#### Global and Partitioned Scheduling

CPEN 432 Real-Time System Design

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### Partitioned Scheduling

## Reasonable Algorithms

- Reasonable allocation (RA): An algorithm that fails to allocate a task to a multiprocessor platform
  only when the task does not fit into any processor on the platform
- When a task is considered for assignment, to which processor does it get assigned?
  - First-fit (FF): the processors are considered ordered in some manner and the task is assigned to the first processor on which it fits
  - Worst-fit (WF): the task is assigned to the processor with the maximum remaining capacity
  - Best-fit (BF): the task is assigned to the processor with the minimum remaining capacity exceeding its own utilization (i.e., on which it fits)
- In what order are the tasks considered for assignment?
  - Decreasing (D): tasks are considered in non-increasing order of their utilizations
  - Increasing (I): tasks are considered in non-decreasing order of their utilizations
  - Unordered (ε): tasks are considered in arbitrary order (i.e., tasks need not be sorted prior to allocation)
- Nine different heuristics: {FF, WF, BF} x {D, I, ε}, i.e., FFD, FFI, FF, WFD, WFI, WF, BFD, BFI, BF

#### Utilization Bounds

- Let  $\alpha$  denote an upper bound on the per-task utilization, and  $\beta = \lfloor 1/\alpha \rfloor$
- For any reasonable allocation algorithm, its utilization bound  $U_b$  is bounded as follows:  $m-(m-1)\alpha \leq U_b \leq (\beta m+1)/(\beta+1)$
- WF and WFI:  $U_b = m (m-1)\alpha$
- FF, FFI, FFD, BF, BFI, BFD, and WFD:  $U_b = (\beta m + 1)/(\beta + 1)$
- What if  $\alpha$  is unknown?

### Reasonable Algorithms

- Nine different heuristics: {FF, WF, BF} x {D, I, ε}
- Each can be implemented extremely efficiently
  - Sorting n tasks:  $O(n \log n)$
  - Choosing a fit for any given task: O(m)
- From Multiprocessor Scheduling for Real-Time Systems (Baruah et al., 2015)
  - \* "... it seems reasonable to actually run the partitioning algorithm, rather than computing the utilization of the task system and comparing against the algorithm's (sufficient, not exact) utilization bound ... from the perspective of actually implementing a real-time system using partitioned scheduling, there is no particular significance to using a utilization bound formula rather than actually trying out the algorithms. Rather, the major benefit to determining these bounds arises from the insight such bounds may provide regarding the efficacy of the algorithm." [Emphasis added]

### Speedup Factors

- Consider partitioning algorithms  $\mathcal{A}_{optimal}$  (optimal) and  $\mathcal{A}_{heuristic}$  (approximate)
- Speedup factor of  $\mathcal{A}_{heuristic}$ 
  - The smallest number f such that any task system that can be partitioned by  $\mathcal{A}_{optimal}$  upon a particular platform can be partitioned by  $\mathcal{A}_{heuristic}$  upon a platform in which each processor is f times faster

Nine different heuristics: {**FF, WF, BF**} x {**D, I, ɛ**} 
$$f_{FFD, WFD, BFD} = \frac{4}{3} - \frac{1}{3m}, f_{WF, WFI} = 2 - \frac{2}{m}, f_{FF, FFI, BF, BFI} = 2 - \frac{2}{m+1}$$

- More expressive than utilization bounds
  - As observed in practice, the speedup factors show that when tasks are considered in non-increasing order of their utilizations (i.e., FFD, WFD, BFD), partitioning is easier
- Question: What is the speedup factor of algorithm  $\mathscr{A}'_{optimal}$  that is also an optimal algorithm?

