



Universal Scheduling on a single machine

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Scheduling

Given a set of jobs and a set of machines.

- Each job has a processing time.
- Each job has to be processed by one of the machines
- Each machine can process at most one job at a time
- A schedule is an assignment of the jobs to the machine together with an order of the jobs on the machines
- Given a schedule the time at which a job completes is called its completion time
- Objectives are functions of the completion times of jobs like maximum completion time or average completion times
- There may be deadlines, release times, precedence constraints etc.

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Given a set J of n jobs to be scheduled on 1 machine

- job j has processing time p_j and weight w_j
- Objective: Minimize the weighted sum of completion times

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We are interested in scheduling policies that are **robust** with respect to machine break- and slow-down



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Also called in the literature **universal scheduling**

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Deterministic	On-line/Off-line	Robust scheduling
Stochastic	2-stage optimization with just 2nd stage (probabilistic analysis)	2-stage optimization decision made in stage 1 (a priori scheduling)

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Specifically, the job being processed at the moment of a machine break down can be resumed without loss of processing as soon as the machine becomes available again

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- in P if all $w_j = 1$
- Without preemption two or more unavailability periods is hard even if all $w_j = 1$

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Price of Robustness

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Price of Robustness is $10/9$

Machine capacity function

The machine capacity function $f(t)$ is the aggregated amount of processing time available up to time t

Robustness against any machine capacity function

Results

- A deterministic polynomial time algorithm that provides a sequence that for any $\epsilon > 0$ and any machine capacity function is at most $4 + \epsilon$ times the optimum
- e -approximate randomized algorithm
- lower bound of 4 for deterministic algorithms
- lower bound of e for randomized algorithms
- Extension to (restricted) precedence constraints
- **FPTAS** for 1 unavailability problem (non-robust)

Deterministic $4 + \epsilon$

Assume maximum speed 1. Let π be job-order and $S(\pi)$ a resulting schedule

- $C_j^{S(\pi)} = \min\{t \mid f(t) \geq C_j^\pi\}$ completion time of job j
- $W^{S(\pi)}(t)$ total weight of jobs not completed by time t

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- With $f(t)$ the machine capacity function

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- KEY LEMMA. *A sequence π of jobs yields a schedule with objective value at most c times the optimum value for all machine capacity functions f if and only if for all t*

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“if” is clear, since

$$\sum_{j \in J} w_j C_j^{S(\pi)} = \int_0^\infty W^\pi(f(t)) dt \leq c \int_0^\infty W^*(f(t)) dt$$

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$$f(t) = \begin{cases} t & \text{if } t \leq t_0 \\ t_0 & \text{if } t_0 < t \leq t_1 \\ t - (t_1 - t_0) & \text{if } t > t_1 \end{cases}$$

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Since by assumption $W^\pi(t_0)/W^*(t_0) > c$ making $t_1 \rightarrow \infty$ causes $\sum_{j \in J} w_j C_j^{S(\pi)} / \sum_{j \in J} w_j C_j^{S(\pi^*)} > c$. Contradiction

Deterministic $4 + \epsilon$

Given our Key Lemma, we can use a result of Bechetti, Leonardi, Marchetti Spaccamela and Pruhs, 2003 who design an algorithm for which they prove $W^\pi(t) \leq 24W^*(t)$ for all t showing that a constant ratio (price of robustness) exists.

We will design an algorithm with $c = 4$

Deterministic $4 + \epsilon$

Idea: put jobs with small w and large p later

Algorithm A.

- Given a set J of jobs construct the sequence backwards
- Iteration $i, i = 0, 1, 2, \dots$: Let J_i be the set of jobs with total weight $\leq 2^i$ and maximum total processing time
- Schedule jobs in

$$J_i \setminus \bigcup_{j < i} J_j$$

in any order at the end, just before all jobs in J_{i-1}

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Note: Algorithm A is not polynomial time

Theorem: A is robust 4-approximate

Proof: by the Key Lemma it is sufficient to show that $W^\pi(t) \leq 4W^*(t)$ for all t .

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$$(1) \quad W^\pi(t) \leq \sum_{i=0}^k w(J_i) \leq \sum_{i=0}^k 2^i = 2^{k+1} - 1 \quad (1)$$

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If $k = 0$ the statement is clearly true

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(1) and (2) combine to the result

A polynomial time algorithm

- Algorithm A requires at each iteration to solve a knapsack problem
- Instead of using exact solution use a $(1 + \epsilon)$ approximation
- Hence we obtain a $(4 + \epsilon)$ approximation with running time polynomial in the number of jobs and $1/\epsilon$

A randomised algorithm

- Randomised Alg: similar to deterministic algorithm
- Weight of jobs selected at iteration i
 - Deterministic: weights sequence $2^0, 2^1, \dots$
 - Random: choose a random value y in $[0, 1]$ and then the sequence of weights is e^{i+y} , $i = 0, 1, \dots$
- **Theorem.** For every instance the randomized algorithm produces a random permutation ρ s.t.

$$E[W^\rho(t)] \leq eW^*(t), \forall t$$

Lower bound

Consider the following sequence of jobs:

$$p_k = k^k, w_k = k, k = M, M + 1, \dots, M^M$$

For any $k \geq M$ we have $\sum_{j < k} p_j < p_k$. Let $T = \sum_k p_k$

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Lemma. If the price of robustness (PoR) of a job order is smaller than α then any job $k > M\alpha$ must be succeeded (not necessarily directly) by some job r with $k/\alpha \leq r \leq k - 1$

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Given a schedule π with PoR: in backward direction starting from the last job in the schedule construct a maximal sequence of jobs with monotonically increasing weight, say length L . Let their weights starting with the last one be w^1, w^2, \dots, w^L . By Lemma for all k , $w^{k+1}/w^k \leq \alpha$ and (for $\alpha < M$) $L \geq M - 1$.

