

Approximation in Deterministic and Stochastic Machine Scheduling

Wissenschaftliche Aussprache

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Vorsitzender: Prof. Dr. Yuri Suris

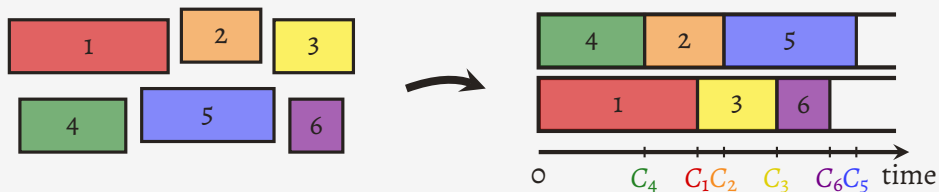
Gutachter: Prof. Dr. Martin Skutella

Prof. Dr. Marc Uetz

Scheduling on Identical Parallel Machines

Given: jobs $j = 1, \dots, n$ with processing times $p_j \geq 0$ and weights $w_j \geq 0$, number of machines m ;

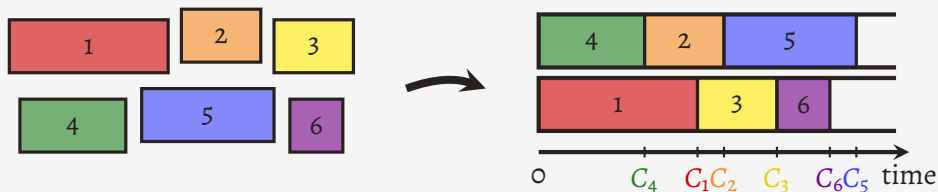
Task: schedule each job for p_j time units on some machine so as to minimize the sum of weighted completion times $\sum_{j=1}^n w_j \cdot C_j$.



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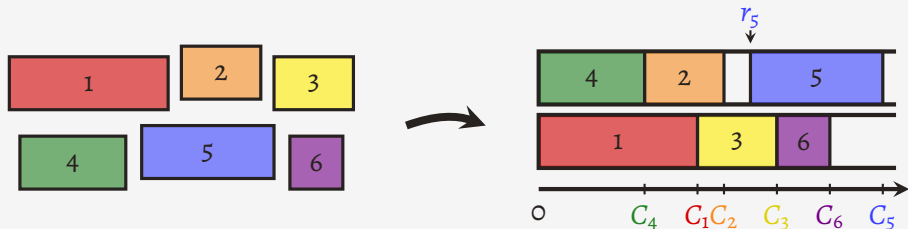
Complexity

- NP-hard for two machines (Bruno et al. 1974)
- strongly NP-hard for arbitrary number of machines (Lageweg, Lenstra 1977)

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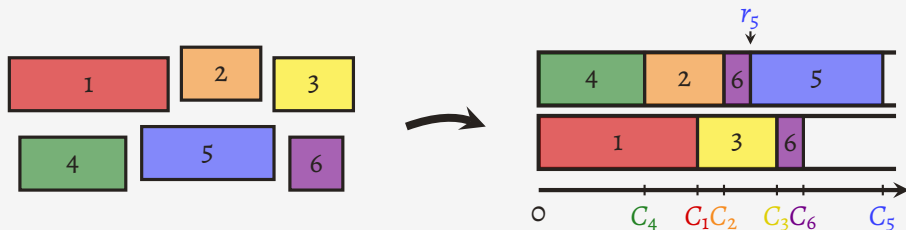
Problem variants

- release dates $r_j \geq 0$

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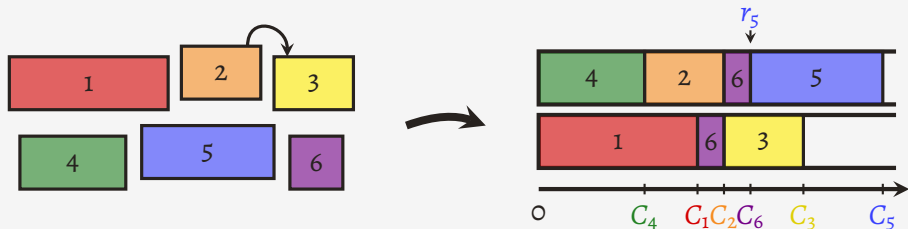
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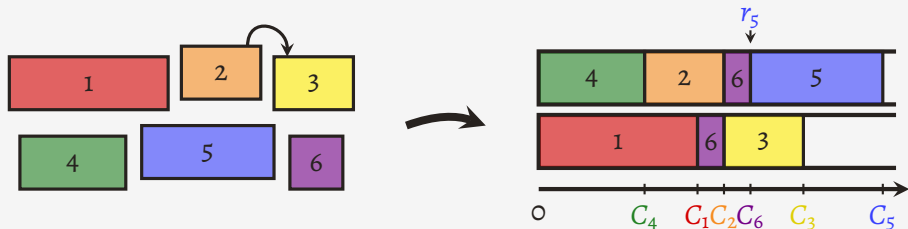
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- release dates $r_j \geq 0$
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Problem variants

- release dates $r_j \geq 0$
- preemptive vs. non-preemptive
- precedence constraints/delays
- heterogeneous machines

Scheduling under Uncertainty

I. Adversarial models

- online job arrivals
- unknown processing times (non-clairvoyant scheduling)

Competitive analysis: compare to offline optimum

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II. Stochastic models

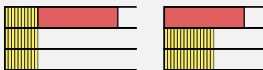
- random processing times with known distributions

Minimize expected total weighted completion time

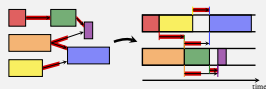
Overview

I. Deterministic Scheduling

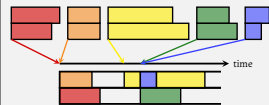
WSPT Rule



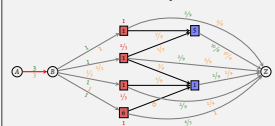
Scheduling with Precedence Delays



Online Non-Migratory Preemptive Scheduling on Unrelated Machines

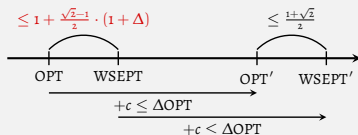


Non-Clairvoyant Online Scheduling with Precedence Delays

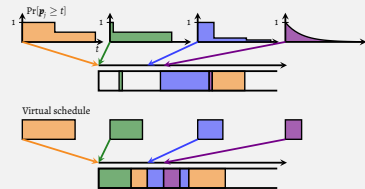


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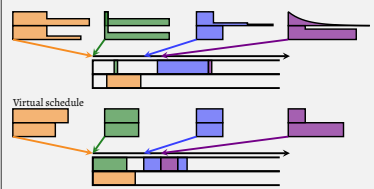
WSEPT Rule



Stochastic Scheduling on Single Machine with Release Dates



Stochastic Online Scheduling on Unrelated Machines



Part I

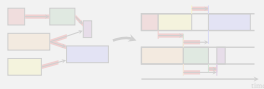
Deterministic Scheduling

Main Results

WSPT Rule



Scheduling with Precedence Delays



Online Non-Migratory Preemptive Scheduling on Unrelated Machines



Non-Clairvoyant Online Scheduling with Precedence Delays



Weighted Shortest Processing Time First (WSPT) rule

Whenever a machine is free, start available job with maximum ratio w_j/p_j on it.

For a fixed number m of machines, what is the maximum possible approximation ratio of the WSPT rule?

WSPT Rule – Previous Work

Theorem (Kawaguchi, Kyan 1986)

For an arbitrary number of machines WSPT rule has tight approximation ratio $\frac{1+\sqrt{2}}{2} \approx 1.207$.

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- Worst case is attained if all jobs have $w_j = p_j$.
- In this case WSPT is list scheduling according to the input order.
- Worst case is attained when $n, m \rightarrow \infty$.

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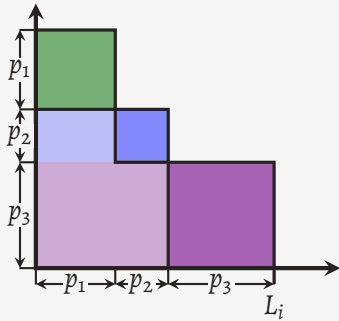
- There is a PTAS for scheduling jobs on an arbitrary number of identical machines. (Skutella, Woeginger 2000)
- For any number m of machines there is an FPTAS for scheduling jobs on m identical machines. (Sahni 1976)

Reduction to Unit Smith Ratio

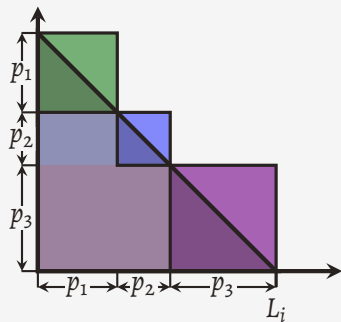
Lemma (Kawaguchi, Kyan 1986; Schwiegelshohn 2011)

On any number of machines the worst-case approximation ratio is attained when $w_j = p_j$ for all j .

Objective Function in Terms of Machine Loads (for $w_j = p_j$)

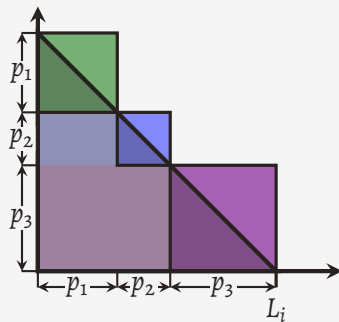


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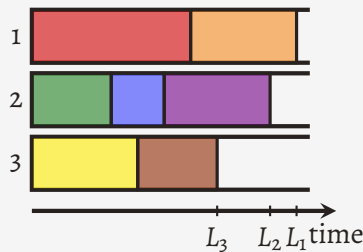


one machine i :

$$\sum_{j \rightarrow i} p_j \cdot C_j = \frac{1}{2} \left(\underbrace{\sum_{j \rightarrow i} p_j}_{=: L_i} \right)^2 + \frac{1}{2} \sum_{j \rightarrow i} p_j^2$$

Objective Function in Terms of Machine Loads (for $w_j = p_j$)one machine i :

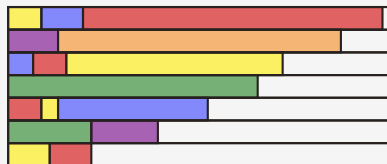
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 m -machine schedule:

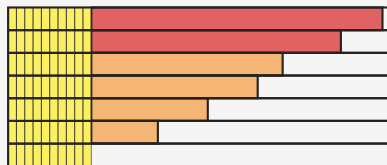
$$\sum_{j=1}^n p_j \cdot C_j = \frac{1}{2} \sum_{i=1}^m L_i^2 + \frac{1}{2} \sum_{j=1}^n p_j^2$$

Reduction to Worst Case

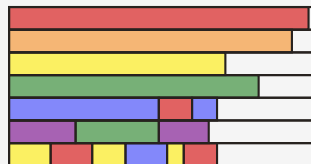
WSPT schedule



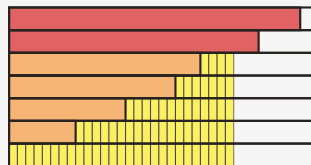
$\sum_i L_i^2$ unchanged
 $\sum_j p_j^2$ decreases by δ



OPT schedule

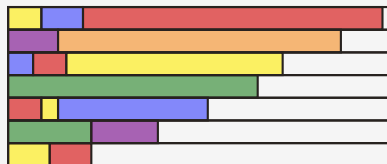


$\sum_i L_i^2$ not increased
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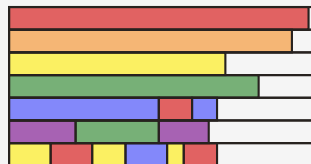


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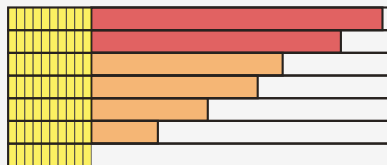
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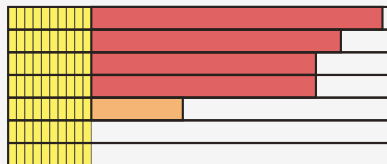


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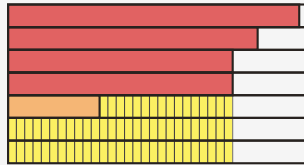
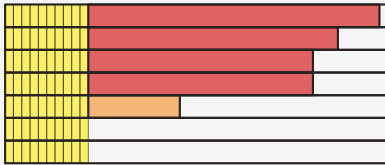
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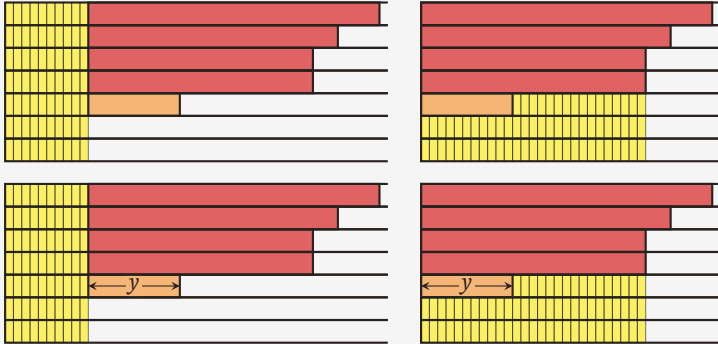


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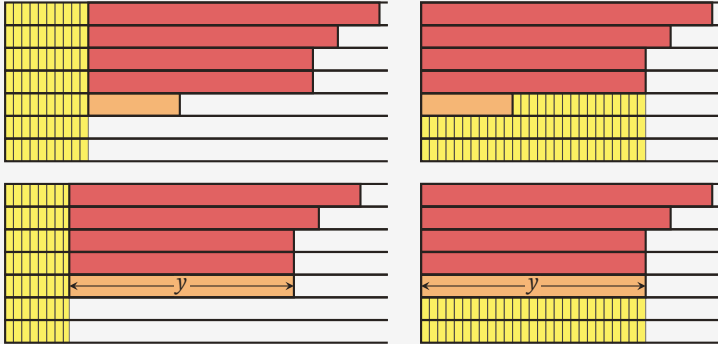
Reduction to Worst Case (Cont.)