### A PTAS for Partitioning

- Optimal partitioning is NP-hard
- Common heuristics have a speedup factor of 4/3 or 2 as  $m \to \infty$
- Is there a Polynomial-Time Approximation Scheme (PTAS) that can achieve a speedup factor of  $1 + \epsilon$ , for any positive constant  $\epsilon$ ?
  - If yes, we can partition a task set to any desired degree of accuracy in polynomial time
  - Of course, we will need faster processors in order to run the tasks :-)
- Next few slides
  - PTAS for partitioning proposed by <u>Hochbaum and Shmoys (1987)</u>
  - Implementation by Chattopadhyay and Baruah (2011)

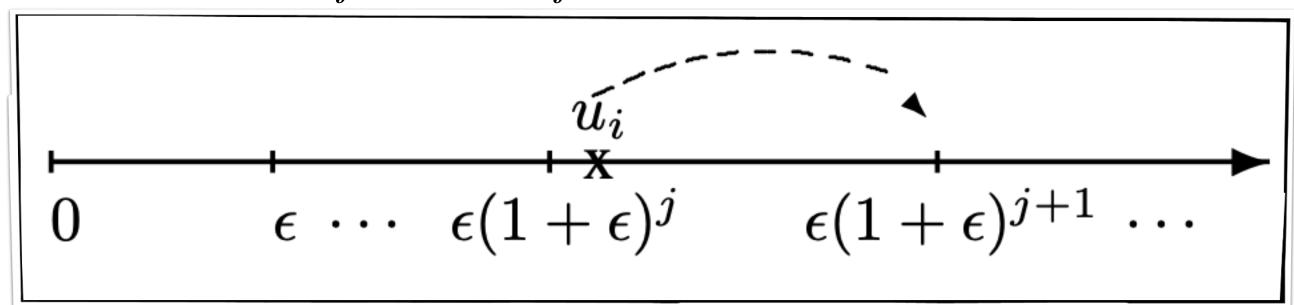
# Key Ideas [1/5]

- Choosing  $\epsilon$ 
  - $\blacktriangleright$  PTAS requires processors that are  $1+\epsilon$  times faster
    - We cannot provide faster processors
    - But we can assume that  $\mathscr{A}_{optimal}$  has only  $\left(\frac{1}{1+\epsilon}\right)^{th}$  of each processor available
    - Thus, if  $\mathscr{A}_{optimal}$  can partition a task set on m processors with an available utilization of  $(1/1+\epsilon)$  on each processor, then  $\mathscr{A}_{PTAS}$  can partition the task set on m processors with full utilization available
  - Suppose we are willing to tolerate a loss of up to 10% of the processor utilization

Thus, 
$$U_{loss}=1-\frac{1}{1+\epsilon}=\frac{\epsilon}{1+\epsilon}=0.1$$
 (i.e.  $10\%$ ), which yields  $\epsilon=\frac{1}{9}$ 

# Key Ideas [2/5]

- Bucketing utilization values
  - The utilization of a task  $u_i$  may vary anywhere from 0 to 1, i.e., infinitely many possibilities
  - Instead, we consider only a finite number of points  $V(\epsilon) = (v_0, v_1, v_2, v_3, ...)$  as valid utilizations
    - Where  $v_j = \epsilon \times (1 + \epsilon)^j \le 1$
  - Any utilization  $v_j < u_i < v_{j+1}$  is inflated to the next valid utilization  $v_{j+1}$



For example,  $V(\epsilon = 0.3) = \{0.3, 0.39, 0.507, 0.6591, 0.8568\}$ 

# Key Ideas [3/5]

- Enumerate all maximal single-processor configurations
  - Each configuration identifies a vector  $\langle x_1, x_2, ..., x_{|V(\epsilon)|} \rangle$ 
    - Where  $x_i$  identifies the number of tasks with utilization  $V(\epsilon)[i]$
    - Such that  $x_1v_1 + x_2v_2 + ...x_{|V(\epsilon)|}v_{|V(\epsilon)|} \le 1$
    - I.e., each configuration identifies a set of tasks that can be scheduled on a uniprocessor
  - The configuration is maximal if no other task can be further added
    - I.e., no  $x_i$  can be incremented without violating the above inequality

Config. ID	0.3000	0.3900	0.5070	0.6591	0.8568
1	3	0	0	0	0
2	2	1	0	0	0
3	1	0	1	0	0
4	1	0	0	1	0
5	0	2	0	0	0
6	0	1	1	0	0
7	0	0	0	0	1

Just seven maximal single-processor configurations for  $\epsilon=0.3$ 

## Key Ideas [4/5]

• Enumerate all maximal multi-processor configurations

0.3	0.39	0.507	0.6591	0.8568	Single-proc. ID's
3	2	1	2	0	[4 4 5 3]
3	4	2	0	0	[6 6 5 1]
0	3	3	0	1	[6 6 6 7]
4	1	1	1	1	[7 4 3 2]
4	0	1	3	0	[4 4 4 3]
					•

