

$\frac{5}{4}(1 + \epsilon)$ -Approximation Algorithm for Scheduling with Rejection Costs Proportional to Processing Times

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Background

Scheduling with Rejection

- Topic of the paper: Scheduling n independent jobs on m machines
 - With a possibility to **reject** some jobs
- Solution (\mathcal{S})
 - Accepted jobs ($A^{\mathcal{S}}$) + associated schedule
 - Rejected jobs ($R^{\mathcal{S}}$)
- Objective
 - minimize($Z^{\mathcal{S}} = C^{\mathcal{S}} + \sum_{j \in R^{\mathcal{S}}} \text{RejectionCost}_j$)
 - where $C^{\mathcal{S}}$ denotes the makespan of jobs in $A^{\mathcal{S}}$

Scheduling with Rejection

- Many studies including
 - Bartal et al.¹: $(2 - \frac{1}{m})$ -approx algo in $\mathcal{O}(n \log n)$
 - Ou et al.²: $(\frac{3}{2} + \epsilon)$ -approx algo in $\mathcal{O}(n \log n + \frac{n}{\epsilon})$
 - Liu and Lu³: $(\frac{3}{2} - \frac{1}{2m})$ -approx algo in $\mathcal{O}(n^3 \log n)$
- **Our paper:** $\frac{5}{4}(1 + \epsilon)$ -approx algo in $\mathcal{O}(m^3(m + n) \log_{1+\epsilon} \rho)$
 - **Assumption:** Rejection costs proportional to processing times

¹Bartal, Y., Leonardi, S., Marchetti-Spaccamela, A., Sgall, J., Stougie, L.: Multi-processor scheduling with rejection. *SIAM Journal Disc Math* 13(1), 64-78, 2000

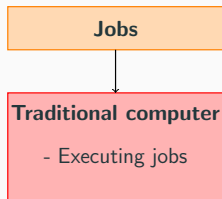
²Ou, J., Zhong, X., Wang, G.: An improved heuristic for parallel machine scheduling with rejection. *European Journal of Operational Research* 241(3), 653-661, 2015

³Liu, P., Lu, X.: New approximation algorithms for machine scheduling with rejection on single and parallel machine. *Journal of Comb Optimization* 40(4), 929-952, 2020

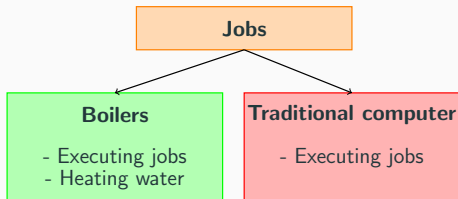
Context – Qarnot platform

- Scheduling pbs in the context of Qarnot platform (PULSE project)
- Qarnot: Cloud provider with highly distributed computational units
 - Recycling heat of computations to provide heat to the hosts
 - Avoiding energy waste on cooling systems

Traditional Datacenter



Qarnot Computing



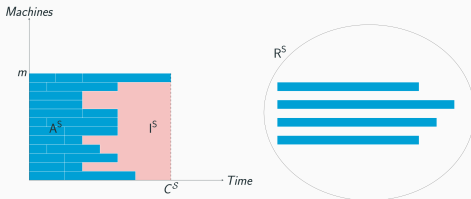
- Today: boiler maintenance scheduled at time T
- Goal: schedule maintenance close to T and minimize energy

Problem Formulation - Scheduling with Rejection

- Inputs:
 - n non-preemptive independent jobs:
 - $\forall j \in \{1, \dots, n\}, p_j$: processing time
 - $\forall j \in \{1, \dots, n\}, \rho p_j$: cost to run on expensive machine (ρ is fixed)
 - m identical machines on boiler (cheap)
 - ∞ identical machines on e.g. public provider (expensive)
- Constraints:
 - Each job requires exactly one machine
 - Each machine can run at most one job at a time
 - All boiler machines are turned ON when any job running on boiler
- Outputs (Decision variables):
 - $\forall i \in \{1, \dots, m\}, X_i$ is the list of jobs assigned to the cheap machine i
- Definitions:
 - $C^S = \max_{i \in \{1, \dots, m\}} \sum_{j \in X_i} p_j$ (no dependencies)
 - $R^S = \sum_{j \notin (X_i)_{i \in \{1, \dots, m\}}} p_j$
- Objective:
$$\text{minimize}(Z^S = m * C^S + \rho R^S)$$

Problem Visualization - Scheduling with Rejection

- Left figure shows boiler machines vs time
 - Blue is occupied time; Pink is idle time
- Right figure shows the jobs on expensive machines (called **rejected**)