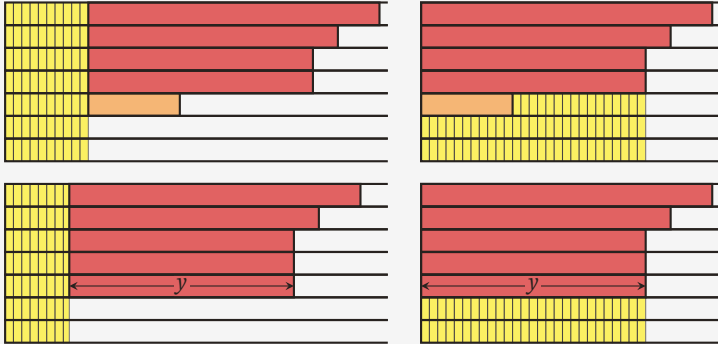
Reduction to Worst Case (Cont.)



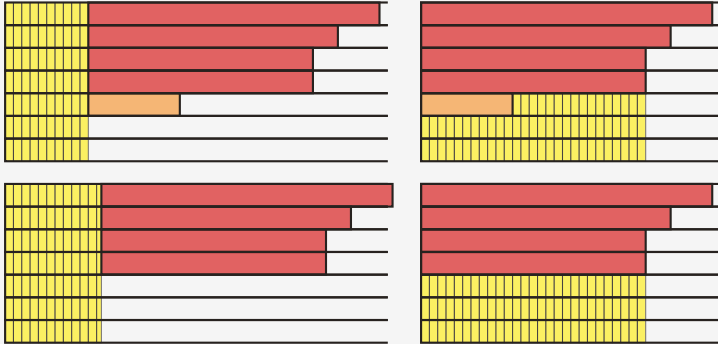
Reduction to Worst Case (Cont.)



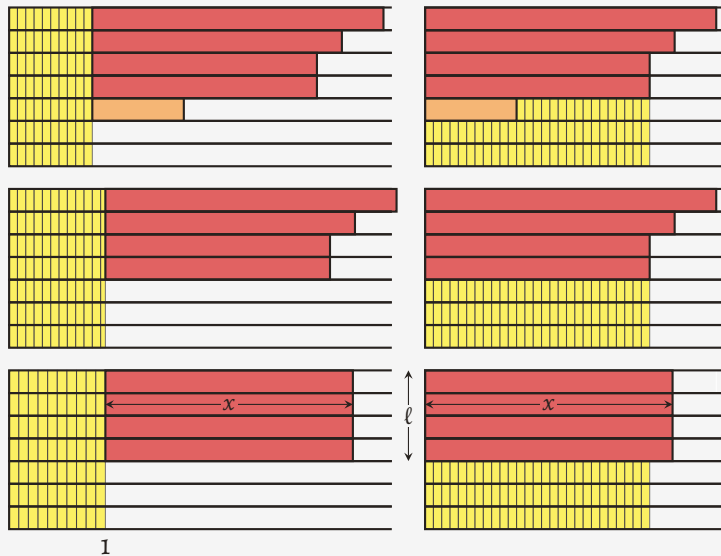
Reduction to Worst Case (Cont.)



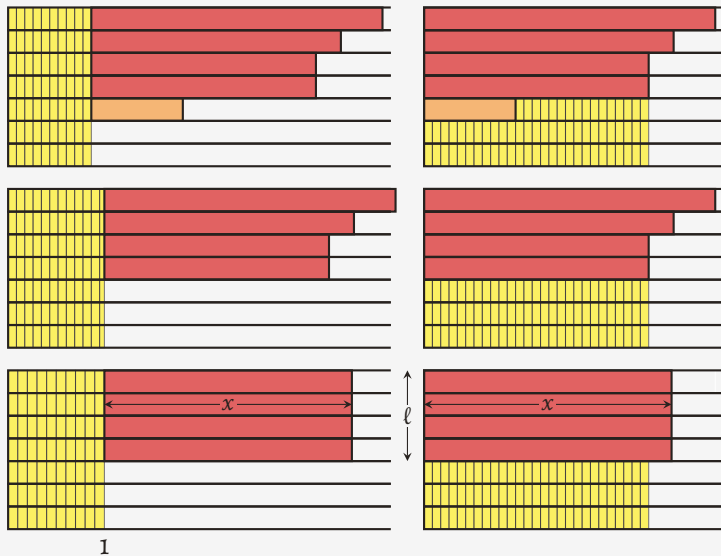
Reduction to Worst Case (Cont.)



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Reduction to Worst Case (Cont.)



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$\sum_j p_j^2$ decreases by δ

1

Find worst ℓ, x by calculation.

$\sum_i L_i^2$ decreases by δ

$\sum_j p_j^2$ decreases by δ

Approximation Ratio of the WSPT Rule on Fixed Number of Machines

Theorem

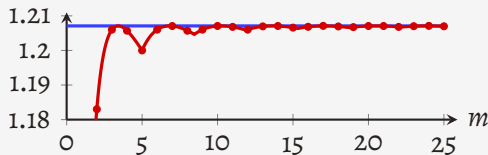
On $m \geq 2$ machines the worst case for the WSPT rule occurs for

1. m/ε small jobs with $w_j = p_j = \varepsilon$ ($\varepsilon \rightarrow 0$),

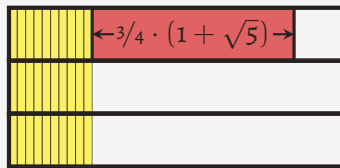
2. $\ell := \lfloor \frac{2m - \sqrt{2m^2 - 1}}{2} \rfloor$ jobs with $w_j = p_j = \frac{m}{\sqrt{\ell \cdot (2m - \ell)} - \ell}$.

In this case the approximation ratio is

$$1 + \frac{\sqrt{\ell \cdot (2m - \ell)} - \ell}{2m}.$$

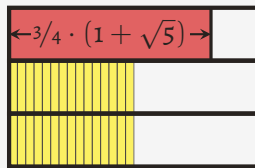


WSPT schedule



1

OPT schedule

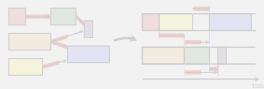


Main Results

WSPT Rule



Scheduling with Precedence Delays



Online Non-Migratory Preemptive Scheduling on Unrelated Machines



Non-Clairvoyant Online Scheduling with Precedence Delays



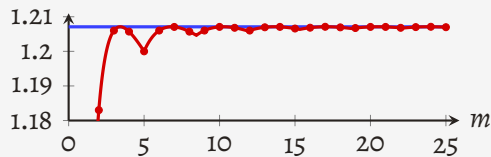
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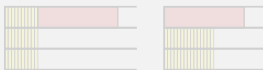
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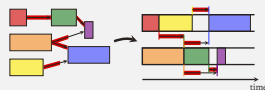


Main Results

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Non-Clairvoyant Online Scheduling with Precedence Delays



Theorem (Li 2020)

There is a 3.386-approximation algorithm for scheduling jobs with precedence constraints.

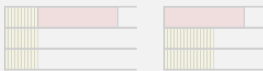
Theorem

*There is a 3.386-approximation algorithm for scheduling jobs with precedence **delays**.*

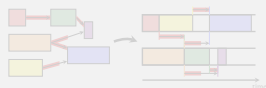
- includes the setting with precedence constraints and release dates.

Main Results

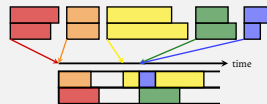
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Theorem

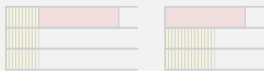
The Greedy-Assignment WSRPT algorithm is a 4-competitive deterministic online algorithm for non-migratory preemptive scheduling on unrelated machines.

Previous results

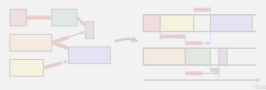
- 7.216-competitive deterministic online algorithm (Gupta et al. 2021+)
- 5.771-competitive randomized online algorithm (Hall et al. 1997; Chakrabarti et al. 1996)
- With $1 + \varepsilon$ speed augmentation, Greedy-Assignment WSRPT algorithm is $(2 + \frac{2}{\varepsilon})$ -competitive for $\sum w_j(C_j - r_j)$. (Anand et al. 2012)
 \rightsquigarrow 8-competitive alg. for $\sum w_j C_j$. (Bansal, Pruhs 2004)

Main Results

WSPT Rule



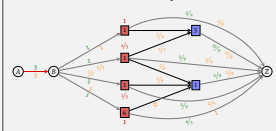
Scheduling with Precedence Delays



Online Non-Migratory Preemptive Scheduling on Unrelated Machines



Non-Clairvoyant Online Scheduling with Precedence Delays



Theorem (Garg et al. 2019)

There is a 10-competitive deterministic non-clairvoyant online algorithm for jobs with precedence constraints.

Theorem

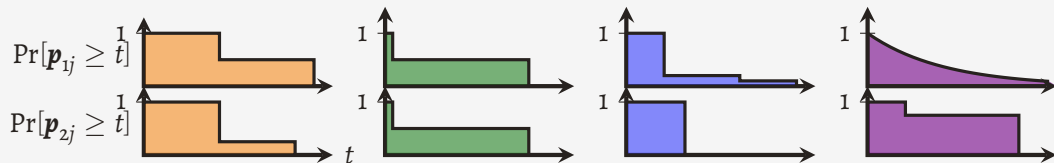
*A simplification of the algorithm of Garg et al. is **8-competitive** for jobs with precedence **delays**.*

Part II

Stochastic Scheduling

Stochastic Scheduling on Unrelated Parallel Machines

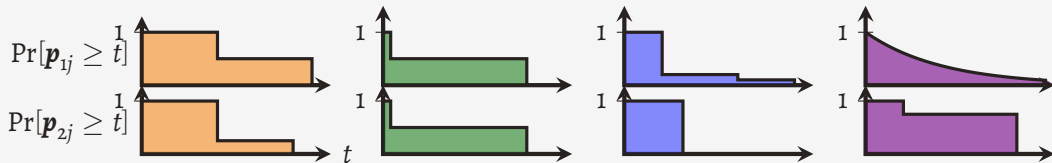
Given: weights $w_j \geq 0$ and **distributions** of independent random processing times $p_{ij} > 0$ of jobs $j = 1, \dots, n$ on machines $i = 1, \dots, m$



Task: find non-preemptive **scheduling policy** Π minimizing the **expected** sum of weighted completion times.

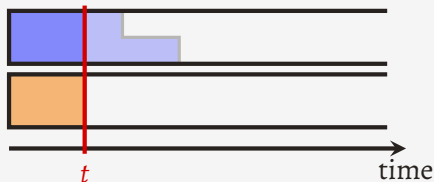
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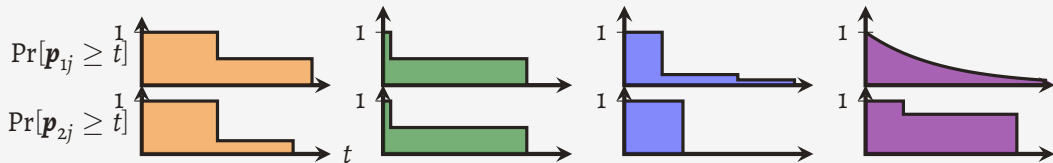
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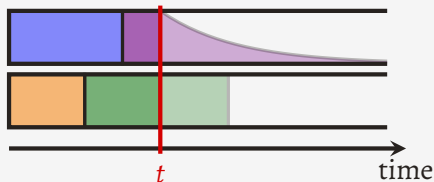
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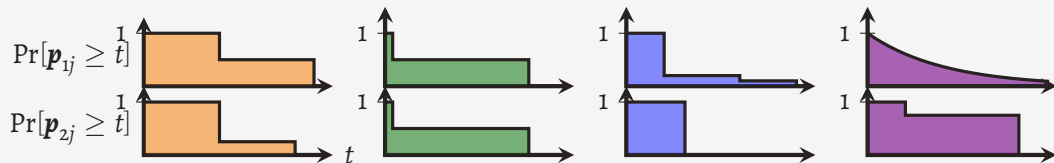
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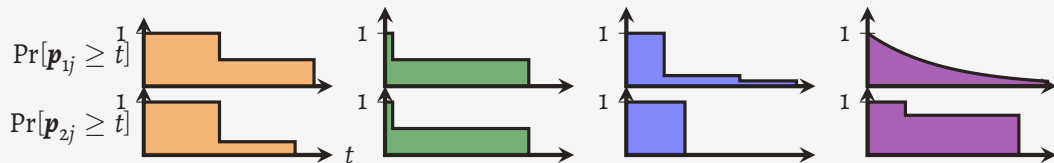


Task: find non-preemptive **scheduling policy** Π minimizing the **expected** sum of weighted completion times.

