



# Optimal Online Load Maximization with Commitment Samin Jamalabadi, Uwe Schwiegelshohn TU Dortmund University Chris Schwiegelshohn Sapienza University of Rome



# $Pm|online, \varepsilon, commit|\sum p_j \cdot (1-U_j)$

- *Pm*: *m* parallel identical machines.
- online: jobs are submitted over time.
  - We do not know the existence nor any properties of future jobs.
- $\varepsilon$ : a job  $J_j$  has deadline  $d_j \ge r_j + \varepsilon \cdot p_j$  with constant slack parameter  $\varepsilon$ .
  - r<sub>j</sub>: submission time of job J<sub>j</sub>
  - *p<sub>j</sub>*: processing time of job *J<sub>j</sub>*
- *commit*: we must decide immediately after submission whether to reject a new job  $J_j$  ( $U_j = 1$ ) or to accept it ( $U_j = 0$ ).
  - For  $U_j = 0$ , we must also immediately fix the start time of the job.
  - We must complete every accepted job on time.
- $\sum p_j \cdot (1 U_j)$ : we want to maximize the total processing time of all accepted jobs.



### Algorithm Choices

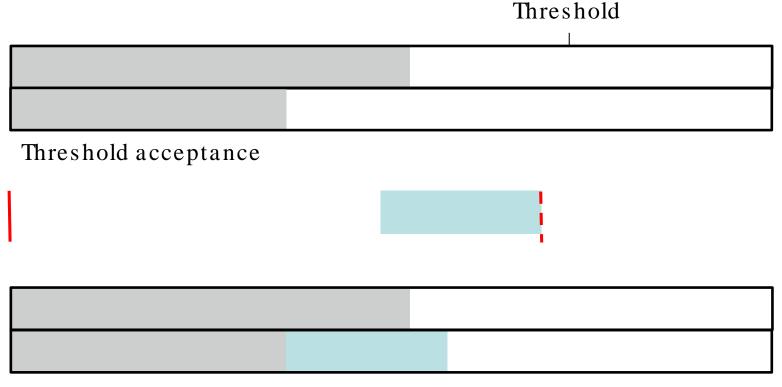
Acceptance

Allocation

Timing



#### Acceptance Algorithms for P2



Greedy acceptance

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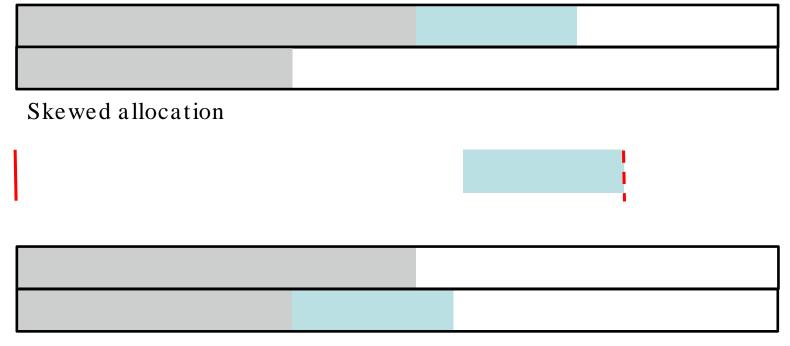
## Algorithm Choices

- Acceptance
  - Greedy: the resulting schedule completes all accepted jobs on time.
  - Threshold: the deadline of an accepted job is at least as large as a deadline threshold.
- Allocation

Timing



#### Allocation for P2 with Greedy Acceptance



**Balanced allocation** 

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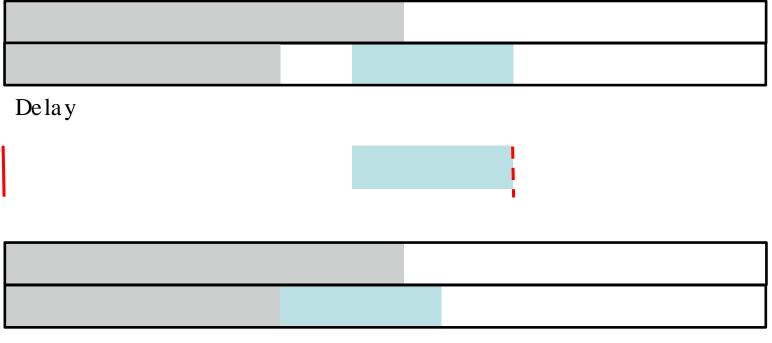