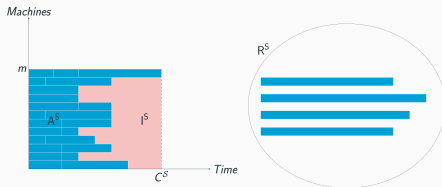


- Objective: $Z^S = m * C^S + \rho R^S$
- $m * C^S$: energy on boiler
- ρR^S : energy on traditional resources

Solution for a makespan bound T

Maintenance with target T

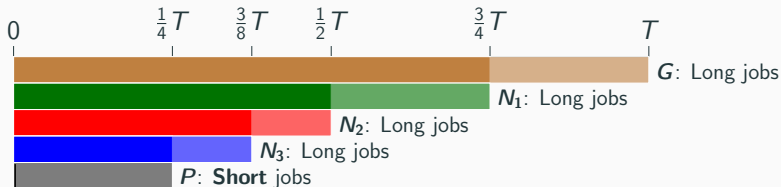
- Optimal Solution if maintenance is scheduled at time T
 - $Z^{opt} = m * T + \rho R^{opt}(T)$
- Our goal: find C^S and R^S such that



- $Z^S = m * C^S + \rho R^S \leq \frac{5}{4} Z^{opt}$

Our approach for fixed makespan T

- Idea from Caprara et al.⁴ for Multiple Subset Sum Problem solution

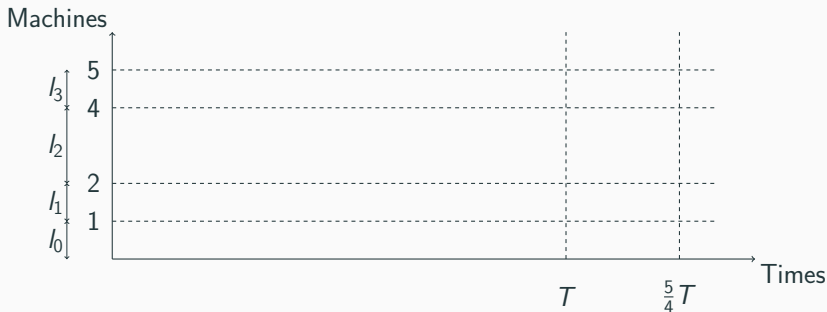


- For fitting in T , only possible long job combinations on a machine:
 - One job possibilities: (G) , (N_1) , (N_2) , (N_3)
 - Two job possibilities: (N_1, N_2) , (N_1, N_3) , (N_2, N_2) , (N_2, N_3) , (N_3, N_3)
 - Three job possibilities: (N_2, N_3, N_3) , (N_3, N_3, N_3)
- Our solution for fixed makespan T :
 - Use only these combinations of long jobs (at most $\frac{5}{4}T$ makespan)
 - Finally, assign short jobs greedily within makespan $\leq \frac{5}{4}T$

⁴Caprara, A., Kellerer, H., Pferschy, U.: A 3/4-approximation algorithm for multiple subset sum. Journal of Heuristics 9(2), 99–111 (March 2003)

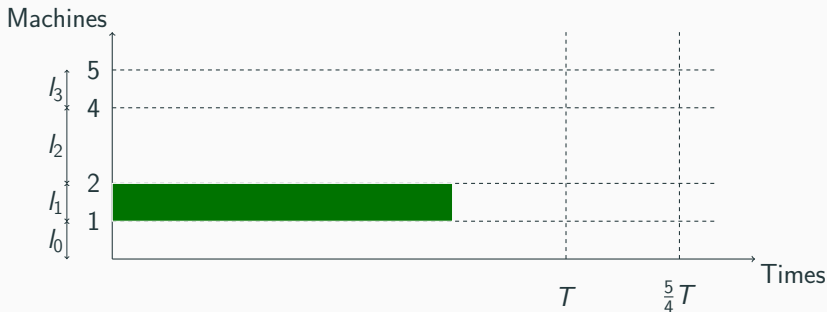
Our algorithm for fixed makespan T : *FillMaxArea*

- Assume: $m=5$, $|G|=0$, $|N_1|=5$, $|N_2|=1$, $|N_3|=5$, $|P|=17$
- l_0 , l_1 , l_2 and l_3 stand for the number of machines that run no long job, one long job, two long jobs and three long jobs, respectively
- Fix $(l_0, l_1, l_2, l_3) = (1, 1, 2, 1)$