- must be **non-anticipative**, i.e., a decision made at time t may only depend on the information known at time t
- function $\mathbb{R}^{mn} \rightarrow \mathbb{R}; (p_{ij}) \mapsto \sum_{j=1}^n w_j \cdot C_j^{\Pi}(p_{ij})$ must be **measurable**.

Stochastic Scheduling on Unrelated Parallel Machines

Given: weights $w_j \geq 0$ and **distributions** of independent random processing times $p_{ij} > 0$ of jobs $j = 1, \dots, n$ on machines $i = 1, \dots, m$



Task: find non-preemptive **scheduling policy** Π minimizing the **expected** sum of weighted completion times.

- must be **non-anticipative**, i.e., a decision made at time t may only depend on the information known at time t
- function $\mathbb{R}^{mn} \rightarrow \mathbb{R}; (p_{ij}) \mapsto \sum_{j=1}^n w_j \cdot C_j^{\Pi}(p_{ij})$ must be **measurable**.

Alternative definitions of scheduling policy (cp. Möhring, Radermacher 1985)

- function $\mathbb{R}^{mn} \rightarrow \mathbb{R}_{\geq 0}^n \times [m]^n$;
- function from state space to action space.

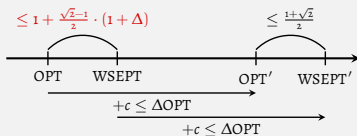
Approximative Policies

- We are interested in simple approximative policies.
- Some performance guarantees depend on an upper bound Δ on the squared coefficients of variation

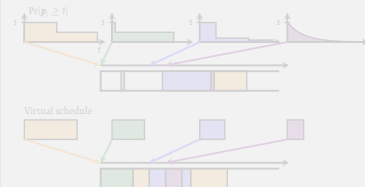
$$CV[\mathbf{p}_{ij}]^2 = \frac{\text{Var}[\mathbf{p}_{ij}]}{E[\mathbf{p}_{ij}]^2}.$$

Main Results

WSEPT Rule



Stochastic Scheduling on Single Machine with Release Dates



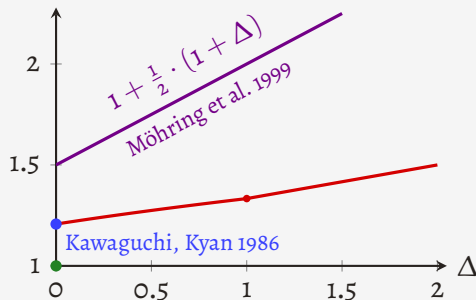
Stochastic Online Scheduling on Unrelated Machines



Theorem

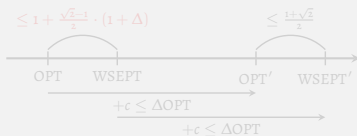
The WSEPT rule has performance guarantee

$$\begin{cases} 1 + \frac{1}{2 \cdot (1 + \sqrt{2(1 + \Delta)})} \cdot (1 + \Delta) & \text{if } \Delta \leq 1; \\ 1 + \frac{1}{6} \cdot (1 + \Delta) & \text{if } \Delta \geq 1. \end{cases}$$

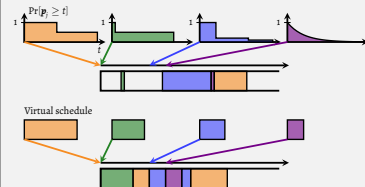


Main Results

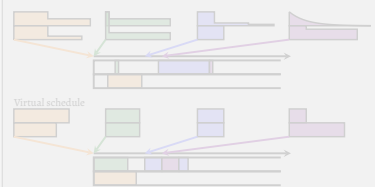
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Stochastic Scheduling on Single Machine with Release Dates



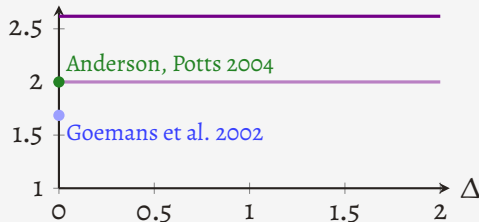
Stochastic Online Scheduling on Unrelated Machines



Theorem (Schulz 2008)

There are a $(1 + \max\{\phi, \frac{\phi+1}{2}(1 + \Delta)\})$ -competitive deterministic and a $(2 + \Delta)$ -competitive randomized online scheduling policy for *identical machines*.

- on single machine $(\phi + 1)$ -competitive (deterministic) and 2-competitive (randomized) ($\phi + 1 \approx 2.618$)

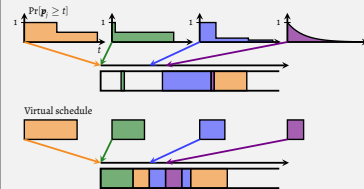


Main Results

WSEPT Rule



Stochastic Scheduling on Single Machine with Release Dates



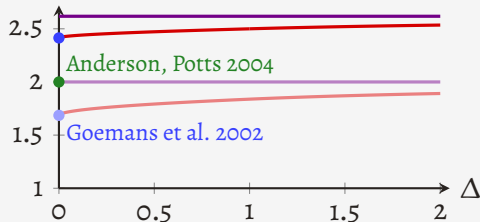
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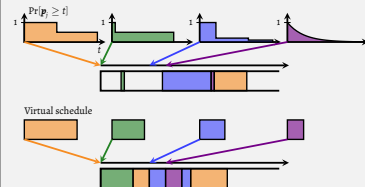
- If Δ is known in advance (*semi-online*), these can be improved.

Main Results

WSEPT Rule



Stochastic Scheduling on Single Machine with Release Dates

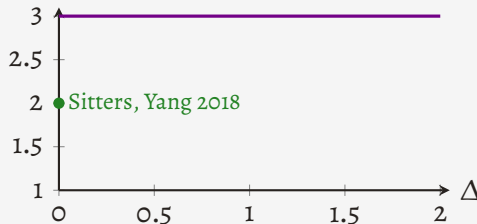


Stochastic Online Scheduling on Unrelated Machines



Theorem (Möhring et al. 1999)

There is an efficient scheduling policy with performance guarantee 3 for scheduling jobs with precedence constraints and release dates on a single machine.

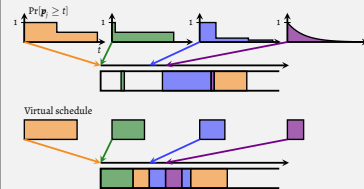


Main Results

WSEPT Rule



Stochastic Scheduling on Single Machine with Release Dates

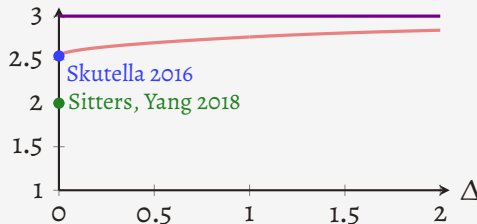


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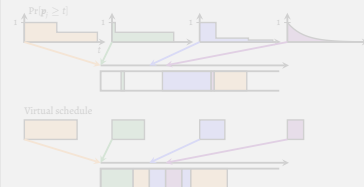


Main Results

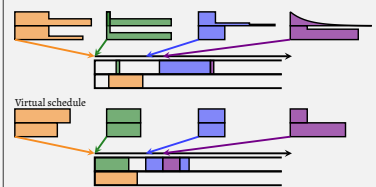
WSEPT Rule



Stochastic Scheduling on Single Machine with Release Dates



Stochastic Online Scheduling on Unrelated Machines



Theorem (Gupta et al. 2021+)

There is a $3.608 \cdot h(\Delta) \cdot (2 + \Delta)$ -competitive deterministic online scheduling policy for unrelated machines, where

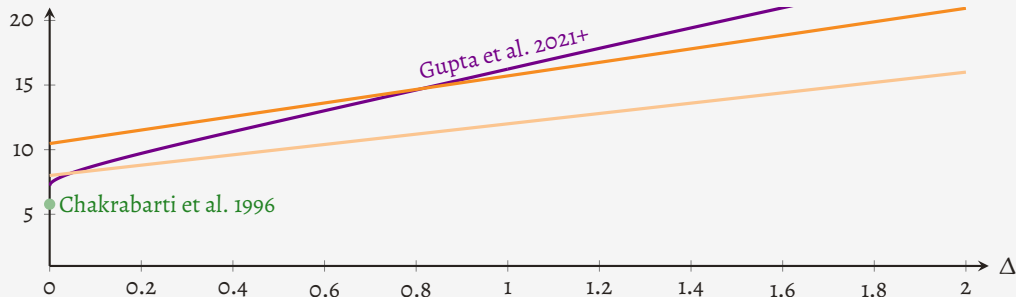
$$h(\Delta) := \begin{cases} 1 + \frac{\sqrt{\Delta}}{2} & \text{if } \Delta \leq 1; \\ 1 + \frac{\Delta}{\Delta+1} & \text{if } \Delta \geq 1. \end{cases}$$

Theorem

There is a $(3 + \sqrt{5}) \cdot (2 + \Delta)$ -competitive deterministic and an $(8 + 4\Delta)$ -competitive *randomized* online scheduling policy for unrelated machines.

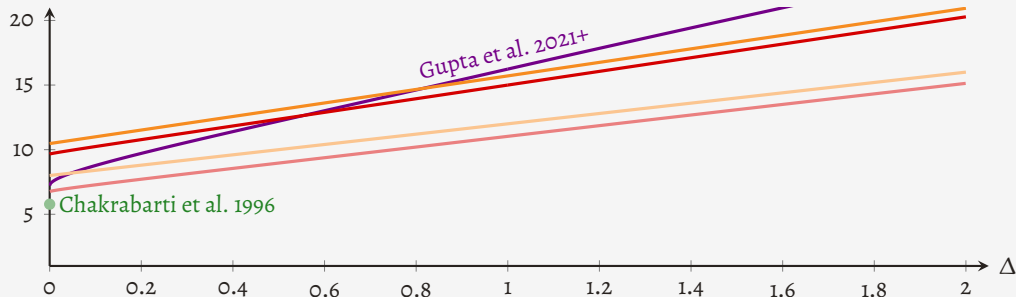
- If Δ is known in advance (semi-online), these can be improved.

Stochastic Online Scheduling on Unrelated Machines



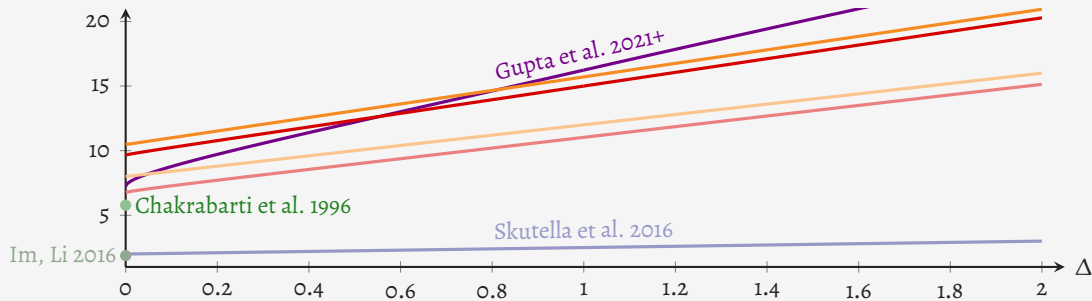
- Jobs j arrive over time;
- at the release date r_j the weight w_j and the distributions of all \mathbf{p}_{ij} , $i \in [m]$, are given;
- if j is scheduled on machine i , the outcome of \mathbf{p}_{ij} becomes known when j completes.
- *Competitive analysis*: compare to optimal scheduling policy.

Stochastic Online Scheduling on Unrelated Machines



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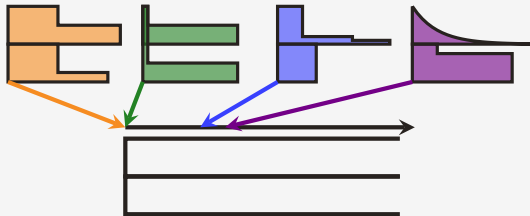
Stochastic Online Scheduling on Unrelated Machines



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Online Scheduling Policies

- The *mean busy time* M_j of job j is the average of all times when it is being processed.

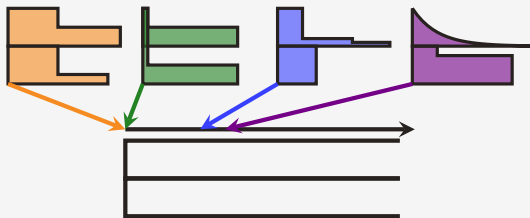


Online Scheduling Policies

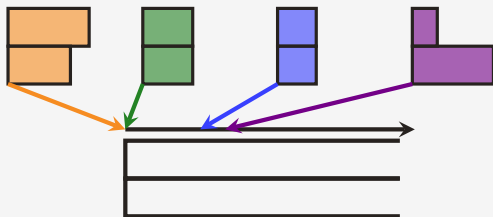
- The *mean busy time* M_j of job j is the average of all times when it is being processed.
- When job j is released, assign it to a machine i with minimum increase of

$$\sum_{j=1}^n w_j \cdot \left(M_j + \frac{p_{ij}}{2} \right)$$

in the virtual preemptive WSPT schedule of deterministic counterparts with $p_{ij} := E[p_{ij}]$.



