



# Parallel Preemptive Online Scheduling with Deadlines and Slack

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## A Basic Online Problem in a Large Computer System





## Dilemma of the Provider

Known request: utilization 1

0 1

10.5

#### Potential future request: utilization 10

Reject of future request due to contract obligation

Competitive ratio of 
$$\sum p_j \cdot (1 - U_j)$$
 is unbounded!  
acceptance indicator  $0/1$ 



# **Competitive Ratio**

- Worst case analysis of online algorithms: competitive analysis
  - Competitive ratio of algorithm *Alg*: max<sub>Instances I</sub>  $\frac{\sum p_j \cdot (1 U_j(Alg))}{\sum p_j \cdot (1 U_j(OPT))}$
- An online algorithm is optimal if the competitive ratio of the algorithm matches the lower bound of the competitive ratio for the online problem.



# A Little Flexibility: the Slack





## Formal Definition of Slack $\varepsilon$







0 1















# Content of this Talk

- Known Results
- Greedy Acceptance Strategy: If you can accept it, execute it!
- ALazy Acceptance Strategy: Do not accept all jobs that you can accept!
- The Sequence Problem: All jobs arrive at time 0 in a sequence.
- Lower Bound: The game of the adversary
- Progression of Time: It is quite a difference!
- Restrictions in Practice



# Known Results without Preemptions

- Single machine without preemptions
  - Greedy is 2 + <sup>1</sup>/ε competitive. The algorithm is optimal (Goldwasser 1999).
- Parallel identical machines without preemption
  - Transfer of the single machine algorithm to parallel identical machines (Kim and Chwa 2001).
  - No lower bound is known.
  - Lee (2003) suggested an algorithm with a claimed competitive ratio  $m + 1 + m \cdot \varepsilon^{-1/m}$ . The real competitive ratio is much larger.



#### Single Machine Bound without Preemptions





#### Single Machine Bound without Preemptions



optimal schedule

PPAM 17, Tuesday, September 12, 2017



# Known Results with Preemptions

- Single machine results with preemption
  - Greedy is  $1 + \frac{1}{\varepsilon}$  competitive. The bound is tight (Das Gupta and Palis 2000).
- Parallel identical machines with preemption and without migration
  - Transfer of the single machine algorithm to parallel identical machines (DasGupta and Palis 2000).
  - The competitive ratio is not correct for large slack values  $\varepsilon$ .
  - Lower bound  $1 + \frac{1}{m \cdot [1+\varepsilon]-1}$  (Das Gupta and Palis 2001).



## Single Machine Bound with Preemptions



#### optimal schedule



# A Simple Check for a Legal Schedule

 $V_{min}(t)$ : minimum amount of processing volume in interval [0,t]





Greedy Acceptance Policy on *m* Parallel Identical Machines

- We accept a new job if there is a valid schedule that completes the new job and all previously accepted jobs in time.
- We use the  $V_{min}$  criterion to test whether there is a valid schedule.
- The worst case single machine scenario applies as well!



#### Greedy Acceptance Policy on 2 Parallel Identical Machines



#### optimal schedule

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#### Lazy Acceptance Policy

- We do not accept every job although it may allow a legal schedule.
- We exchange the criterion  $V_{min}(t)$  by a criterion  $V_{sim}(t) \le V_{min}(t)$ .
- We exchange our threshold  $m \cdot t$  by  $F(m, \varepsilon) \cdot t \leq m \cdot t$ .

$$V_{sim}(t) = \sum_{j \in J} \begin{cases} 0 & \text{for } d_j > t \\ p_j & \text{for } d_j \le t \end{cases}$$
$$f(m, \varepsilon) = \frac{\varepsilon}{1 + \varepsilon} \cdot \sum_{i=0}^{m-1} \left(\frac{1 + \varepsilon}{\varepsilon}\right)^{\frac{i}{m}} \qquad F(m, \varepsilon) = \left(\frac{1 + \varepsilon}{\varepsilon}\right)^{\frac{1}{m}} \cdot f(m, \varepsilon)$$



# Proof of the Existence of a Legal Schedule (Key Lemma)

- We test whether  $V_{min}(d_j) \le d_j \cdot m$  always holds if the new criterion does not exceed the new threshold.
- We reduce the instance space that we must examine.
  - We apply transformations that cannot decrease  $V_{min}(d_j)$  while they do not increase  $V_{sim}(d_i)$  for any  $d_i > d_j$ .
- We analyze all job sequences of the reduced instance space.



#### Job Removal Transformation





## Spread Generation Transformation





# V<sub>0</sub>-Transformation





# **V-Transformation**





# p-Transformation





# Worst Case Sequence

$$1. \quad p_{>j} \le \frac{d_{>j}}{1+\varepsilon}$$

2. 
$$p_i = \frac{d_i}{1+\varepsilon}$$
 for every deadline  $d_i > d_{>j}$ 

3. 
$$V_{sim}(d_i) = d_i \cdot F(m, \varepsilon)$$
 for all  $d_i \ge d_j$ .

- 4. There are no jobs i with  $d_i p_j \ge d_j$ .
- 5. There are no two jobs with the same deadline  $d_i > d_j$ .