Example configurations for m=4 and  $\epsilon=0.3$  (out of 140)

Config. ID	0.3000	0.3900	0.5070	0.6591	0.8568
1	3	0	0	0	0
2	2	1	0	0	0
3	1	0	1	0	0
4	1	0	0	1	0
5	0	2	0	0	0
6	0	1	1	0	0
7	0	0	0	0	1

Just seven maximal single-processor configurations for  $\epsilon=0.3$ 

## Key Ideas [5/5]

$$\left[\frac{1}{5}, \frac{1}{5}, \frac{1}{3}, \frac{7}{20}, \frac{9}{25}, \frac{2}{5}, \frac{1}{2}, \frac{1}{2}, \frac{3}{4}\right]$$

Example task set

- Task assignment
  - Step 1: Round up task utilizations to values in  $V(\epsilon = 0.3) = \{0.3, 0.39, 0.507, 0.6591, 0.8568\}$
  - Step 2: Ignore "small" tasks with utilization less than  $\epsilon/1+\epsilon$
  - Step 3: For the remaining "large" tasks, identify a matching configuration from the multi-processor lookup table
  - Step 4: Assign these "large" tasks to appropriate processors based on the chosen configuration
  - Step 5: Assign each "small" task to any processor upon which it fits

0.3	0.39	0.507	0.6591	0.8568	Single-proc. ID's
3	2	1	2	0	[4 4 5 3]
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0	3	3	0	1	[6 6 6 7]
4	1	1	1	1	[7 4 3 2]
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<u>4</u>	<u> </u>	1	<u> </u>	U	[4 4 4 3]

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1	0	1	0	0
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0	2	0	0	0
0	1	1	0	0
0	0	0	0	1
	3 2 1 1 0 0	3 0 2 1 1 0 1 0 0 2 0 1 0 0	3     0       2     1       1     0       1     0       0     0       0     1       1     0       0     1       1     1       0     0       0     0       0     0	0.3000     0.3900     0.5070     0.6591       3     0     0     0       2     1     0     0       1     0     1     0       0     2     0     0       0     1     1     0       0     1     1     0       0     0     0     0

Just seven maximal single-processor configurations for  $\epsilon=0.3$ 

## Global Scheduling

#### Global vs. Partitioned: Pros & Cons

- ☐ Global scheduling
  - Automatic load balancing
  - ✓ Lower avg. response time
  - ✓ Simpler implementation
  - ✓ Optimal schedulers exist
  - ✓ More efficient reclaiming
  - Migration costs
  - Inter-core synchronization
  - X Loss of cache affinity
  - Weak scheduling framework

- Partitioned scheduling
  - Supported by automotive industry (e.g., AUTOSAR)
  - No migrations
  - ✓ Isolation between cores
  - Mature scheduling framework
  - Cannot exploit unused capacity
  - Rescheduling not convenient
  - X NP-hard allocation

#### The Dhall Effect

- Consider an implicit-deadline sporadic task system of (m + 1) tasks to be scheduled upon an m-processor platform
  - Tasks  $T_1, ..., T_m$  have parameters  $(e_i = 1, P_i = P)$
  - Task  $T_{m+1}$  has parameters  $e_{m+1} = P_{m+1} = P+1$

• 
$$U = m\left(\frac{1}{P}\right) + \frac{P+1}{P+1} = \frac{m}{P} + 1$$
 
$$\left[\lim_{P \uparrow \infty} U = 1\right]$$

- Is this task set schedulable with Global EDF if all tasks are released simultaneously?
- What happens if we increase the number of processors?

This task system is not EDF-schedulable despite having a utilization close to 1

The utilization bound of global EDF is **very poor**: it is arbitrarily close to one regardless of the number of processors.

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- Question: Is this task set partitioned-schedulable?
  - Answer: Yes! for example, when m < P, we need only two processors!