Virtual schedule

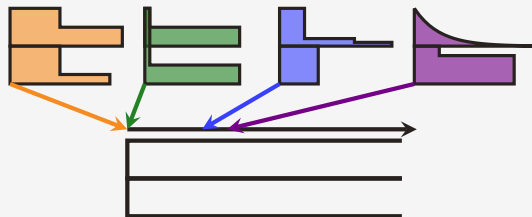


Online Scheduling Policies

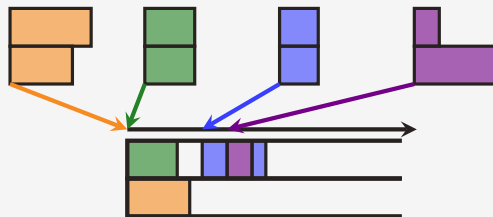
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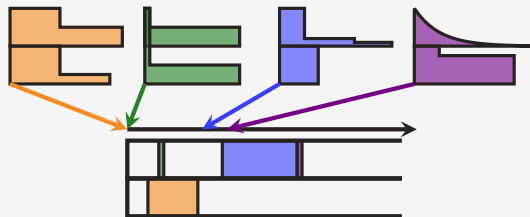
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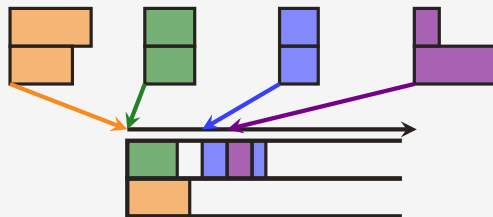
$$\sum_{j=1}^n w_j \cdot \left(M_j + \frac{p_{ij}}{2} \right)$$

in the virtual preemptive WSPT schedule of deterministic counterparts with $p_{ij} := E[p_{ij}]$.

- On each machine, schedule the jobs by single machine (semi-)online policy from previous chapter.



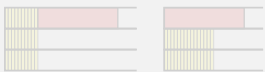
Virtual schedule



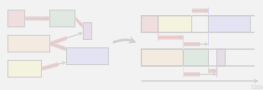
Overview

I. Deterministic Scheduling

WSPT Rule



Scheduling with Precedence Delays



Online Non-Migratory Preemptive Scheduling on Unrelated Machines



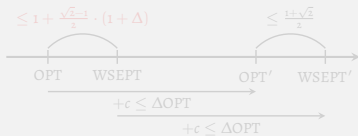
Non-Clairvoyant Online Scheduling with Precedence Delays



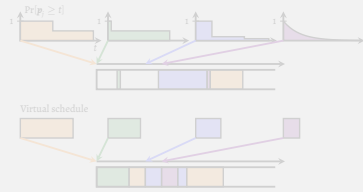
Thank you!

II. Stochastic Scheduling

WSEPT Rule



Stochastic Scheduling on Single Machine with Release Dates



Stochastic Online Scheduling on Unrelated Machines



Additional Slides

Details on WSPT Rule

Clairvoyant Online List Model

Scheduling with Precedence Delays on Identical Machines

Clairvoyant Online Time Model

Analysis of the WSEPT rule

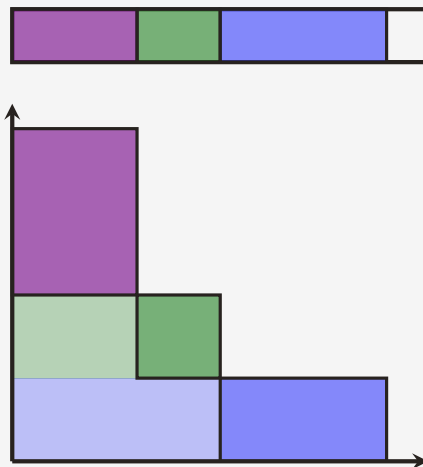
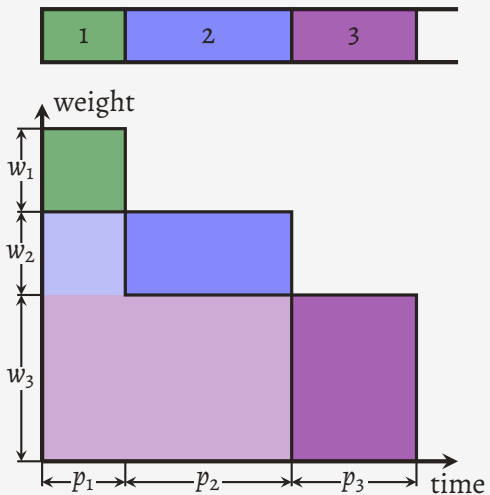
Stochastic Online Scheduling on Unrelated Machines

Stochastic Scheduling with Precedence Constraints

References

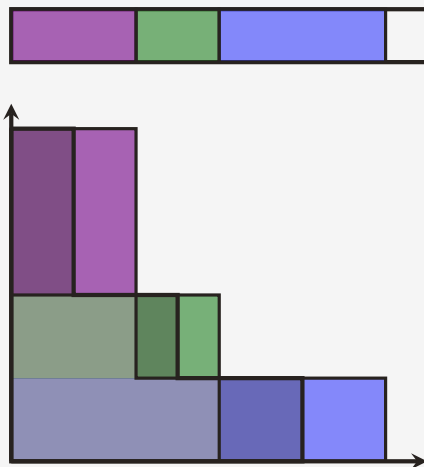
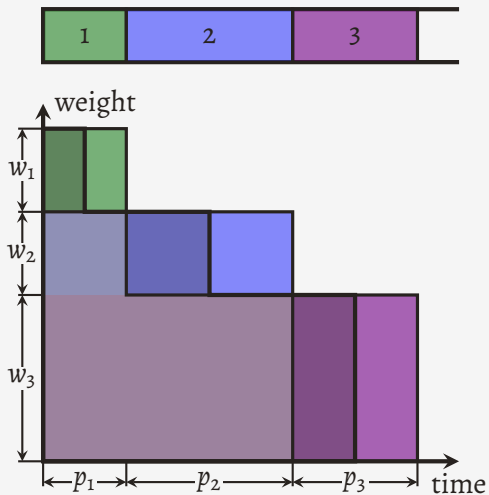
Two-Dimensional Gantt Charts

Eastman et al. 1964



Two-Dimensional Gantt Charts

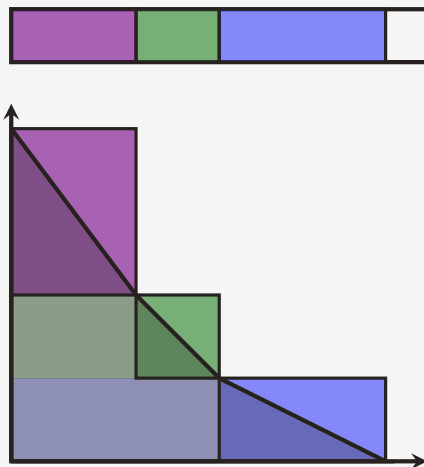
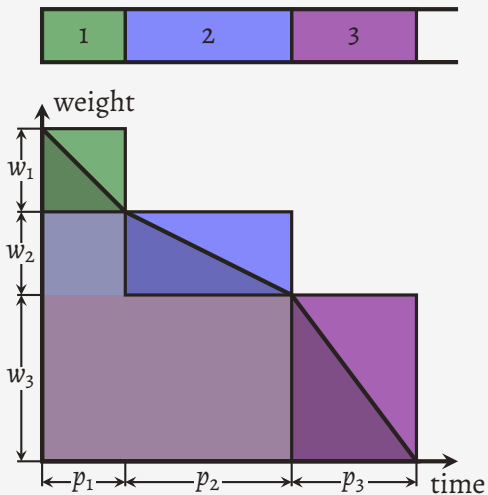
Eastman et al. 1964



$$\sum_{j=1}^n w_j \cdot C_j = \sum_{j=1}^n w_j \cdot M_j + \frac{1}{2} \cdot \sum_{j=1}^n w_j \cdot p_j$$

Two-Dimensional Gantt Charts

Eastman et al. 1964

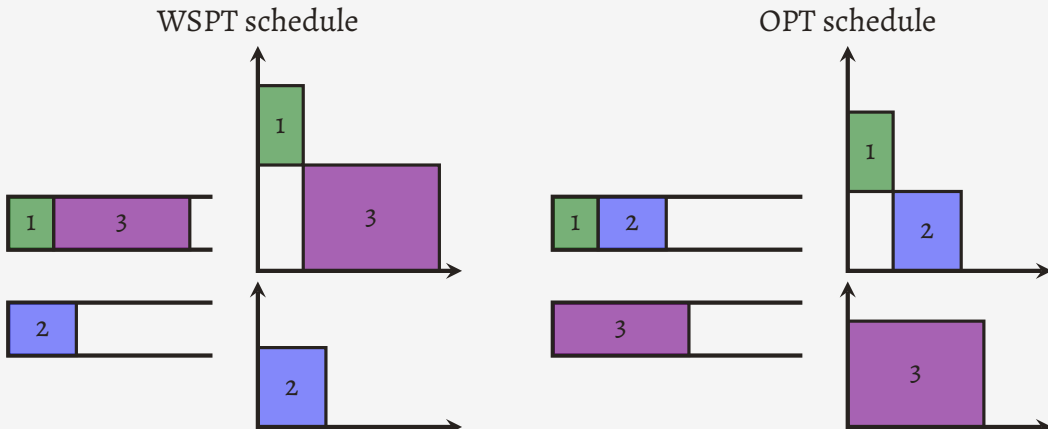


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Reduction to Unit Smith Ratio

Lemma (Kawaguchi, Kyan 1986; Schwiegelshohn 2011)

On any number of machines the worst-case approximation ratio is attained when $w_j = p_j$ for all j .

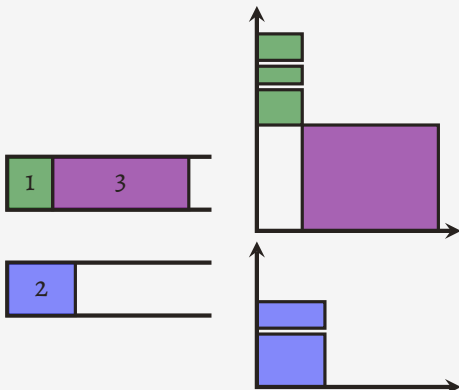


Reduction to Unit Smith Ratio

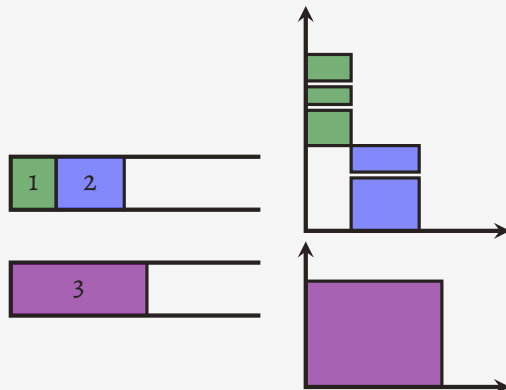
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WSPT schedule



OPT schedule

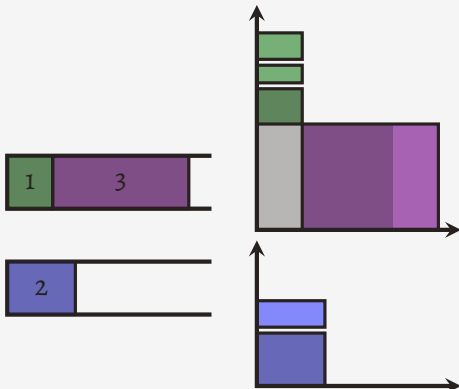


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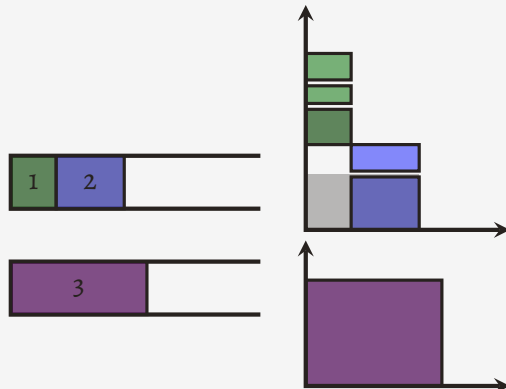
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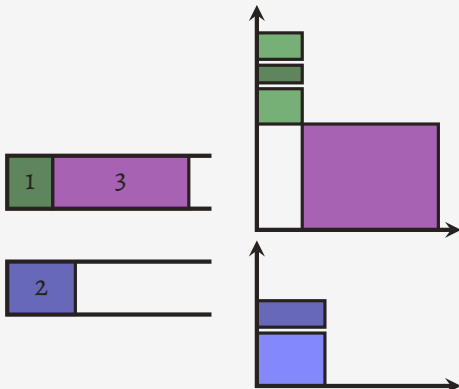