# Algorithm Choices

- Acceptance
  - Greedy: the resulting schedule completes all accepted jobs on time.
  - Threshold: the deadline of an accepted job is at least as large as a deadline threshold.
- Allocation
  - Skewed: the candidate machine with the highest load
  - Balanced: the candidate machine with the minimal load (for greedy) or the candidate machine that increases the threshold by the smallest amount (for threshold).
- Timing



#### Job Starting for P2 with Greedy Acceptance



Semi-active

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# Algorithm Choices

- Acceptance
  - Greedy: the resulting schedule completes all accepted jobs on time.
  - Threshold: the deadline of an accepted job is at least as large as a deadline threshold.
- Allocation
  - Skewed: the candidate machine with the highest load
  - Balanced: the candidate machine with the minimal load (for greedy) or the candidate machine that increases the threshold by the smallest amount (for threshold).
- Timing
  - Semi-active: as early as possible on the allocated machine.
  - Delay: possible intermediate idle time on the allocated machine.



## **Threshold Calculation**

- The machines are indexed in decreasing order of their outstanding loads.
- For machine  $m_i$ , we calculate a machine specific threshold using a function  $f_i(\varepsilon)$  and the outstanding load of the machine at time t.

$$d_{lim,i}\Big|_t = load(m_i)\Big|_t \cdot f_i(\varepsilon) + t$$

• The threshold is the maximum of the machine specific thresholds.

$$d_{lim}\Big|_t = \max_{1 \le i \le m} d_{lim,i}\Big|_t$$

• We set  $f_m(\varepsilon) = \frac{1+\varepsilon}{\varepsilon}$  and determine the remaining  $f_i(\varepsilon)$  recursively.  $\frac{m \cdot f_i(\varepsilon) + 1}{\sum_{h=1}^{i-1} f_i(\varepsilon) - (i-1) + 1} = const \text{ for } 1 \le i \le m$ 

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## **Competitive Ratio**

Threshold allocation with a skewed and semi-active schedule has the competitive ratio

 $m \cdot f_1(\varepsilon) + 1 \ge 2m + 1$  for  $f_1(\varepsilon) \ge 2$ .

•  $\varepsilon_T = \arg \max\{f_1(\varepsilon) = 2\}$  decreases with increasing *m*.

m	2	3	4	5
$\varepsilon_T$	0.2857	0.0900	0.0291	0.0098

Greedy allocation with a skewed and semi-active schedule has the competitive ratio

$$\frac{1}{m} + \frac{1+\varepsilon}{\varepsilon} \text{ for } 0 < \varepsilon \leq 1.$$



## Greedy Acceptance

Intuitively, greedy acceptance is the simplest approach.

• The competitive ratio of greedy acceptance is identical to the competitive ratio of the following min-threshold approach for  $0 < \varepsilon \leq 1$ :

$$d_{lim,i}\Big|_{t} = load(m_{i})\Big|_{t} \cdot \frac{1+\varepsilon}{\varepsilon} + t$$

$$d_{lim}|_t = \min_{1 \le i \le m} d_{lim,i} \Big|_t$$

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# **Proof Concepts**

- Key lemma for the (max-)threshold approach:
  - Allocation of a new job to a machine without the maximum outstanding load will turn this machine into the machine with the maximum outstanding load if  $f_1(\varepsilon) \ge 2$  holds.
- Partitioning of the resulting schedule into several intervals
  - We determine how much load of every interval cannot be executed outside of this interval in any optimal schedule that has accepted the corresponding jobs.



## **Previous Results**

• For m = 1, greedy acceptance (Goldwasser 1999, 2003) with a semiactive schedule has the tight competitive ratio