Our algorithm for fixed makespan T : *FillMaxArea*

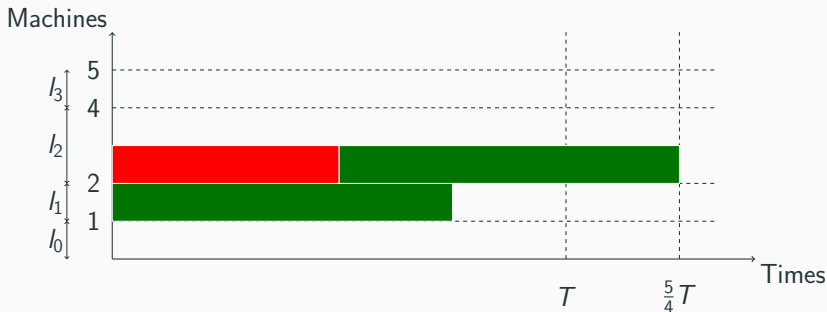
- Assume: $m=5$, $|G|=0$, $|N_1|=4$, $|N_2|=1$, $|N_3|=5$, $|P|=17$
- l_0 , l_1 , l_2 and l_3 stand for the number of machines that run no long job, one long job, two long jobs and three long jobs, respectively
- Fix $(l_0, l_1, l_2, l_3) = (1, 1, 2, 1)$



- Start with singletons, i.e. (G) , (N_1) , (N_2) , (N_3)
 - No job in $G \implies$ Take the longest job in N_1

Our algorithm for fixed makespan T : *FillMaxArea*

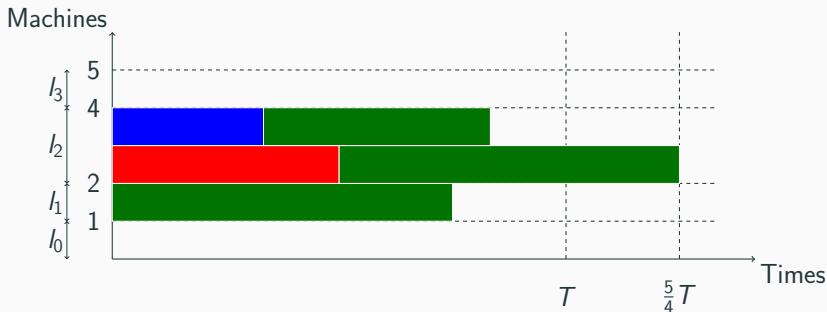
- Assume: $m=5$, $|G|=0$, $|N_1|=3$, $|N_2|=0$, $|N_3|=5$, $|P|=17$
- l_0 , l_1 , l_2 and l_3 stand for the number of machines that run no long job, one long job, two long jobs and three long jobs, respectively
- Fix $(l_0, l_1, l_2, l_3) = (1, 1, 2, 1)$



- Now assign pairs, i.e. (N_1, N_2) , (N_1, N_3) , (N_2, N_2) , (N_2, N_3) , (N_3, N_3)
 - Longest job in N_2 and longest job in N_1

Our algorithm for fixed makespan T : *FillMaxArea*

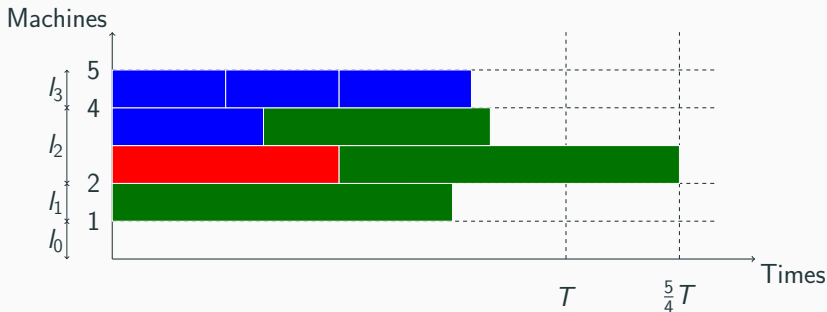
- Assume: $m=5$, $|G|=0$, $|N_1|=2$, $|N_2|=0$, $|N_3|=4$, $|P|=17$
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- Fix $(l_0, l_1, l_2, l_3) = (1, 1, 2, 1)$



- Assign pairs, i.e. (N_2, N_1) , (N_3, N_1) , (N_2, N_2) , (N_3, N_2) , (N_3, N_3)
 - No job in $N_2 \implies$ Longest job in N_3 and longest job in N_1