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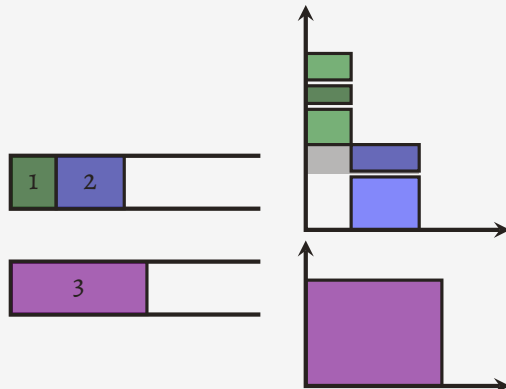
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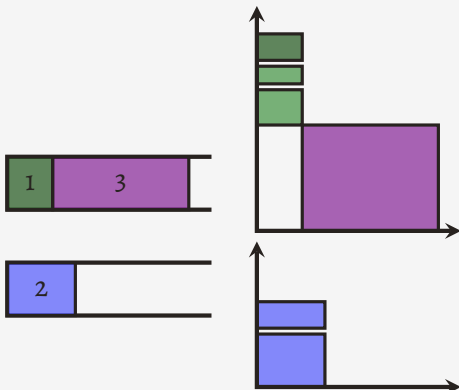


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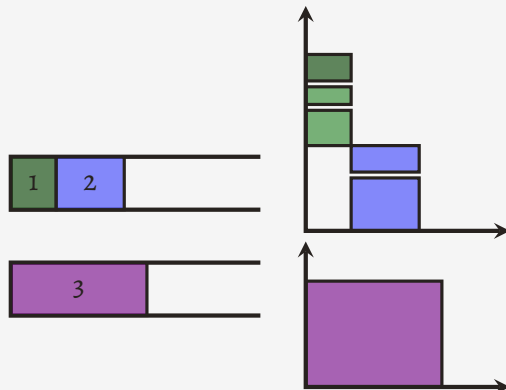
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WSPT schedule



OPT schedule



Approximation of the WSPT Rule for Sum of Weighted α -Points

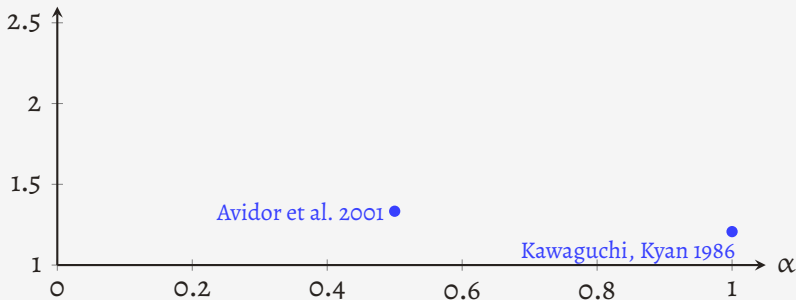
For $\alpha \in (0, 1]$ the α -point $C_j(\alpha)$ of job j is the first moment when an α -fraction of j has been finished.

- $C_j(1) = C_j$
- $C_j(\frac{1}{2}) = M_j$

Approximation of the WSPT Rule for Sum of Weighted α -Points

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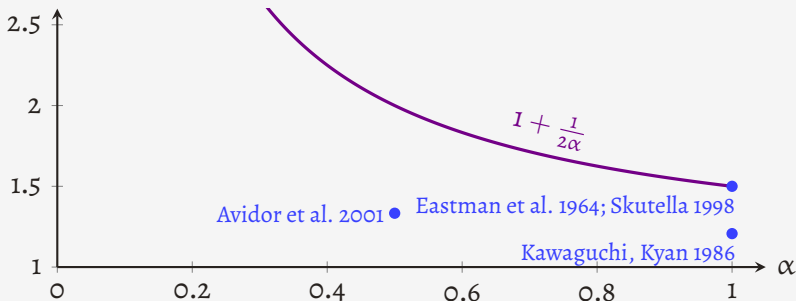
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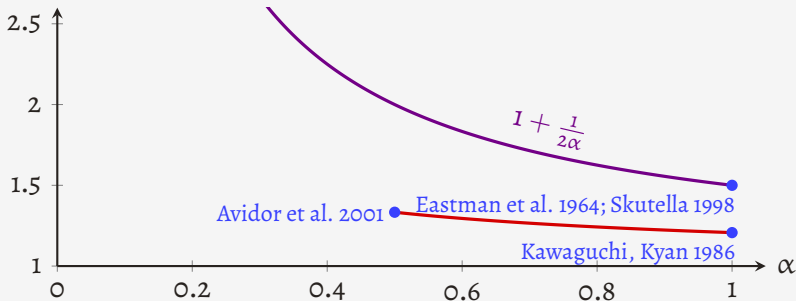
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Theorem

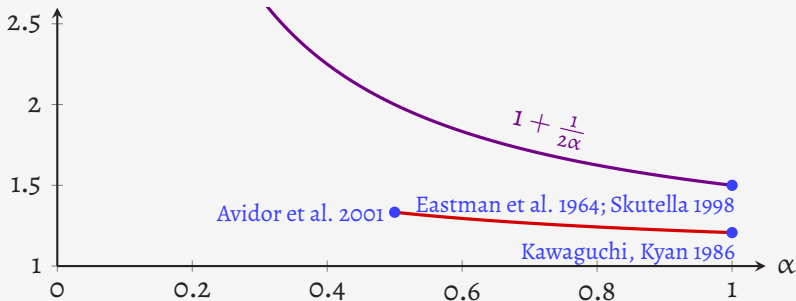
For $\alpha \in [\frac{1}{2}, 1]$ the WSPT rule has approximation ratio

$$1 + \frac{1}{2\alpha + \sqrt{8\alpha}}$$

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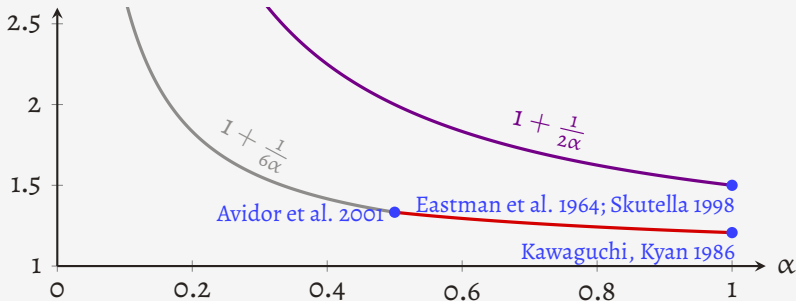
$$1 + \frac{1}{2\alpha + \sqrt{8\alpha}}$$

- Also computed worst case for $\alpha \in (\frac{1}{2}, 1]$ on $m \geq 2$ machines.

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$$1 + \frac{1}{2\alpha + \sqrt{8\alpha}}$$

- Also computed worst case for $\alpha \in (\frac{1}{2}, 1]$ on $m \geq 2$ machines.

Open Problem: Online List Assignment Model

- jobs arrive one by one; must be assigned to machines immediately
- on each machine, assigned jobs are optimally sequenced (WSPT)

Min-increase algorithm

Assign each job to a machine minimizing the increase of the current objective value.

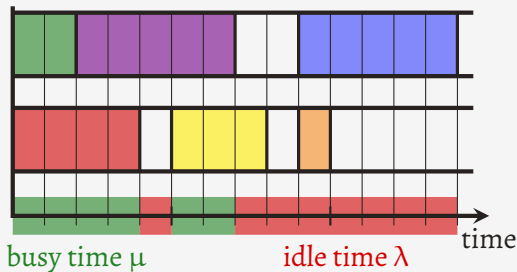
Known results:

- $(\frac{3}{2} - \frac{1}{2m})$ -competitive (cp. e.g. Megow et al. 2006)
- If jobs arrive in order of non-increasing or non-decreasing $\frac{w_j}{p_j}$, then Min-increase achieves competitive ratio $\frac{1+\sqrt{2}}{2}$.

Conjecture (Stougie 2017)

Min-increase has competitive ratio $\frac{1+\sqrt{2}}{2}$.

Scheduling with Precedence Delays on Identical Machines



Consider partial schedule when the blue job has been assigned.

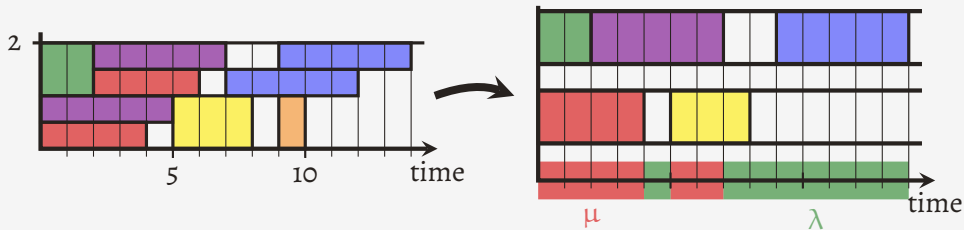
$$C_{\text{blue}} = \mu + \lambda.$$

For $\alpha \in (0, \frac{1}{2}]$ it holds that

$$\mu \leq \frac{1}{\alpha} \cdot C_{\text{blue}}^{\text{LP}}, \quad \lambda \leq \frac{1}{1 - \alpha} \cdot C_{\text{blue}}^{\text{LP}}.$$

For $\alpha = \frac{1}{2}$ this gives a 4-approximation algorithm.

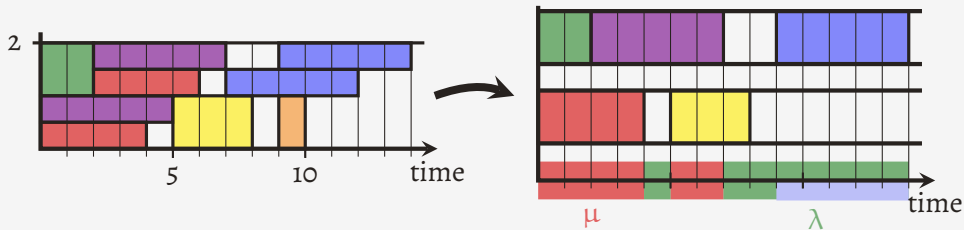
Bounding the Idle Time



Trace back why the **blue** job is not completed earlier.

$$\lambda \leq \dots$$

Bounding the Idle Time

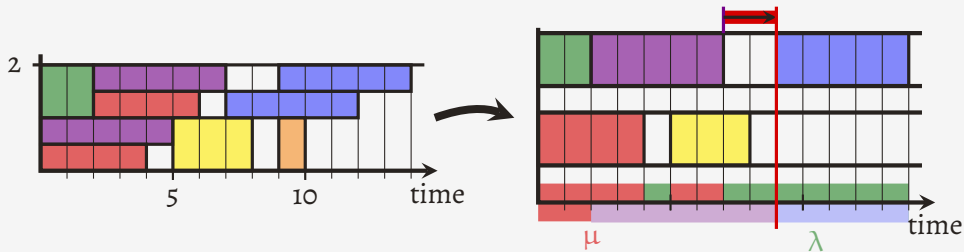


Trace back why the **blue** job is not completed earlier.

- The **blue** job has to be processed (can be idle time).

$$\lambda \leq p_{\text{blue}} + \dots$$

Bounding the Idle Time

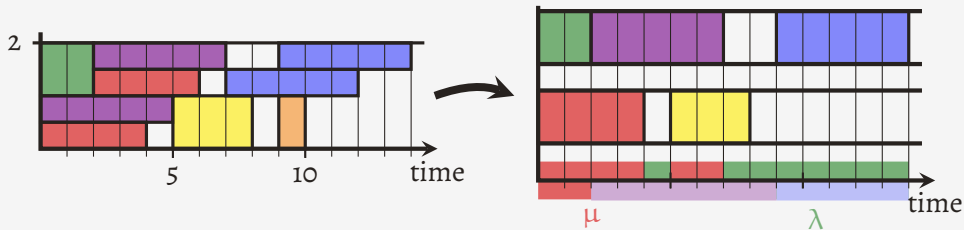


Trace back why the **blue** job is not completed earlier.

- The **blue** job has to be processed (can be idle time).
- The **blue** job has to wait for the **purple** job (can be idle time).

$$\lambda \leq p_{\text{blue}} + d_{\text{purple,blue}} + p_{\text{purple}}$$

Bounding the Idle Time

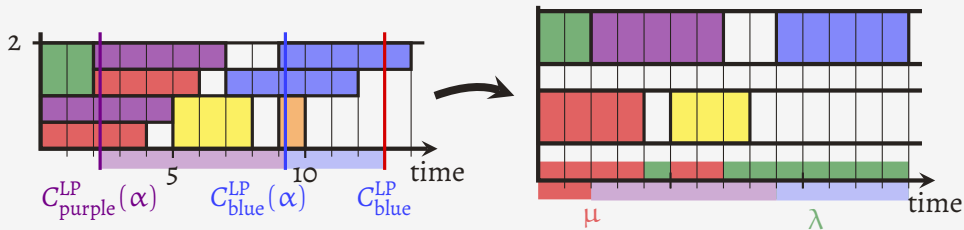


Trace back why the **blue** job is not completed earlier.

- The **blue** job has to be processed (can be idle time).
- The **blue** job has to wait for the **purple** job (can be idle time).
- The **purple** job has to wait because before all machines are busy (cannot be idle time).

$$\lambda \leq p_{\text{blue}} + d_{\text{purple,blue}} + p_{\text{purple}}$$

Bounding the Idle Time

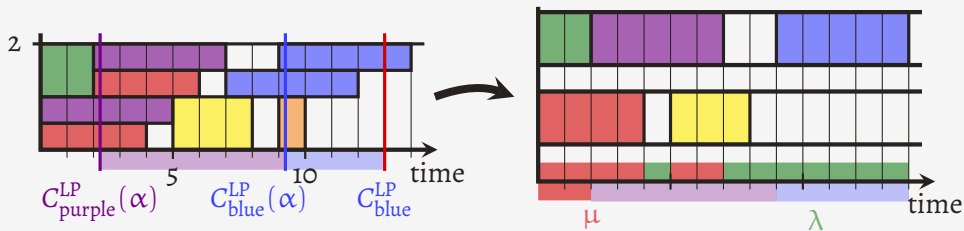


Trace back why the **blue** job is not completed earlier.

