Minimize the cardinality of a real-time task set through Task Clustering

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Context

- Focus on hard real-time systems
- Interest in programming of large systems specified as high-level functionalities



• Up to ≈ 1000 high-level functionalities in RT system software (e.g. aileron command, read pressure sensor, etc.)

Task Clustering

Problem

- Functionalities implemented via real-time threads (tasks) by programmers
- RT operating systems (OS) support a **limited number** of concurrent threads (several tens of OS tasks)

Task Clustering

RTOS limitations

Having numerous threads:

- Scheduling overhead
 - Scheduler level: handle large queues
 - Increase of context switches
 - \Rightarrow increase the risk of cache misses (larger WCET)
- Memory
 - Task level: one stack by task
 - $\circ~$ Scheduler level: increase in the number of priorities required
 - \Rightarrow number of preemption increases
 - \Rightarrow execution stack grows

Task Clustering

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\rightarrow Several functionalities grouped together in a thread Manually made in industry (error prone, tedious)

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System model used

Program = a set of tasks τ_i :

• T_i: period



- D_i : constrained relative deadline $(D_i \leq T_i)$
- *C_i*: worst case execution time (WCET)

Current limitations: independent and synchronous tasks (offset = 0) in a uniprocessor system.

Fixed task-priority and fixed-job-priority assignment considered.

Objective

- Automatically grouping functionalities into tasks to minimize their number:
- while respecting original timing constraints,
- while preserving schedulability.



Cluster model

• **Cluster** τ_i and τ_j into τ_{ij} with $D_i \leq D_j$



- $\circ \ C_{ij} = C_i + C_j$
- $T_{ij} = T_i = T_j$ (by restriction we only regroup tasks with equal periods)
- Which deadline for the cluster?





(b) System after clustering τ_i with τ_j Schedulability preserved \Rightarrow **Zero-cost clustering**

Case 1: Maximum deadline D_j
o if (D_i − C_i ≤ D_i) or generalizing (R_i − C_i ≤ D_i)

• and $\tau_{ij} = \tau_j \leq D_i$ or generalizing (T_j) • and $\tau_{ij} = \tau_j = \tau_j$ in that order



(a) Initial system with tasks τ_i, τ_x et τ_j



(b) System after clustering τ_i with τ_j Schedulability preserved \Rightarrow **Zero-cost clustering**

- Case 2: Minimum deadline D_i
 - Taking minimum deadline ensures respect of both initial ones





(a) Initial system with tasks τ_i, τ_x et τ_j



(b) System after clustering τ_i with τ_j
System may become unschedulable after clustering

- Case 2: Minimum deadline D_i
 - Taking minimum deadline ensures respect of both initial ones
 - $\circ \quad \tau_{ij} \quad \tau_i | \tau_j \quad \text{or} \quad \tau_{ij} \quad \tau_j | \tau_i \quad \text{(order does not matter)}$



(a) Initial system with tasks τ_i, τ_x et τ_j



(b) System after clustering τ_i with τ_i

- System may become unschedulable after clustering
- \Rightarrow Schedulability must be checked after each non zero-cost clustering!

Cluster model

Valid cluster

Theorem

Let S be a task set and Φ be a priority assignment. Let S' the task set S after clustering of two tasks τ_i and τ_j .

 \mathcal{S}' is schedulable under $\Phi \Rightarrow \mathcal{S}$ is schedulable under $\Phi.$

However, the converse is not always true.

Definition

The clustering is valid iff schedulability is preserved after clustering

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Complexity: a schedulability problem

Reminder:

• sufficient test

- often in linear complexity
- ensures system schedulability
- sub-optimal
- necessary test
 - \circ does not ensure schedulability
- **exact** \Rightarrow sufficient AND necessary test
 - generally NP-hard

In our case, use of sufficient or exact tests to guarantee system correctness

Complexity: and a partitionning problem

Example

Some possible combinations (15 possibilities) of 4 tasks τ_a, τ_b, τ_c et τ_d :



- Combinatorial explosion: number of possible clusterings in the Bell number range (e.g., $B_{500}=10^{844})$

Overall Complexity

- NP-hard: reduced from bin-packing with fragile objects
- Given:
 - \circ a set of bins of **capacity** c_i
 - $\circ\,$ a set of object with a size s_i and a fragility f_i
- Assign objects to a minimum number of bins such that for each bin *j*:

$$\sum s_i \leq c_j$$

- $\sum s_i \leq \min(f_i)$
- Task clustering analogy:
 - $\circ \ \mathsf{bin} \Rightarrow \mathsf{cluster}$
 - $\circ \ \mathsf{object} \Rightarrow \mathsf{task}$
 - $\circ \ \ \mathsf{fragility} \Rightarrow \mathsf{deadline}$

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 \rightarrow Motivates to work towards a heuristic

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Heuristic principles

- Classical optimization techniques are based on cost functions to choose "promising candidates"
 - \Rightarrow Schedulability test used as a cost function. (Applied to greedy BFS, but works for simulated annealing, A*, etc.)
- Perform in priority zero-cost clustering
- Use sufficient test when exact impracticable

- Idea: Successive clusterings from an initial task set
- Heuristic cost function based on response time Rcomputation: $h(S) = \sum_{n=1}^{|S|} \frac{R_n}{D_n}$ $(\frac{R_n}{D_n} \text{ closer to 1 means less margin for the scheduler}).$



• Sustainable unschedulability: a task set deemed unschedulable remains so after clustering \Rightarrow avoid useless search ^{20 de 24}

(A)(B)(C)(D)

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schedulable best schedulable

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Results under DM

Results under EDF

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We presented in this talk:

- the task clustering problem,
- its complexity,
- some heuristic principles,
- and a first heuristic.

Current and future work:

- adding precedences between tasks
- then applying task clustering to multi-processor systems