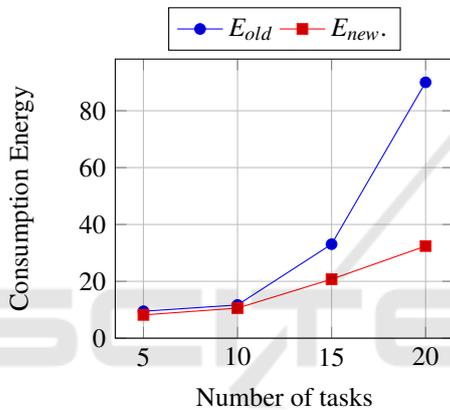


Figure 5: Evaluation of the Energy Consumption.

### 5.5 Evaluation

In order to generalize the performance evaluation of our strategy, we generate a random system with random task set. The experiments are carried-out on Intel Core i5-4200U processor running at 2.8 GHz with 4GB of cache memory. The curve in Figure 4 shows the variation of the reconfiguration time of the system described in section 3 depending on the number of tasks and the number of implementations.

In Figure 4 we compare the reconfiguration time using the proposed approach with the normal reconfiguration time. It can be seen that the proposed approach allows obtaining a lower reconfiguration time. This is due to the task merging technique.

Figure 5 depicts the impact of our approach on the total energy consumed by the system.

In this figure, we compute the energy consumption and we compare it with the energy consumed by the system before applying the proposed approach. It is clear from this figure that we have obtained better results.

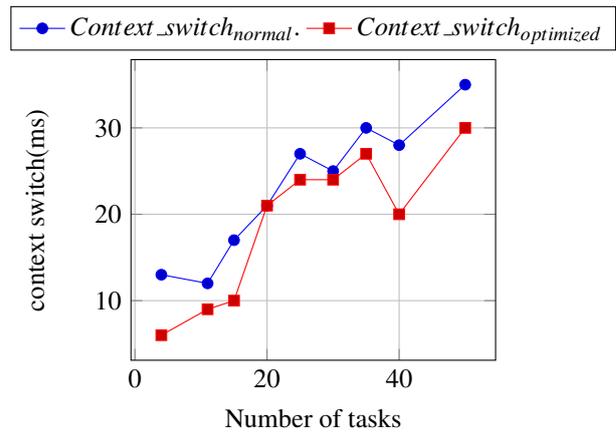


Figure 6: Evaluation of the Context switching.

We also compare in Figure 6 the context switching of the proposed approach to the context switching before applying this approach in a randomly generated system with a number of tasks that varies between 6 and 50. We note that this comparison shows the efficiency of the proposed approach.

The originality of this paper is manifested in the fact that the proposed approach allows the implementation of the multi-core reconfigurable real-time systems by reducing: (i) the task number, (ii) the energy consumption, (iii) the reconfiguration time, (iv) the time overhead in terms of context switching, and (v) the redundancies between implementations.

## 6 CONCLUSIONS

In this paper, we have described a process for the semi-automatic synthesis of an energy-aware POSIX code for multi-core reconfigurable real-time systems. Firstly, we showed how from the input specification model we generate an initial task model, assign tasks to the cores using partitioning scheduling while meeting timing properties. Secondly, in order to reduce the time overhead, energy consumption and the redundancies between the implementations caused by the huge number of tasks, we proposed a MILP formulation for the problem, which can find the optimal solution for the proposed system model. Thirdly, we generate the optimized POSIX code from the optimized task model. We evaluated the performance of the proposed approach by comparing the obtained results against the performance of the system before applying the approach. As a future work, we want (i) to introduce other criteria to be optimized, and (ii) to fully automate the proposed approach by introducing transformation techniques.

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