Our algorithm for fixed makespan T : *FillMaxArea*

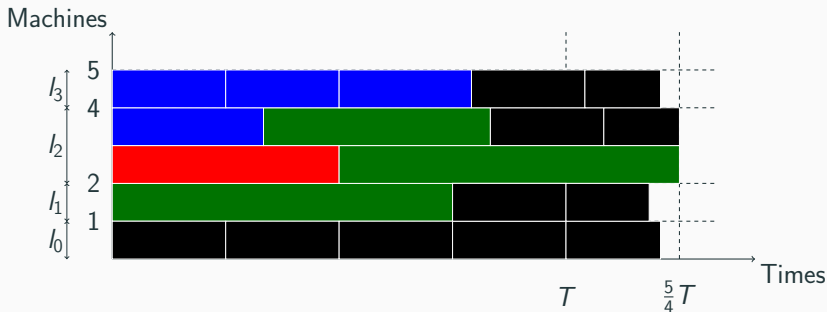
- Assume: $m=5$, $|G|=0$, $|N_1|=2$, $|N_2|=0$, $|N_3|=1$, $|P|=17$
- l_0 , l_1 , l_2 and l_3 stand for the number of machines that run no long job, one long job, two long jobs and three long jobs, respectively
- Fix $(l_0, l_1, l_2, l_3) = (1, 1, 2, 1)$



- Assign triplets, i.e. (N_3, N_3, N_2) , (N_3, N_3, N_3)
 - No job in $N_2 \implies$ 3 longest jobs in N_3

Our algorithm for fixed makespan T : *FillMaxArea*

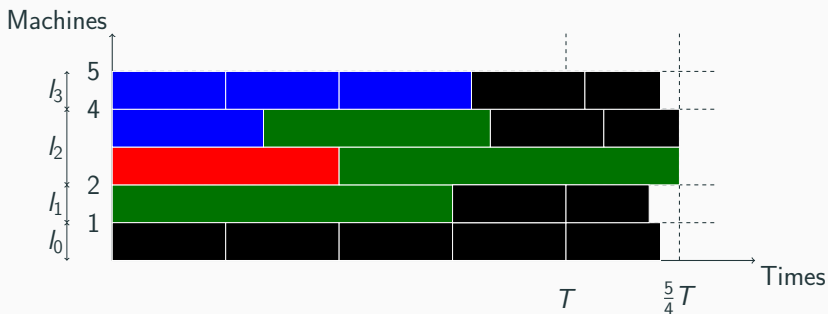
- Assume: $m=5$, $|G|=0$, $|N_1|=2$, $|N_2|=0$, $|N_3|=1$, $|P|=6$
- l_0 , l_1 , l_2 and l_3 stand for the number of machines that run no long job, one long job, two long jobs and three long jobs, respectively
- Fix $(l_0, l_1, l_2, l_3) = (1, 1, 2, 1)$



- Assign P jobs greedily while makespan $\leq \frac{5}{4}T$

Our algorithm for fixed makespan T : *FillMaxArea*

- Assume: $m=5$, $|G|=0$, $|N_1|=2$, $|N_2|=0$, $|N_3|=1$, $|P|=6$
- l_0 , l_1 , l_2 and l_3 stand for the number of machines that run no long job, one long job, two long jobs and three long jobs, respectively
- Fix $(l_0, l_1, l_2, l_3) = (1, 1, 2, 1)$



- Find solution for each $(l_0, l_1, l_2, l_3) \rightarrow \mathcal{O}(m^3)$ iterations
 - Choose solution with largest $\sum_{j \in \text{AcceptedJobs}} p_j$

FillMaxArea - Main result

Overall algorithm:

- Iterate above algorithm for all l_0, l_1, l_2, l_3 such that $\sum l_i = m$
- there are "only" $\Theta(m^3)$ such quadruplets...
- and keep the solution that minimizes $Z^S = m * C^S + \rho R^S$

RESULT

For any T , let \dagger be the solution obtained by $\text{FillMaxArea}(J, m, T)$.
Then, $C^\dagger \leq \frac{5}{4}T$ and $R^\dagger \leq R^*(T)$, where $R^*(T) = \min_{S, C^S \leq T} R^S$.

FillMaxArea - Main result

Overall algorithm:

- Iterate above algorithm for all l_0, l_1, l_2, l_3 such that $\sum l_i = m$
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Sketch of Proof:

- $C^\dagger \leq \frac{5}{4}T$: straightforward given the upper bounds of combinations
- FillMaxArea allocates as much work as possible on the boiler given l_0, l_1, l_2, l_3
 - $R^\dagger \leq R^*(T)$



RESULT

For any T , let \dagger be the solution obtained by $\text{FillMaxArea}(J, m, T)$.
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LEMMA

For any T , let \dagger be the solution obtained by $\text{FillMaxArea}(J, m, T)$.
We can bound its cost by: $Z^\dagger \leq \frac{5}{4}Tm + \rho R^*(T)$.