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Using w^k as the number of the job: At time $t_k = T - p(w^k + 1)$ in the schedule π a.o. jobs w^1, \dots, w^{k+1} are not completed. OPT has not completed job $w^k + 1$. PoR of π is at least

$$\alpha \geq \frac{\sum_{j=1}^{k+1} w^j}{w^k + 1}, \quad \forall k = 1, \dots, M - 1$$

The recurrence implies that $\alpha \geq 4$

QED

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However, Key Lemma does not hold:

$$r_i = i, p_i = w_i = 1, i = 1, \dots, n, r_{n+1} = n + 1, p_{n+1} = 1, w_{n+1} = M.$$

Clearly the order $\pi = 1, 2, \dots, n, n + 1$ satisfies the inequality.

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Breakdown at $[n, n + 1]$. According to π , job n is processed at time $n + 1$. Long breakdown starting at time $t \in [n + 2, n + 3)$. Job $n + 1$ uncompleted. In OPT job n is uncompleted, yielding ratio of $M/1 = M$.

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Theorem. The price of robustness in the presence of release dates is

$$\Omega\left(\frac{\log n}{\log \log n}\right)$$

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Clearly the order $\pi = 1, 2, \dots, n, n + 1$ satisfies the inequality.

Breakdown at $[n, n + 1]$. According to π , job n is processed at time $n + 1$. Long breakdown starting at time $t \in [n + 2, n + 3)$. Job $n + 1$ uncompleted. In OPT job n is uncompleted, yielding ratio of $M/1 = M$.

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$$\Omega\left(\frac{\log n}{\log \log n}\right)$$

(in fact through an adversarial instance with all unit weights $w_i = 1$)

Release dates

Giving the order as before, ignoring release dates, and at any time processing the highest priority job in that order that is available satisfies $W^\pi(t) \leq 4W^*(t)$, $\forall t$

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No upperbound so far

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Let j first job in ℓ . Then breakdown at $[r_j, r_0]$ and $[r_0 + 2^j - 1, L]$. At time r_0 all jobs are released. At time r_0 (end of first breakdown), job j starts processing and is not completed by $r_0 + 2^j - 1$. Hence all jobs of ℓ are uncompleted, whereas an optimal schedule can complete all jobs except j . Choose L large enough.

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For all $j \in \ell_i$ breakdown $[r_j, r_j + \epsilon]$ and for all $j \in \ell_{i+1}, \dots, \ell_k$ breakdown $[r_j, r_j + 2^j] = [r_j, r_j + p_j]$. At time $2^n - 1$ all jobs in $\ell_i, \ell_{i+1}, \dots, \ell_k$ are uncompleted. Compare with a schedule that leaves only the last job of ℓ_i and all jobs in $\ell_{i+1}, \dots, \ell_k$ uncompleted. Thus a breakdown during $[2^n - 1, L]$ for L large enough yields the proof.

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Proof of Theorem: Let α performance guarantee. Using Lemma 3, any $|\ell_i| \leq \alpha^{k-i+1}$ (by induction).

$$n = \sum_{i=1}^k |\ell_i| \leq \sum_{i=1}^k \alpha^{k-i+1} \leq \alpha^{k+1}.$$

By Lemma 2 $k \leq \alpha$. Therefore $\log n = O(\alpha \log \alpha)$ hence the lower bound.

Release dates

Positive result:

- Worst-case ratio of **5 (tight)** if $w_j/p_j = \beta$, for all j .

Release dates

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- Worst-case ratio of **5 (tight)** if $w_j/p_j = \beta$, for all j .
- For $w_j/p_j = \beta$, for all j the price of robustness **at least 3**.

Other results

- The algorithm can be extended to special cases of jobs with precedence constraints
- The first constant ratio algorithm for scheduling in the presence of unavailable periods
- If there is only one unavailability period (speed 0) then there exists a FPTAS

Independently achieved by Kellerer and Strusevich, but ours has smaller running time

Open problems

- Release dates:
 - Find a universal schedule that has performance ratio $O(\log n)$.
- Multiple machines:
 - Consider two machines; if both machines can break simultaneously then it is easy to see that there is no robust approximation algorithm unless $P=NP$ (reduction from partition)
 - Good robust model for multiple machines; e.g. (at most) 1 machine breaking/slowing down at a time

Open problems

- Expensive intervals:
 - Some time intervals are freely available and for all other (expensive) time units can be made available at cost c per time unit. Still minimising $\sum w_j C_j$
 - Without release dates easy, but not trivial
 - With release time and $p_j = 1$ is easy, but not trivial
 - With release times and general processing time not known (guess that it is easy without weights)