#### The Dhall Effect

- Dhall's Effect shows the limitation of global EDF and RM: both utilization bounds tend to 1, independently of the value of m.
- Researchers lost interest in global scheduling for ~25 years, since late 1990s.
- Such a limitation is related to EDF and RM, not to global scheduling in general

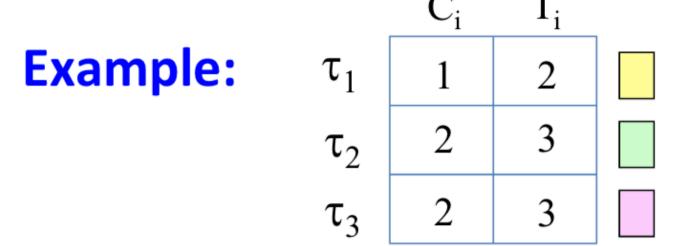
#### Global Scheduling: Negative Results

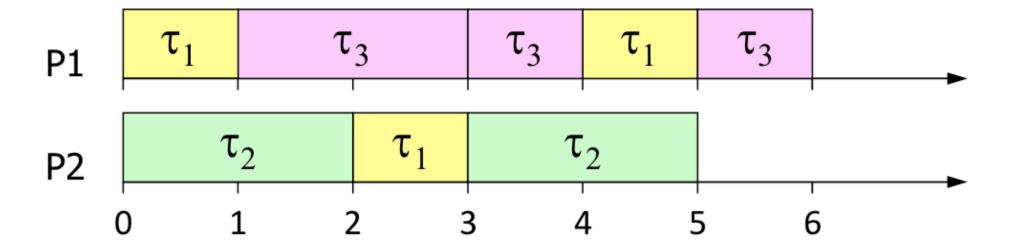
#### Weak theoretical framework

- Unknown critical instant
- Global EDF is not optimal
- Any global job-fixed (or task-dynamic) priority scheduler is not optimal
- Optimality only for implicit deadlines
- Many sufficient tests (most of them incomparable)

#### Global vs. Partitioned

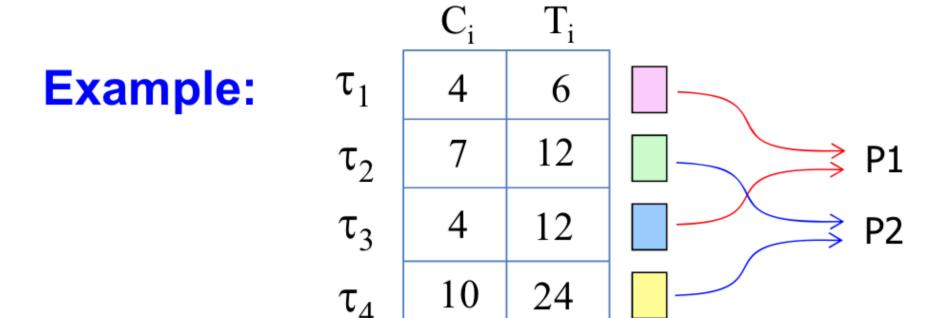
• There are tasks that are schedulable only with a global scheduler!

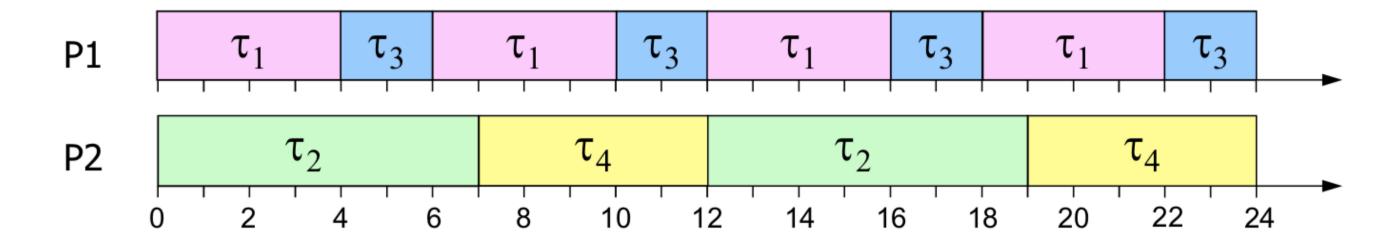




#### Global vs. Partitioned

• But there are also task sets that are schedulable only with a partitioned scheduler





All 4! = 24 global priority assignments lead to deadline miss.

#### Global vs. Partitioned

• Example of an unfeasible global schedule with  $\pi_1 > \pi_2 > \pi_3 > \pi_4$ 