RESULT

For any T , let \dagger be the solution obtained by $\text{FillMaxArea}(J, m, T)$. Then, $C^\dagger \leq \frac{5}{4}T$ and $R^\dagger \leq R^*(T)$, where $R^*(T) = \min_{S, C^S \leq T} R^S$.

LEMMA

For any T , let \dagger be the solution obtained by $\text{FillMaxArea}(J, m, T)$. We can bound its cost by: $Z^\dagger \leq \frac{5}{4}Tm + \rho R^*(T)$.

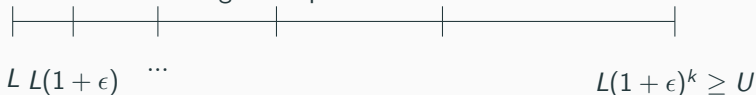
Proof:

- Remember: Cost of \dagger is $Z^\dagger = m * C^\dagger + \rho R^\dagger$
- $C^\dagger \leq \frac{5}{4}T$
- $R^\dagger \leq R^*(T)$
 - $Z^\dagger \leq \frac{5}{4}Tm + \rho R^*(T) \leq \frac{5}{4}Z^{\text{opt}}$

Determining the makespan bound T

Determining Makespan: BEKP method

- If C^{OPT} is known \implies $FillMaxArea(C^{OPT})$ is a $\frac{5}{4}$ -approx. algorithm
- In the general problem C^{OPT} is not known...
- Assume: Lower and Upper bound (L and U) on makespan such that
 - Rejecting all jobs when makespan $C^{OPT} \leq L$ and $C^{OPT} \geq U$ is valid
- Iterate over remaining makespans



- ϵ : User defined small coefficient
- Iteration number: $k = \lceil \frac{\log(\frac{U}{L})}{\log(1+\epsilon)} \rceil$

Determining Makespan: BEKP method

Algorithm 1 $BEKP(J, m)$

- 1: The solution where all jobs are rejected is X_0
 - 2: Compute U and L
 - 3: Set k as the first value such that $(1 + \epsilon)^k L \geq U$
 - 4: **for** each $C_i \in \{L, (1 + \epsilon)L, (1 + \epsilon)^2 L, \dots, (1 + \epsilon)^k L\}$ **do**
 - 5: $X_i = \text{FillMaxArea}(J, m, C_i)$
 - 6: **return** A schedule with the lowest cost among X_0 and all X_i
-

Determining Makespan: BEKP method

Algorithm 2 $BEKP(J, m)$

- 1: The solution where all jobs are rejected is X_0
 - 2: Compute U and L
 - 3: Set k as the first value such that $(1 + \epsilon)^k L \geq U$
 - 4: **for** each $C_i \in \{L, (1 + \epsilon)L, (1 + \epsilon)^2 L, \dots, (1 + \epsilon)^k L\}$ **do**
 - 5: $X_i = \text{FillMaxArea}(J, m, C_i)$
 - 6: **return** A schedule with the lowest cost among X_0 and all X_i
-

WHAT IS LEFT?

- Compute L and U
- Proof: $BEKP$ is a $\frac{5}{4}(1 + \epsilon)$ -approximation algorithm.

Computing L and U

- W : sum of all processing times of jobs in ready queue
- $R^*(C)$: minimum possible area for the rejected jobs for makespan C
- Lower bound function on the cost: $f(C) = Cm + \rho R^*(C)$

Computing L and U

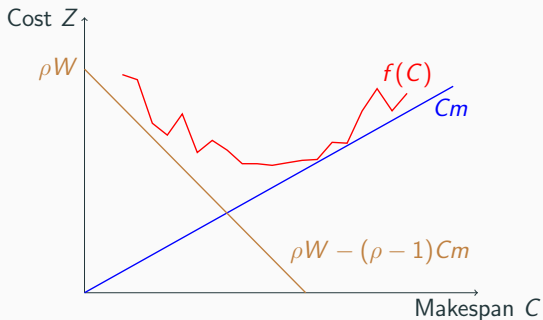
- W : sum of all processing times of jobs in ready queue
- $R^*(C)$: minimum possible area for the rejected jobs for makespan C
- Lower bound function on the cost: $f(C) = Cm + \rho R^*(C)$
- $f(C) \geq Cm$

Computing L and U

- W : sum of all processing times of jobs in ready queue
- $R^*(C)$: minimum possible area for the rejected jobs for makespan C
- Lower bound function on the cost: $f(C) = Cm + \rho R^*(C)$
- $f(C) \geq Cm$
- $f(C) \geq Cm + \rho(W - Cm)$