- The **blue** job has to be processed (can be idle time).
- The **blue** job has to wait for the **purple** job (can be idle time).
- The **purple** job has to wait because before all machines are busy (cannot be idle time).

$$\lambda \leq \underbrace{p_{\text{blue}}}_{= \frac{1}{1-\alpha} (C_{\text{blue}}^{\text{LP}}(\alpha) - C_{\text{blue}}^{\text{LP}}(\alpha))} + \underbrace{d_{\text{purple,blue}} + p_{\text{purple}}}_{\leq \frac{1}{1-\alpha} (C_{\text{blue}}^{\text{LP}}(\alpha) - C_{\text{purple}}^{\text{LP}}(\alpha))}$$

Bounding the Idle Time



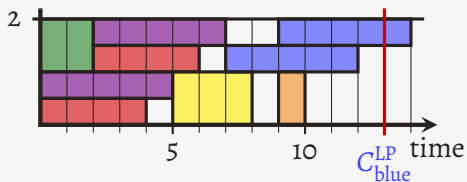
Trace back why the **blue** job is not completed earlier.

- The **blue** job has to be processed (can be idle time).
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$$\lambda \leq \underbrace{p_{\text{blue}}}_{= \frac{1}{1-\alpha} (C_{\text{blue}}^{\text{LP}} - C_{\text{blue}}^{\text{LP}}(\alpha))} + \underbrace{d_{\text{purple,blue}} + p_{\text{purple}}}_{\leq \frac{1}{1-\alpha} (C_{\text{blue}}^{\text{LP}}(\alpha) - C_{\text{purple}}^{\text{LP}}(\alpha))} \leq \frac{1}{1-\alpha} \cdot C_{\text{blue}}^{\text{LP}}.$$

Bounding the Expected Busy Time

Overview

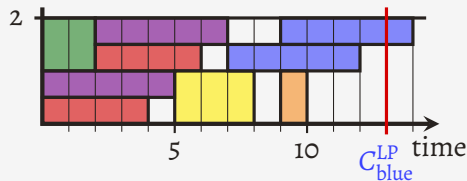


For $\alpha \in (0, \frac{1}{2}]$ let $J_\alpha := \{j \in N \mid C_j^{LP}(\alpha) \leq C_{blue}^{LP}\}$.

Upper bound on busy time: $\mu \leq \frac{p(J_\alpha)}{m}$.

Bounding the Expected Busy Time

Overview



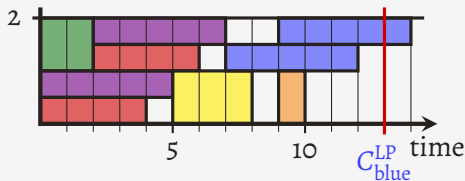
For $\alpha \in (0, \frac{1}{2}]$ let $J_\alpha := \{j \in N \mid C_j^{LP}(\alpha) \leq C_{blue}^{LP}\}$.

Upper bound on busy time: $\mu \leq \frac{p(J_\alpha)}{m}$.

Lower bound on LP-completion time: $C_{blue}^{LP} \geq \frac{E[p(J_\alpha)]}{2m}$.

Bounding the Expected Busy Time

Overview



For $\alpha \in (0, \frac{1}{2}]$ let $J_\alpha := \{j \in N \mid C_j^{LP}(\alpha) \leq C_{blue}^{LP}\}$.

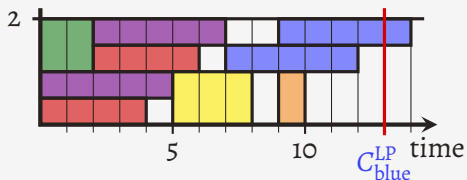
Upper bound on busy time: $\mu \leq \frac{p(J_\alpha)}{m}$.

Lower bound on LP-completion time: $C_{blue}^{LP} \geq \frac{E[p(J_\alpha)]}{2m}$.

$$E[\mu] \leq 2 \cdot C_{blue}^{LP}.$$

Bounding the Expected Busy Time

Lower bound on LP-completion time

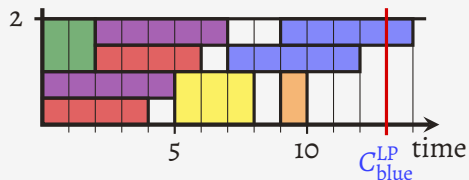


For $\alpha \in (0, \frac{1}{2}]$ let $J_\alpha := \{j \in N \mid C_j^{\text{LP}}(\alpha) \leq C_{\text{blue}}^{\text{LP}}\}$.

Claim: $C_{\text{blue}}^{\text{LP}} \geq \frac{1}{2} \cdot \frac{E[p(J_\alpha)]}{m}$.

Bounding the Expected Busy Time

Lower bound on LP-completion time



For $\alpha \in (0, \frac{1}{2}]$ let $J_\alpha := \{j \in N \mid C_j^{\text{LP}}(\alpha) \leq C_{\text{blue}}^{\text{LP}}\}$.

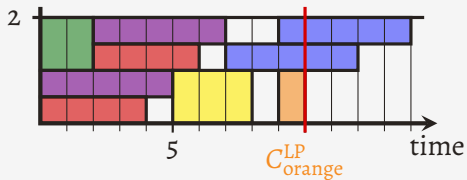
Claim: $C_{\text{blue}}^{\text{LP}} \geq \frac{1}{2} \cdot \frac{E[p(J_\alpha)]}{m}$.

In the example $C_j^{\text{LP}}(\frac{1}{2}) \leq C_{\text{blue}}^{\text{LP}}$ for all $j \in N$, whence $J_\alpha = N$ for all α . Therefore,

$$C_{\text{blue}}^{\text{LP}} \geq \frac{1}{2} \cdot \frac{p(N)}{m} = \frac{1}{2} \cdot \frac{E[p(J_\alpha)]}{m}.$$

Bounding the Expected Busy Time

Lower bound on LP-completion time

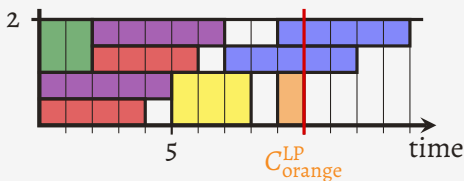


For $\alpha \in (0, \frac{1}{2}]$ let $J_\alpha := \{j \in N \mid C_j^{LP}(\alpha) \leq C_{orange}^{LP}\}$.

Claim: $C_{orange}^{LP} \geq \frac{1}{2} \cdot \frac{E[p(J_\alpha)]}{m}$.

Bounding the Expected Busy Time

Lower bound on LP-completion time



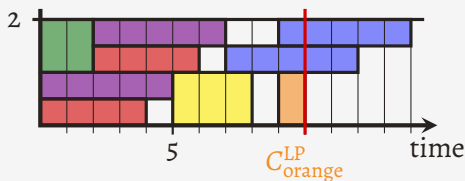
For $\alpha \in (0, \frac{1}{2}]$ let $J_\alpha := \{j \in N \mid C_j^{LP}(\alpha) \leq C_{orange}^{LP}\}$.

Claim: $C_{orange}^{LP} \geq \frac{1}{2} \cdot \frac{E[p(J_\alpha)]}{m}$.

$$p(J_\alpha) = \begin{cases} p_{green} + p_{purple} + p_{red} + p_{yellow} + p_{orange} + p_{blue} & \text{if } \alpha \leq \frac{2}{5}; \\ p_{green} + p_{purple} + p_{red} + p_{yellow} + p_{orange} & \text{if } \alpha > \frac{2}{5}. \end{cases}$$

Bounding the Expected Busy Time

Lower bound on LP-completion time



For $\alpha \in (0, \frac{1}{2}]$ let $J_\alpha := \{j \in N \mid C_j^{LP}(\alpha) \leq C_{orange}^{LP}\}$.

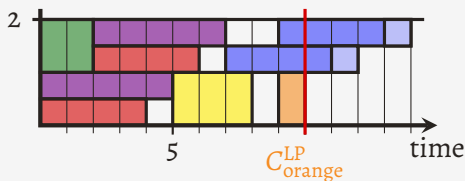
Claim: $C_{orange}^{LP} \geq \frac{1}{2} \cdot \frac{E[p(J_\alpha)]}{m}$.

$$p(J_\alpha) = \begin{cases} p_{green} + p_{purple} + p_{red} + p_{yellow} + p_{orange} + p_{blue} & \text{if } \alpha \leq \frac{2}{5}; \\ p_{green} + p_{purple} + p_{red} + p_{yellow} + p_{orange} & \text{if } \alpha > \frac{2}{5}. \end{cases}$$

$$E[p(J_\alpha)] = p_{green} + p_{purple} + p_{red} + p_{yellow} + p_{orange} + \frac{4}{5} \cdot p_{blue}$$

Bounding the Expected Busy Time

Lower bound on LP-completion time



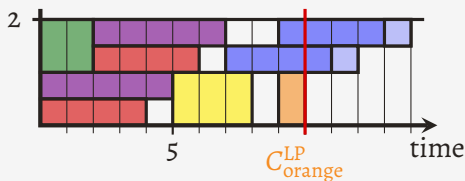
For $\alpha \in (0, \frac{1}{2}]$ let $J_\alpha := \{j \in N \mid C_j^{LP}(\alpha) \leq C_{orange}^{LP}\}$.

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Bounding the Expected Busy Time

Lower bound on LP-completion time



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$$E[p(J_\alpha)] = p_{green} + p_{purple} + p_{red} + p_{yellow} + p_{orange} + \frac{4}{5} \cdot p_{blue}$$

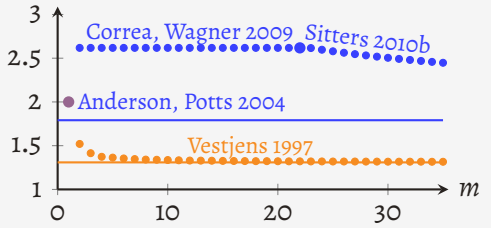
The modified jobs J' have midpoints bounded by C_{orange}^{LP} . Therefore,

$$C_{orange}^{LP} \geq \frac{1}{2} \cdot \frac{p(J')}{m} = \frac{1}{2} \cdot \frac{E[p(J_\alpha)]}{m}.$$

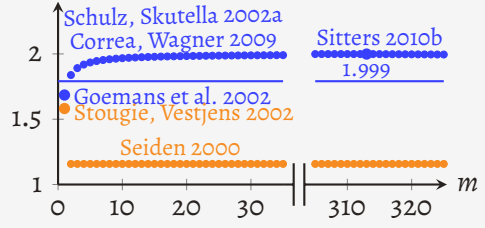
Clairvoyant Online Time Model on Identical Machines

Non-preemptive scheduling

Deterministic algorithms

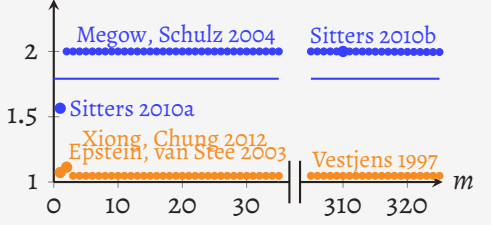


Randomized algorithms

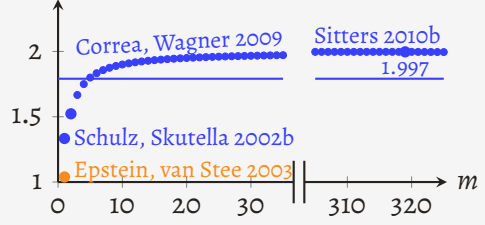


Preemptive scheduling

Deterministic algorithms



Randomized algorithms



Non-Migratory Preemptive Scheduling on Unrelated Machines

Assume that all $p_{ij}, r_j \in 2\mathbb{Z}$, and let $T \in \mathbb{Z}$ be an upper bound on the makespan.