Sequence of jobs with geometrically increasing deadlines and tight slack 
$$d_i = d_{>i} \cdot \left(\frac{1+\varepsilon}{\varepsilon}\right)$$

 $\frac{1}{m}$ 



# Algorithm Limit test

- The algorithm uses  $t_{min} = \operatorname{argmax}_{\tau} \{ V_{min}(\tau) |_t = (\tau t) \cdot f(m, \varepsilon) \}.$
- For each submitted job j do
  - use  $r_j = t$  to determine  $t_{min}$ ;
  - if  $d_j \ge t_{min}$  then
    - accept job j; update t<sub>min</sub>;
  - else
    - reject job j;
  - end if
- end for



# Sequence Problem

- A sequence problem is an online problem in which all jobs are submitted at time 0.
  - But we must make our decision on any job before seeing the next job.
- In some online problems, worst cases occur in the corresponding sequence problem.
  - 1| $\varepsilon$ , online, pmtn| $\sum p_j \cdot (1 U_j)$
- In other online problems, progression of time leads to more complexity and a larger competitive ratio.
  - 1| $\varepsilon$ , online|  $\sum p_j \cdot (1 U_j)$



# Competitive Ratio with Preemption and Migration





## Competitive Ratio and the Number of Machines





Lower Bound Base Phase

- Submission of many small jobs
- The adversary stops the submission if the planned area is covered with these jobs.
- If we do not accept enough small jobs then the adversary submits enough jobs to cover the whole area until the deadline.





# Lower Bound Filling Phase

- Submission of long jobs with tight slack.
- The adversary stops the submission if we accept one job.
- If we do not accept a job then the adversary submits enough jobs to cover the area  $[m \cdot (1 + \varepsilon)] \cdot p_j$ .





Lower Bound Completion

- Repetition of this procedure with exponentially increasing processing times and deadlines.
- For the final job type, the adversary first exponentially increases the processing time and then reduces it by a very small amount δ. The adversary selects the deadline to generate a tight slack. We cannot accept any of these jobs.





# Slack and the Progression of Time





# Modification of the Key Lemma

- We replace  $V_{sim}$  by a more complex criterion  $V_{acc}$ .
- We add a new transformation *large job splitting*.
- The lemma still holds.

$$V_{acc}(t) = \sum_{j \in J} \begin{cases} 0 & \text{for } d_j - p_j \ge t \\ t - d_j + p_j & \text{for } d_j - \frac{d_j}{1 + \varepsilon} \ge t > d_j - p_j \\ max \left\{ p_j - \frac{d_j}{1 + \varepsilon}, 0 \right\} & \text{for } d_j > t > d_j - \frac{d_j}{1 + \varepsilon} \\ p_j & \text{for } t \ge d_j \end{cases}$$



# Large Job Splitting Transformation





Impact of the Schedule

- The schedule has an impact on worst case sequence situations with a small competitive ratio (here large slack values).
- Many small jobs with deadline 1 and a total processing time  $m/_2$ .
- $m/_2$  jobs with processing time 1 and suitably large deadline.
- There are two basic strategies for allocation
  - Balancing
  - Concentration
- Greedy scheduling with preemption has a worse competitive ratio on parallel identical machines than on a single machine for large slack values.



# Allocation with Concentration Strategy

our schedule

optimal schedule





# Allocation with Balancing Strategy

our schedule

optimal schedule





Competitive Ratio of the Greedy Acceptance Policy

• The greedy acceptance policy for the problem  $P_m[\varepsilon, \text{pmtn}, \text{online}] \sum p_j \cdot (1 - U_j)$  has a competitive ratio of at least

$$\frac{4\varepsilon^2 + 14\epsilon + 2}{4\varepsilon^2 + 9}.$$

- This term is larger than  $\frac{1+\varepsilon}{\varepsilon}$  for  $\varepsilon > 7$ .
- The result also holds for preemption without migration since the proof does not use migration. Therefore, the result corrects the claim by Das Gupta and Palis (2000).



# Influence of Progression of Time on the Threshold

- Our algorithm guarantees the validity of the threshold for all times in the sequence problem:  $V_{sim}(t) \le F(m, \varepsilon) \cdot t$
- After progression of time, this condition may not hold anymore.
  - The condition for our key lemma is not true anymore!
- We can prove that the algorithm still works when we use allocation with concentration strategy.

• Competitive ratio 
$$max\left\{\frac{m \cdot (1+\varepsilon)}{f(m,\varepsilon)}, \frac{m \cdot (1+\varepsilon)}{4f(m,\varepsilon)}+1\right\}$$

Impact of allocation with concentration strategy for large values of  $\varepsilon$ .



# **Restrictions in Practice**

- Preemption with migration for resource management
  - We use migration in case of failures. Can we use it for resource management?
- Heterogeneity of resources
  - Most large computer systems are not homogeneous.
- Multiple resources
  - The analysis only considers a single resource. How about computation and bandwidth and storage?
- Priority of some jobs
  - Can we handle jobs with different priorities?
- Type of jobs
  - How can we handle interactive or parallel jobs?