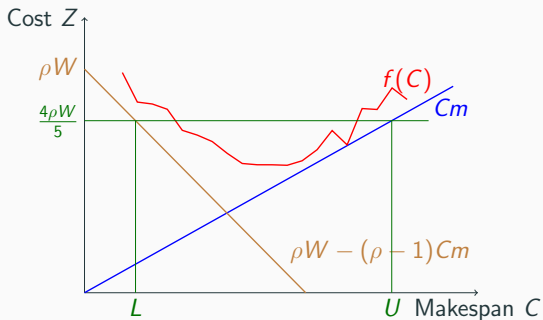
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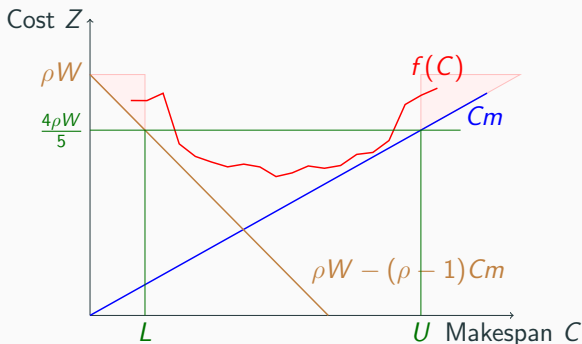
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- Lower bound function on the cost: $f(C) = Cm + \rho R^*(C)$
- $f(C) \geq Cm$
- $f(C) \geq Cm + \rho(W - Cm)$



- $H = \frac{4\rho W}{5}$ (We set this value)

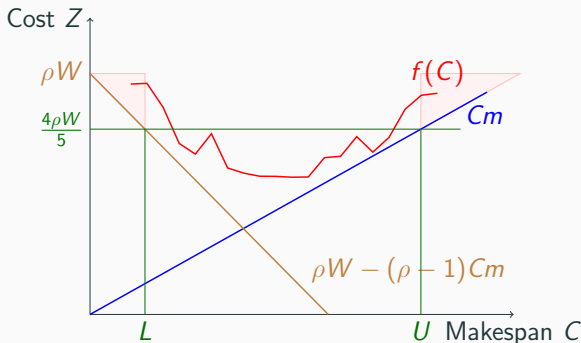
Computing L and U



- If we reject all jobs for makespan values $\leq L$ and $\geq U$, worst cost:

😊 $\frac{Z^{X_0}}{Z^{OPT}} \leq \frac{\rho W}{\frac{4\rho W}{5}} = \frac{5}{4}$ (Valid solution)

Computing L and U



- If we reject all jobs for makespan values $\leq L$ and $\geq U$, worst cost:
😊 $\frac{Z^{X_0}}{Z^{OPT}} \leq \frac{\rho W}{\frac{4\rho W}{5}} = \frac{5}{4}$ (Valid solution)
- Remember: $k = \lceil \frac{\log(\frac{U}{L})}{\log(1+\epsilon)} \rceil$
- $U = \frac{4\rho W}{5m}$ and $L = \frac{\rho W}{5m(\rho-1)} \implies \frac{U}{L} = 4(\rho-1)$
- 😊 For $\rho \leq 10$ and $\epsilon \geq 0.05$, $k = 74$ (Cheap iteration)

Proof of the approximation ratio

THEOREM

For any positive ϵ , \mathcal{BEP} is a $\frac{5}{4}(1 + \epsilon)$ -approximation algorithm.

Proof

- For $C^{OPT} < L$ or $C^{OPT} > U$ already proved.

Proof of the approximation ratio

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For any positive ϵ , \mathcal{BEP} is a $\frac{5}{4}(1 + \epsilon)$ -approximation algorithm.

Proof

- For $C^{OPT} < L$ or $C^{OPT} > U$ already proved. **For $L \leq C^{OPT} \leq U$:**



Proof of the approximation ratio

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- For $C^{OPT} < L$ or $C^{OPT} > U$ already proved. **For $L \leq C^{OPT} \leq U$:**



- X_i : Solution by using C_j
- $Z^{X_i} \leq \frac{5}{4}mC_j + \rho R^*(C_j)$

Proof of the approximation ratio

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- X_i : Solution by using C_i
- $Z^{X_i} \leq \frac{5}{4}mC_i + \rho R^*(C_i)$
- As $C^{OPT} \leq C_i$
 - $R^*(C_i) \leq R^*(C^{OPT})$

Proof of the approximation ratio

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Proof

- For $C^{OPT} < L$ or $C^{OPT} > U$ already proved. **For $L \leq C^{OPT} \leq U$:**