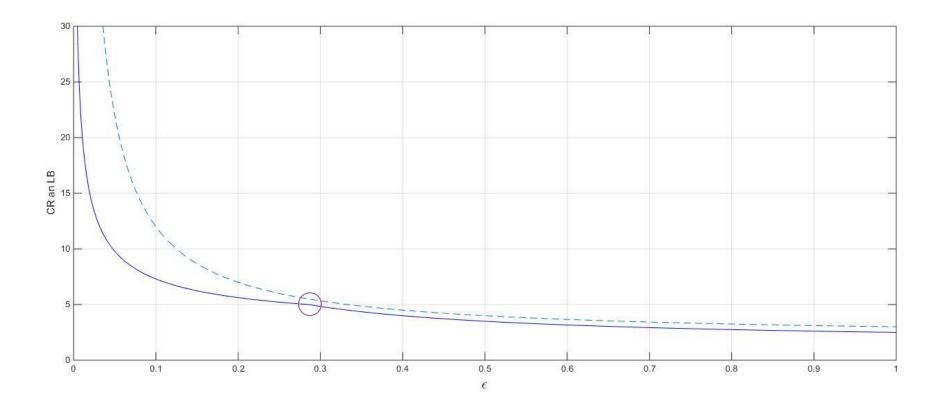
$$1 + \frac{1+\varepsilon}{\varepsilon}$$
 for  $0 < \varepsilon$ .

• For m > 1, greedy allocation with a balanced and semi-active schedule has the competitive ratio (Kim, Chwa 2001)

$$1 + \frac{1+\varepsilon}{\varepsilon}$$
 for  $0 < \varepsilon$ .



## Results for P2



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## **Competitive Ratio**

• Gap between greedy and (max-)threshold allocation at  $\varepsilon_T$ 

m	2	3	4	5
Gap	0	5.44	26.61	92.24

- For m > 2, we need m algorithms covering different intervals within (0,1].
- The algorithms are a combination of min-thresholds and max-thresholds.



#### Lower Bound for P2 and Greedy Allocation

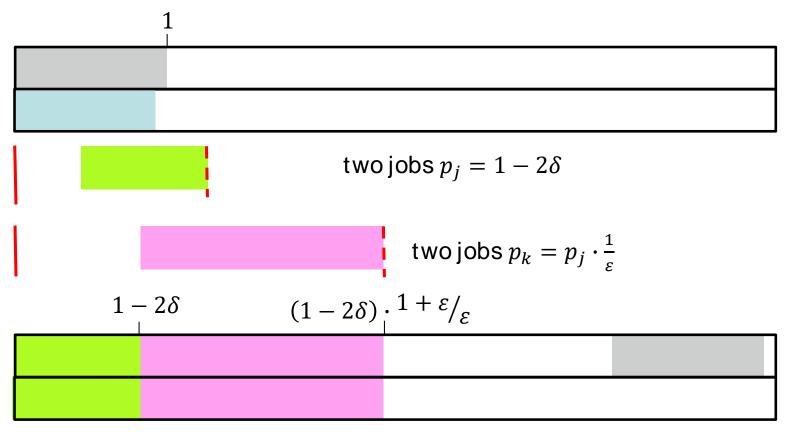


optimal schedule c = 5

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#### Lower Bound for P2 and Greedy Allocation

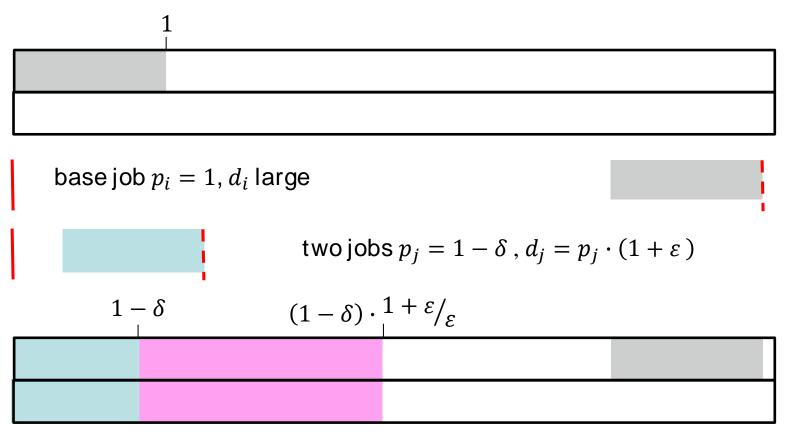


optimal schedule  $c = (1 + 2 \cdot \frac{1+\varepsilon}{\varepsilon})/2$ 

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#### Lower Bound for P2 and (Max-)Threshold Allocation