Variables: y_{ijt} , $i \in [m]$, $j \in [n]$, $t \in \{r_j, \dots, T-1\}$, indicating how long job j is processed on machine i during time slot $(t, t+1]$.

$$\begin{aligned}
 \text{(LP) min} \quad & \sum_{j=1}^n w_j \cdot \sum_{i=1}^m \sum_{t=r_j}^{T-1} \left(\frac{y_{ijt}}{2} + \frac{y_{ijt}}{p_{ij}} \cdot \left(t + \frac{1}{2} \right) \right) \\
 \text{s. t.} \quad & \sum_{i=1}^m \sum_{t=r_j}^{T-1} \frac{y_{ijt}}{p_{ij}} = 1 \quad \forall j \in [n] \\
 & \sum_{j:r_j \geq t} y_{ijt} \leq 1 \quad \forall i \in [m], t \in \{0, \dots, T-1\} \\
 & y_{ijt} \geq 0 \quad \forall i \in [m], j \in [n], t \in \{r_j, \dots, T-1\}
 \end{aligned}$$

Dual LP

$$(LP) \min \sum_{j=1}^n w_j \cdot \sum_{i=1}^m \sum_{t=r_j}^{T-1} \left(\frac{y_{ijt}}{2} + \frac{y_{ijt}}{p_{ij}} \cdot \left(t + \frac{1}{2} \right) \right)$$

$$\text{s. t. } \sum_{i=1}^m \sum_{t=r_j}^{T-1} \frac{y_{ijt}}{p_{ij}} = 1 \quad \forall j \in [n]$$

$$\sum_{j:r_j \geq t} y_{ijt} \leq 1 \quad \forall i \in [m], t \in \{0, \dots, T-1\}$$

$$y_{ijt} \geq 0 \quad \forall i \in [m], j \in [n], t \in \{r_j, \dots, T-1\}$$

$$(D) \max \sum_{j=1}^n \chi_j - \sum_{i=1}^m \sum_{t=0}^{T-1} \psi_{it}$$

$$\text{s. t. } \frac{\chi_j}{p_{ij}} \leq \psi_{it} + w_j \left(\frac{t + 1/2}{p_{ij}} + \frac{1}{2} \right) \quad \forall i \in [m], j \in [n], t \in \{r_j, \dots, T-1\}$$

$$\psi_{it} \geq 0 \quad \forall i \in [m], t \in \{0, \dots, T-1\}$$

Dual Solution

- For $j \in [n]$ let $i(j)$ be the machine to which j is assigned by the Greedy-Assignment WSRPT algorithm.
- For $i \in [m]$ and $t \in \mathbb{Z}_{\geq 0}$ let $U_{i,t}$ be the set of jobs completed after time t on machine i in the Greedy-Assignment WSRPT schedule.

$$\chi_j := \frac{1}{2} \cdot \text{cost}(j \rightarrow i(j)) \quad \text{for all } j \in [n],$$

$$\psi_{it} := \frac{w(U_{i,2t})}{2} \quad \text{for all } i \in [m], t \in \{0, \dots, T-1\},$$

Cost of Dual Solution

$$\begin{aligned}
 \text{(D) max} \quad & \sum_{j=1}^n \chi_j && - \sum_{i=1}^m \sum_{t=0}^{T-1} \psi_{it} \\
 \text{s. t.} \quad & \frac{\chi_j}{p_{ij}} \leq \psi_{it} + w_j \cdot \left(\frac{t + 1/2}{p_{ij}} + \frac{1}{2} \right) && \forall i, j, t \\
 & \psi_{it} \geq 0 && \forall i, t
 \end{aligned}$$

Cost of Dual Solution

$$\begin{aligned}
 \text{(D) max} \quad & \sum_{j=1}^n \frac{\text{cost}(j \rightarrow i(j))}{2} - \sum_{i=1}^m \sum_{t=0}^{T-1} \frac{w(U_{i,2t})}{2} \\
 \text{s. t.} \quad & \frac{X_j}{p_{ij}} \leq \psi_{it} + w_j \cdot \left(\frac{t + 1/2}{p_{ij}} + \frac{1}{2} \right) \quad \forall i, j, t \\
 & \psi_{it} \geq 0 \quad \forall i, t
 \end{aligned}$$

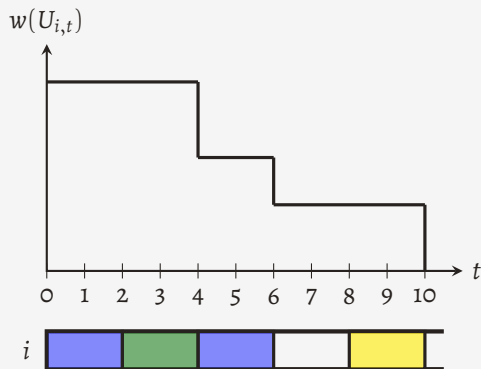
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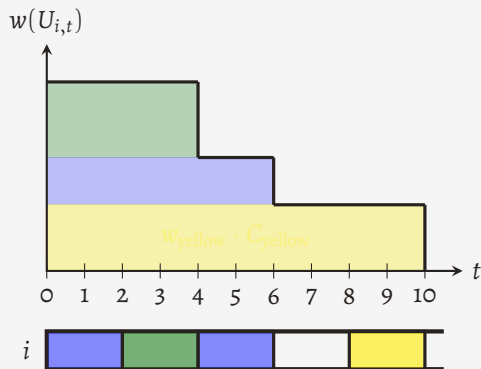
$$= \frac{1}{2} \cdot Z^{\text{GA-WSRPT}} - \frac{1}{4} \cdot Z^{\text{GA-WSRPT}} = \frac{1}{4} \cdot Z^{\text{GA-WSRPT}}$$



Cost of Dual Solution

$$\sum_{j=1}^n \frac{\text{cost}(j \rightarrow i(j))}{2} - \sum_{i=1}^m \sum_{t=0}^{T-1} \frac{w(U_{i,2t})}{2}$$

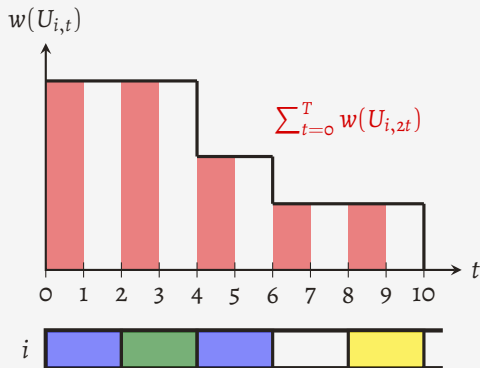
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Cost of Dual Solution

$$\sum_{j=1}^n \frac{\text{cost}(j \rightarrow i(j))}{2} - \sum_{i=1}^m \sum_{t=0}^{T-1} \frac{w(U_{i,2t})}{2}$$

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Feasibility of Dual Solution

$$\begin{aligned}
 \text{(D) max} \quad & \sum_{j=1}^n \chi_j - \sum_{i=1}^m \sum_{t=0}^{T-1} \psi_{it} \\
 \text{s. t.} \quad & \frac{\chi_j}{p_{ij}} \leq \psi_{it} + w_j \cdot \left(\frac{t+1/2}{p_{ij}} + \frac{1}{2} \right) \quad \forall i, j, t \\
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 \text{s. t.} \quad & \frac{\text{cost}(j \rightarrow i(j))}{2 \cdot p_{ij}} \leq \frac{w(U_{i,2t})}{2} + w_j \cdot \left(\frac{t + 1/2}{p_{ij}} + \frac{1}{2} \right) \quad \forall i, j, t \\
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Feasibility of Dual Solution

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Weighted Shortest Expected Processing Time First Rule

WSEPT rule

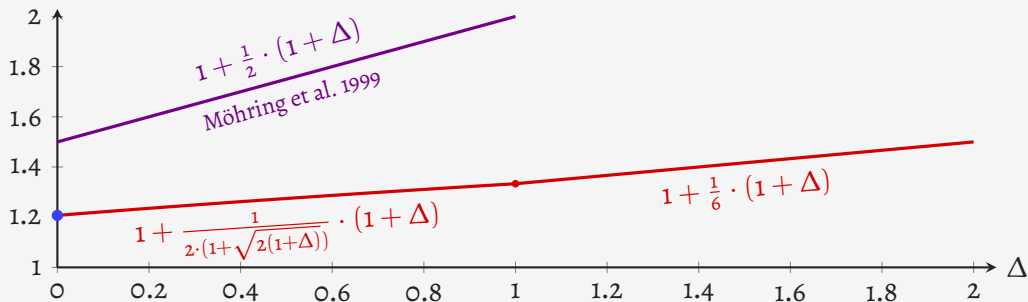
Whenever a machine is free, start available job with maximum ratio $w_j/E[p_j]$ on it.

- WSEPT is optimal if
 - there is only one machine (Rothkopf 1966)
 - all jobs have unit weight, and the processing times are pairwise stochastically comparable (Weber et al. 1986)
- Even for unit-weight jobs WSEPT has no constant performance guarantee. (Cheung et al. 2014; Im et al. 2015)
- The approximation ratio can be bounded in terms of

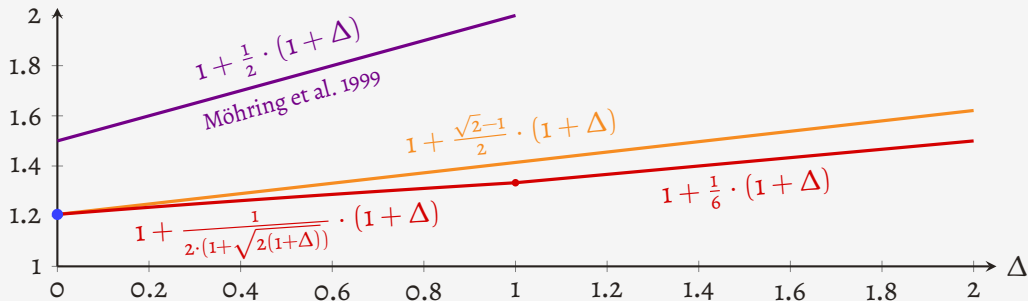
$$\Delta := \max_{j \in \{1, \dots, n\}} \frac{\text{Var}[p_j]}{E[p_j]^2}.$$

(Möhring et al. 1999)

Performance Guarantees for the WSEPT Rule



Performance Guarantees for the WSEPT Rule



Proof of the Performance Guarantee

Idea: Consider random weights $\mathbf{w}'_j := \frac{P_j}{E[P_j]} \cdot w_j$.

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\implies in every realization the WSEPT rule schedules the jobs in non-increasing order of $\frac{w'_j}{p_j}$.

$$\sum_{j=1}^n \mathbf{w}'_j \cdot \mathbf{C}_j^{\text{WSEPT}} \leq \frac{1 + \sqrt{2}}{2} \cdot \sum_{j=1}^n \mathbf{w}'_j \cdot \mathbf{C}_j^{\text{OPT}(\mathbf{p}, \mathbf{w}') } \leq \frac{1 + \sqrt{2}}{2} \cdot \sum_{j=1}^n \mathbf{w}'_j \cdot \mathbf{C}_j^{\text{OPT}}.$$

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Lemma

For any stochastic scheduling policy Π it holds that

$$E \left[\sum_{j=1}^n \mathbf{w}'_j \cdot \mathbf{C}_j^{\Pi} \right] = E \left[\sum_{j=1}^n w_j \cdot \mathbf{C}_j^{\Pi} \right] + \underbrace{\sum_{j=1}^n w_j \cdot \frac{\text{Var}[p_j]}{E[p_j]}}_{=c}.$$

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$$E \left[\sum_{j=1}^n \mathbf{w}'_j \cdot \mathbf{C}_j^{\Pi} \right] = E \left[\sum_{j=1}^n w_j \cdot \mathbf{C}_j^{\Pi} \right] + \underbrace{\sum_{j=1}^n w_j \cdot \frac{\text{Var}[p_j]}{E[p_j]}}_{=c}.$$

Proof

$$E[\mathbf{p}_j \mathbf{C}_j^{\Pi}] = E[\mathbf{p}_j \mathbf{S}_j] + E[\mathbf{p}_j^2] = E[\mathbf{p}_j] \cdot (E[\mathbf{S}_j] + E[\mathbf{p}_j]) + \text{Var}[\mathbf{p}_j] = E[\mathbf{p}_j] \cdot E[\mathbf{C}_j] + \text{Var}[\mathbf{p}_j]. \quad \square$$

Proof of the Performance Guarantee

Idea: Consider random weights $\mathbf{w}'_j := \frac{p_j}{E[p_j]} \cdot w_j$.

\implies in every realization the WSEPT rule schedules the jobs in non-increasing order of $\frac{w'_j}{p_j}$.

$$E \left[\sum_{j=1}^n \mathbf{w}'_j \cdot \mathbf{C}_j^{\text{WSEPT}} \right] \leq \frac{1 + \sqrt{2}}{2} \cdot E \left[\sum_{j=1}^n \mathbf{w}'_j \cdot \mathbf{C}_j^{\text{OPT}(\mathbf{p}, \mathbf{w}')} \right] \leq \frac{1 + \sqrt{2}}{2} \cdot E \left[\sum_{j=1}^n \mathbf{w}'_j \cdot \mathbf{C}_j^{\text{OPT}} \right].$$

Lemma

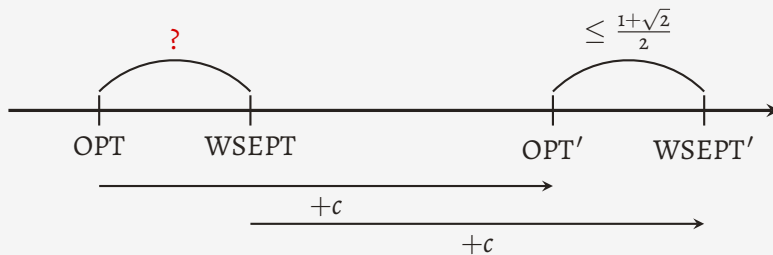
For any stochastic scheduling policy Π it holds that

$$E \left[\sum_{j=1}^n \mathbf{w}'_j \cdot \mathbf{C}_j^{\Pi} \right] = E \left[\sum_{j=1}^n w_j \cdot \mathbf{C}_j^{\Pi} \right] + \underbrace{\sum_{j=1}^n w_j \cdot \frac{\text{Var}[p_j]}{E[p_j]}}_{=: c \leq \sum_j w_j \cdot \Delta \cdot E[p_j] \leq \Delta \text{OPT}}$$

Proof

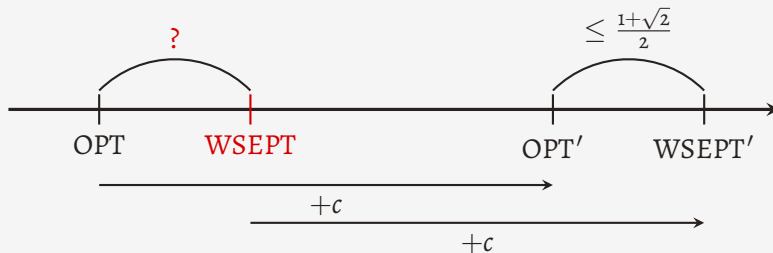
$$E[\mathbf{p}_j \mathbf{C}_j^{\Pi}] = E[\mathbf{p}_j \mathbf{S}_j] + E[\mathbf{p}_j^2] = E[\mathbf{p}_j] \cdot (E[\mathbf{S}_j] + E[\mathbf{p}_j]) + \text{Var}[\mathbf{p}_j] = E[\mathbf{p}_j] \cdot E[\mathbf{C}_j] + \text{Var}[\mathbf{p}_j]. \quad \square$$

Proof of the Performance Guarantee (Cont.)