- X_i : Solution by using C_i
- $Z^{X_i} \leq \frac{5}{4}mC_i + \rho R^*(C_i)$
- As $C^{OPT} \leq C_i$
 - $R^*(C_i) \leq R^*(C^{OPT})$
- As $C_i \leq (1 + \epsilon)C^{OPT}$
 - $Z^{X_i} \leq \frac{5}{4}(1 + \epsilon)mC^{OPT} + \rho R^*(C^{OPT}) \leq \frac{5}{4}(1 + \epsilon)Z^{OPT}$

Conclusion - Future work

- In practice (simulations)
 - Very good qualitative results (much lower than $\frac{5}{4}$)
 - Low computational time ($Km^3(m+n)$ where m machines, n jobs)
- *BEKP* compared to *LIULU*
 - Improves approximation ratio (for proportional rejection costs)
 - Proportional rejection costs can be extended to other contexts
 - Improves complexity wrt total number of jobs
- Next step: consider other scheduling problems related to Qarnot
 - With deadlines (each job has a duration and a deadline)
 - the true problem is online and non clairvoyant...
 - probably very difficult to prove something...

Thank You

Backup slides

Experiments

Experimental setting

- Run in sequential on Miriel nodes of Plafrim
 - 2 INTEL Xeon E5-2680v3 12-core 2.50 GHz processors with 128 GB
- Processing times are generated through lognormal distribution
 - Mean 3
 - Standard deviations (σ): "0.5", "0.7", and "1.0"
 - Large σ means higher variance in processing times
- Number of machines: $m = 20$
- Number of jobs: $n = 4m$
- Our simulation code is available as free software in⁵

⁵Beaumont, O., Eyraud-Dubois, L., Korkmaz, E., Pilla, L.L.: Experimental codes and results for the paper "a $5/4(1+\epsilon)$ -approximation algorithm for scheduling with rejection costs proportional to processing times".

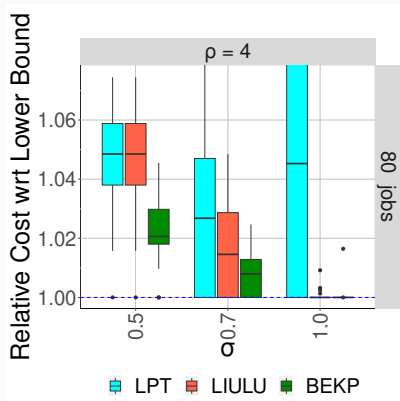
<https://inria.hal.science/hal-04517532>, accessed: March 25, 2024

Methods compared

- *BEKP*: Our method
 - $\mathcal{O}(m^3(m+n) \log_{1+\epsilon} \rho)$
 - $\frac{5}{4}(1+\epsilon)$ approximation
- *LIULU*: The algorithm proposed by Liu and Lu⁶
 - $\mathcal{O}(n^3 \log n)$
 - $(\frac{3}{2} - \frac{1}{2m})$ approximation
 - Assumes arbitrary rejection costs
- *LPT*: A cheap and naive solution (No rejection)
 - Uses Longest Processing Time first method

⁶Liu, P., Lu, X.: New approximation algorithms for machine scheduling with rejection on single and parallel machine. *Journal of Comb Optimization* 40(4), 929-952 (2020)

Comparison of scheduler costs



- Each box-plot represents 30 different experimental cases
- *LPT*: No guarantee on the cost bound
- *BEKP*: Better costs in general compared to *LIULU*

Our algorithm for fixed makespan T

Algorithm 3 *FillMaxArea*(J, m, T)

- 1: Generate G, N_1, N_2, N_3 and P subsets of J
 - 2: **for** each $j = (l_0, l_1, l_2, l_3)$ such that $l_0 + l_1 + l_2 + l_3 = m$ and $l_1 + 2l_2 + 3l_3 \leq n$
do
 - 3: $X_j \leftarrow \emptyset$
 - 4: $X_j \leftarrow X_j \cup \text{AssignFrom}(\{(G), (N_1), (N_2), (N_3)\}, l_1)$
 - 5: $X_j \leftarrow X_j \cup \text{AssignFrom}(\{(N_2, N_1), (N_3, N_1), (N_2, N_2), (N_3, N_2), (N_3, N_3)\}, l_2)$
 - 6: $X_j \leftarrow X_j \cup \text{AssignFrom}(\{(N_3, N_3, N_2), (N_3, N_3, N_3)\}, l_3)$
 - 7: If $l_0 + |X_j| < m$, discard X_j
 - 8: Add jobs from P greedily (in any order) to X_j , keeping makespan $\leq \frac{5}{4} T$
 - 9: $X^* = \{X_j \mid \max_j A^{X_j}\}$
 - 10: **return** X^*
-