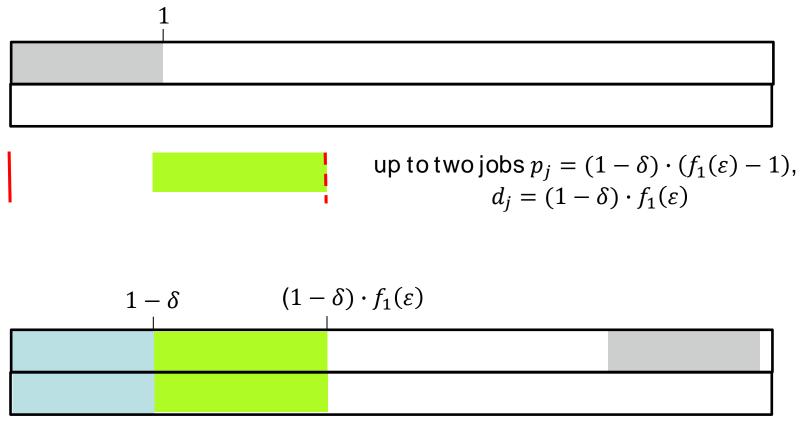


optimal schedule  $c = (1 + 2 \cdot \frac{1+\varepsilon}{\varepsilon})/2$ 

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#### Lower Bound for P2 and (Max-)Threshold Allocation

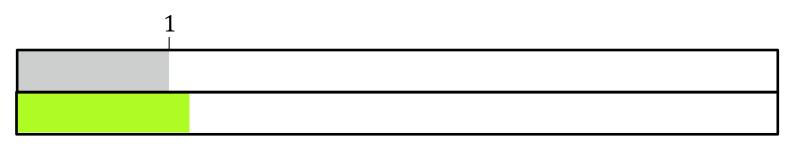


optimal schedule  $c = 1 + 2f_1(\varepsilon)$ 

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#### Lower Bound for P2 and (Max-)Threshold Allocation



two jobs 
$$p_k = (1 - \delta) \cdot \frac{1}{\varepsilon}$$
  
 $1 - \delta \qquad (1 - \delta) \cdot \frac{1 + \varepsilon}{\varepsilon}$ 

optimal schedule  $c = (1 + 2 \cdot \frac{1 + \varepsilon}{\varepsilon}) / f_1(\varepsilon) = 1 + 2f_1(\varepsilon)$ 

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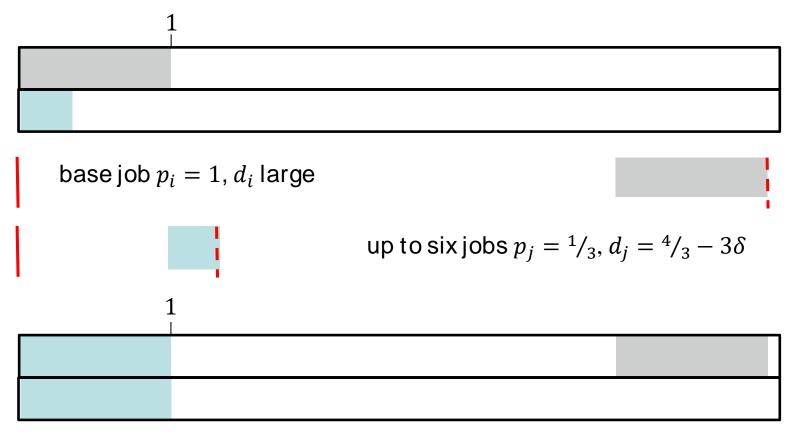


## Interval $(1, \infty)$ for $\varepsilon$

- The optimal competitive ratio is larger for any  $\varepsilon \in (0,1]$  than the optimal competitive ratio for any  $\varepsilon \in (1,\infty)$ .
  - The problem becomes easier for  $\varepsilon > 1$ ?
  - Previous results seem to support this claim.
- Observation: The presented optimal online algorithms for  $\varepsilon \in (0,1]$  only use semi-active schedules avoiding any start-time problem
- It is not possible to obtain the competitive ratio  $\frac{1}{m} + \frac{1+\varepsilon}{\varepsilon}$  for all  $\varepsilon \in (1, \infty)$  when using only semi-active schedules.
  - We consider an example with  $\varepsilon = 2$  and the P2 environment.
- Progression of time limits the competitive ratio for large  $\varepsilon$  and  $m \ge 3$ .