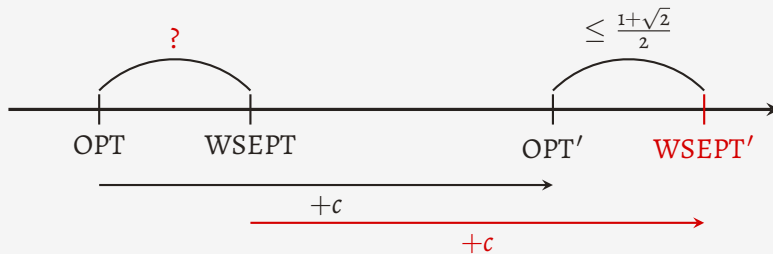
WSEPT

Proof of the Performance Guarantee (Cont.)



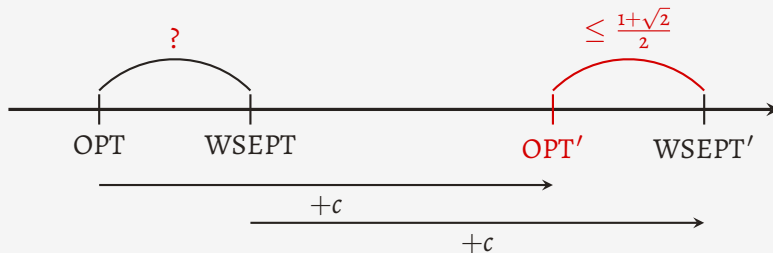
WSEPT

Proof of the Performance Guarantee (Cont.)



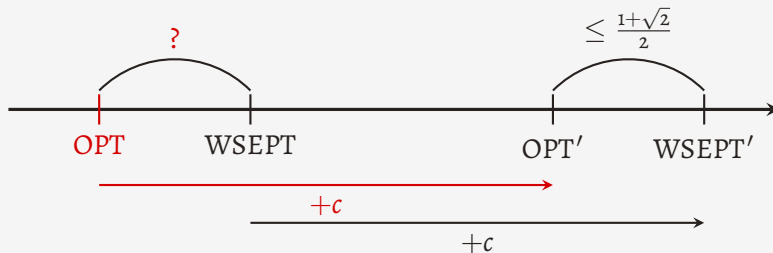
$$\text{WSEPT} = \text{WSEPT}' - c$$

Proof of the Performance Guarantee (Cont.)



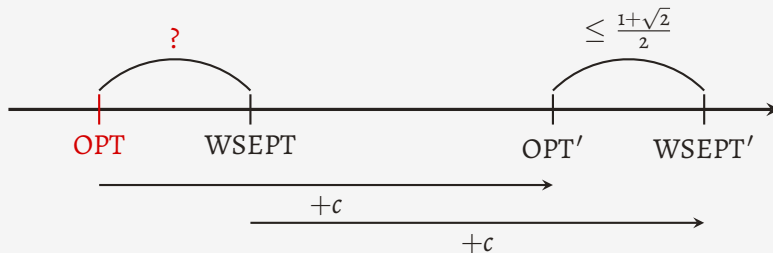
$$\text{WSEPT} = \text{WSEPT}' - c \leq \frac{1 + \sqrt{2}}{2} \cdot \text{OPT}' - c$$

Proof of the Performance Guarantee (Cont.)



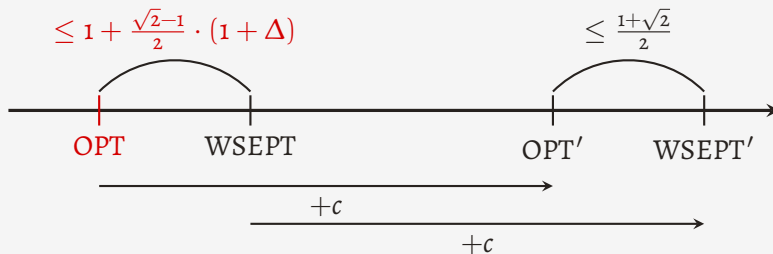
$$\text{WSEPT} = \text{WSEPT}' - c \leq \frac{1 + \sqrt{2}}{2} \cdot \text{OPT}' - c = \frac{1 + \sqrt{2}}{2} \cdot (\text{OPT} + c) - c$$

Proof of the Performance Guarantee (Cont.)



$$\begin{aligned}
 \text{WSEPT} &= \text{WSEPT}' - c \leq \frac{1 + \sqrt{2}}{2} \cdot \text{OPT}' - c = \frac{1 + \sqrt{2}}{2} \cdot (\text{OPT} + c) - c \\
 &= \text{OPT} + \frac{\sqrt{2} - 1}{2} \cdot (\text{OPT} + c)
 \end{aligned}$$

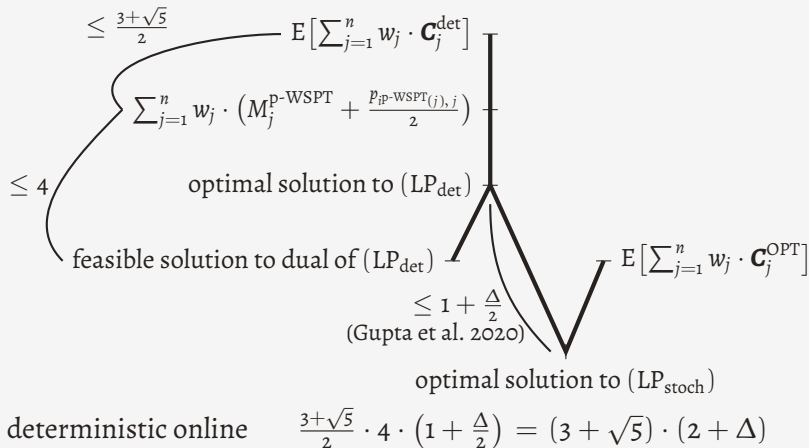
Proof of the Performance Guarantee (Cont.)



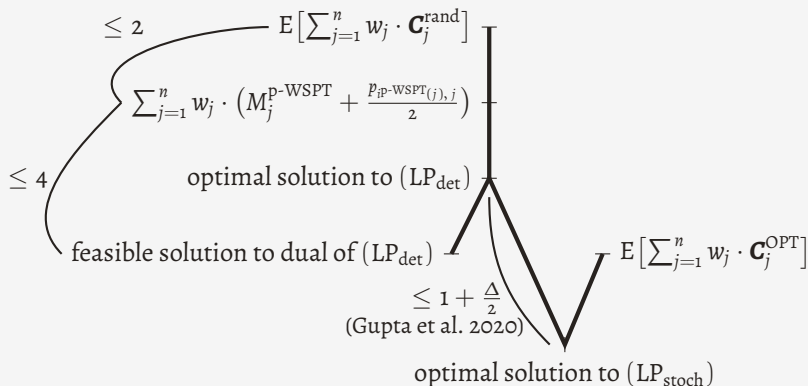
$$\begin{aligned}
 \text{WSEPT} &= \text{WSEPT}' - c \leq \frac{1 + \sqrt{2}}{2} \cdot \text{OPT}' - c = \frac{1 + \sqrt{2}}{2} \cdot (\text{OPT} + c) - c \\
 &= \text{OPT} + \frac{\sqrt{2} - 1}{2} \cdot (\text{OPT} + c) \leq \left(1 + \frac{\sqrt{2} - 1}{2} \cdot (1 + \Delta)\right) \cdot \text{OPT}
 \end{aligned}$$

$c \leq \Delta \text{OPT}$

Analysis of Online Scheduling Policies for Unrelated Machines



Analysis of Online Scheduling Policies for Unrelated Machines



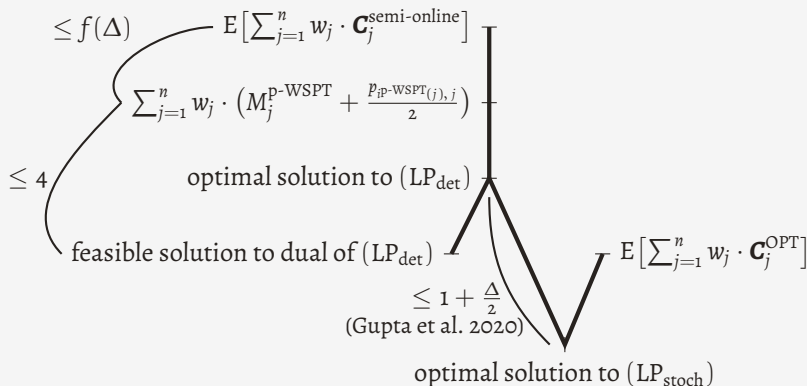
deterministic online

$$\frac{3+\sqrt{5}}{2} \cdot 4 \cdot (1 + \frac{\Delta}{2}) = (3 + \sqrt{5}) \cdot (2 + \Delta)$$

randomized online

$$2 \cdot 4 \cdot (1 + \frac{\Delta}{2}) = 4 \cdot (2 + \Delta)$$

Analysis of Online Scheduling Policies for Unrelated Machines

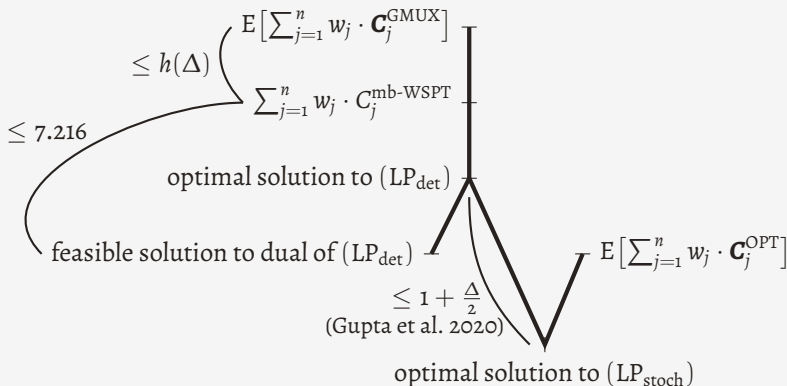


deterministic online $\frac{3+\sqrt{5}}{2} \cdot 4 \cdot (1 + \frac{\Delta}{2}) = (3 + \sqrt{5}) \cdot (2 + \Delta)$

randomized online $2 \cdot 4 \cdot (1 + \frac{\Delta}{2}) = 4 \cdot (2 + \Delta)$

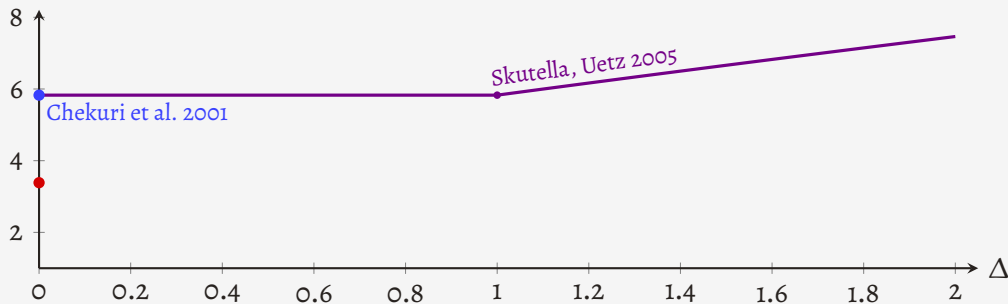
semi-online $f(\Delta) \cdot 4 \cdot (1 + \frac{\Delta}{2}) = f(\Delta) \cdot 2 \cdot (2 + \Delta)$









Analysis of Online Scheduling Policies for Unrelated Machines

















deterministic online	$\frac{3+\sqrt{5}}{2} \cdot 4 \cdot (1 + \frac{\Delta}{2}) = (3 + \sqrt{5}) \cdot (2 + \Delta)$
randomized online	$2 \cdot 4 \cdot (1 + \frac{\Delta}{2}) = 4 \cdot (2 + \Delta)$
semi-online	$f(\Delta) \cdot 4 \cdot (1 + \frac{\Delta}{2}) = f(\Delta) \cdot 2 \cdot (2 + \Delta)$
Gupta et al. 2021+	$h(\Delta) \cdot 7.216 \cdot (1 + \frac{\Delta}{2}) = 3.608 \cdot h(\Delta) \cdot (2 + \Delta)$









Stochastic Scheduling with Precedence Constraints and Release Dates

















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