Our algorithm for fixed makespan T

Algorithm 4 *AssignFrom*(*combs*, l)

- 1: *Result* $\leftarrow \emptyset$
 - 2: Remove all combinations from *combs* where at least one set within the combination is empty
 - 3: **while** $|Result| \leq l$ and *combs* is not empty **do**
 - 4: Denote by (K_1, K_2, \dots, K_k) the first combination in *combs*
 - 5: $j_1 \leftarrow$ the largest job from K_1 , $j_2 \leftarrow$ the largest remaining job from $K_2 \dots$
 - 6: Continue until $j_k \leftarrow$ the largest remaining job from K_k
 - 7: $Result = Result \cup (j_1, j_2, \dots, j_k)$
 - 8: Remove all combinations from *combs* where at least one set within the combination is empty
 - 9: **return** *Result*
-

Comparison of real-life scheduler running times

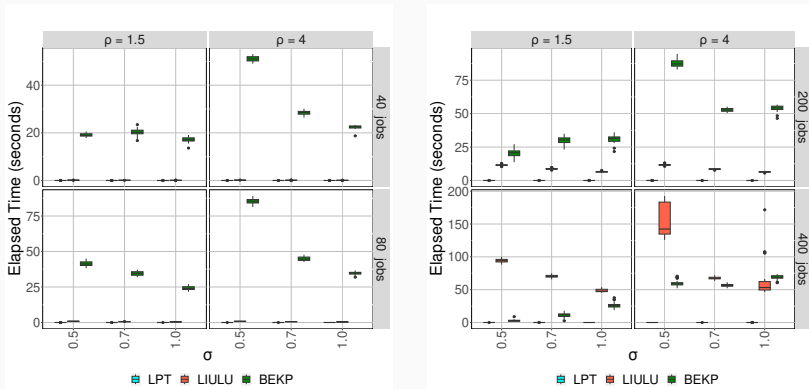


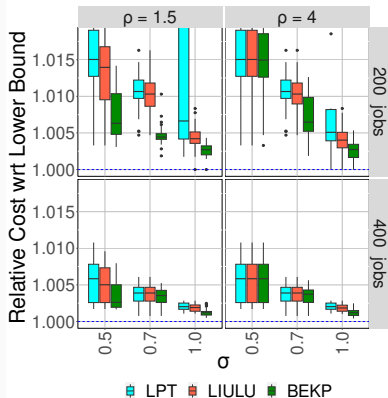
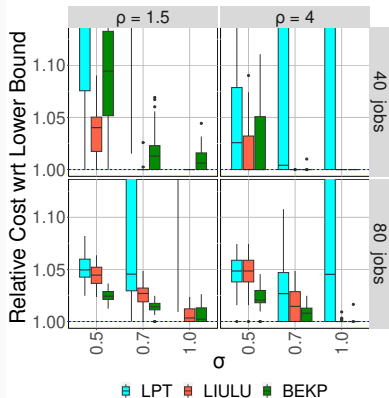
Figure 1: Comparison of LPT , $LIULU$ and $BEKP$ using $m = 20$ for different number of jobs, different values for ρ and σ . Each box-plot represents 30 different experimental cases for the corresponding configuration.

Methods compared

- **BEKP**: Our method
 - $\mathcal{O}(m^3(m+n)\log_{1+\epsilon}\rho)$
 - $\frac{5}{4}(1+\epsilon)$ approximation
- **LIULU**: The algorithm proposed by Liu and Lu⁷
 - $\mathcal{O}(n^3\log n)$
 - $(\frac{3}{2} - \frac{1}{2m})$ approximation
 - Assumes arbitrary rejection costs
- **LPT**: A cheap and naive solution (No rejection)
 - Uses Longest Processing Time first method
- **Lower Bound**: Reference method
 - ILP: $x_i = 1$ if job is accepted; else $x_i = 0$
 - $\forall i \in J, C \geq x_i p_i$ and $C \geq \sum_{i \in J} (x_i p_i) / m$
 - minimize $Cm + \sum_{i \in J} \rho(1 - x_i)p_i$
- No performance optimization for methods

⁷Liu, P., Lu, X.: New approximation algorithms for machine scheduling with rejection on single and parallel machine. Journal of Comb Optimization 40(4), 929-952 (2020)

Comparison of scheduler costs



- Each box-plot represents 30 different experimental cases
- *LPT*: No guarantee on the cost bound
- *BEKP*: Better costs in general compared to *LIULU*