Lower Bound for P2,  $\varepsilon = 2$ , and Semi-active Schedules

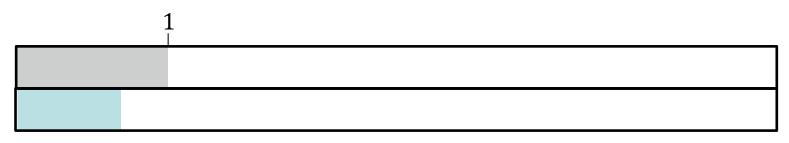


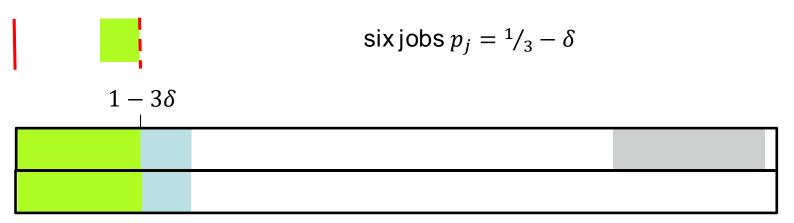
optimal schedule  $c = \frac{9}{4}$ 

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Lower Bound for *P*2,  $\varepsilon = 2$ , and Semi-active Schedules



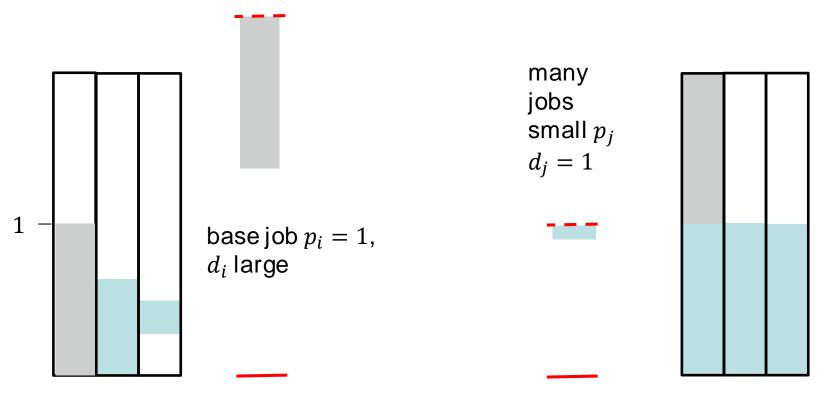


optimal schedule  $c = \frac{11}{5}$ 

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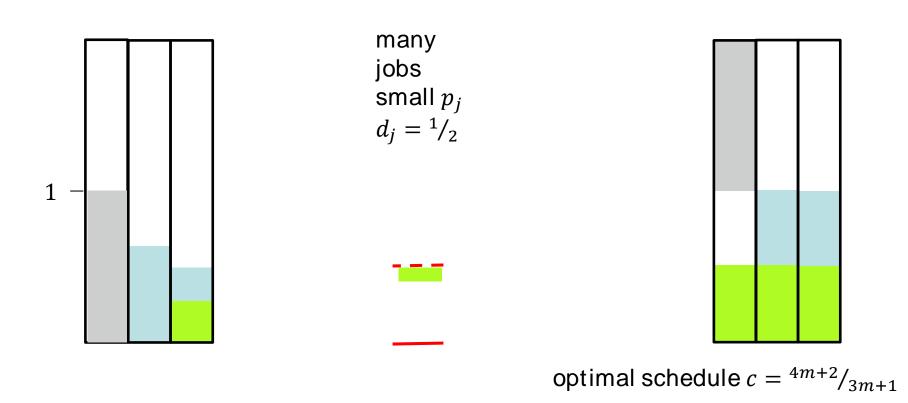
#### Lower Bound for Pm and Large $\varepsilon$



optimal schedule c = 2



#### Lower Bound for Pm and Large $\varepsilon$





## Conclusion

- For the problem  $Pm|online, \varepsilon, commit|\sum p_j \cdot (1-U_j)$ , we presented online algorithms with an optimal competitive ratio for  $0 < \varepsilon \leq 1$ .
- The optimal algorithm consists of m different algorithms that are each valid for a subinterval of (0,1] of  $\varepsilon$ .
- The algorithms use thresholds, skewed allocation and semi-active schedules.
- For  $\varepsilon > 1$ , the optimal competitive ratio is smaller but the algorithms are more complicated.