

A Subexponential Time Algorithm for Makespan Scheduling of Unit Jobs with Precedence Constraints

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Karol Węgrzycki



Saarland
University

$$Pm|prec, p_j = 1|C_{\max}$$

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m identical
parallel machines

$$m = 2$$

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Given:

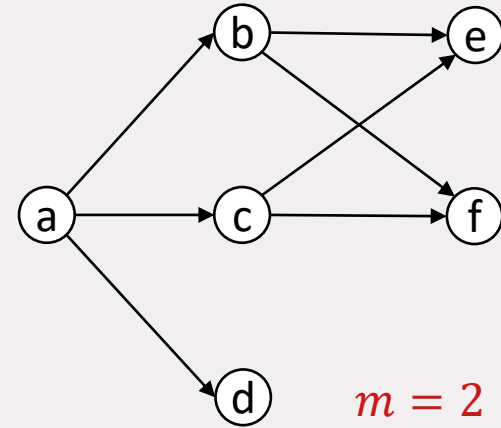
- n jobs of length 1

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Given:

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- A precedence graph G

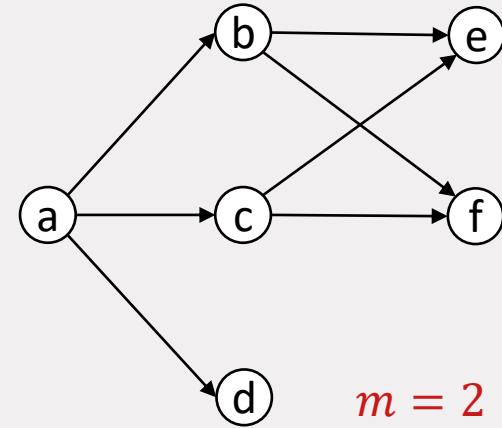


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Given:

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- A precedence graph G
- $T \in \mathbb{N}$

Q: Is there a schedule of makespan T ?

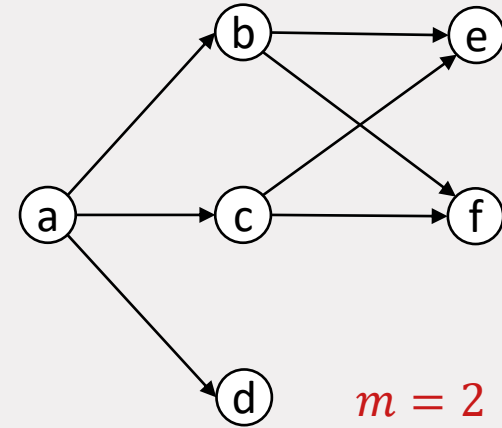


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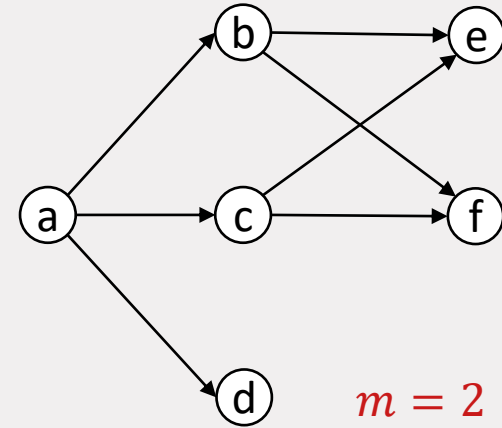
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Observation:

Jobs of length one \Rightarrow 'timeslots'



time \rightarrow

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List of Open Problems by Garey and Johnson 1979

1. Graph Isomorphism
2. Subgraph Homeomorphism
3. Graph genus
4. Chordal graph completion
5. Chromatic index
6. Spanning tree parity problem
7. Partial order dimension
8. Precedence constrained 3-processor scheduling
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$2^{O((\log n)^3)}$ time
[Babai 2017]

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$2^{O(\sqrt{n} \cdot \log n)}$ time
This talk

Literature overview $Pm|prec, p_j = 1|C_{\max}$

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Before:

$Pm|prec, p_j = 1|C_{\max}$ can be solved in $O\left(2^n \cdot \binom{n}{m}\right)$ time.

Our Result

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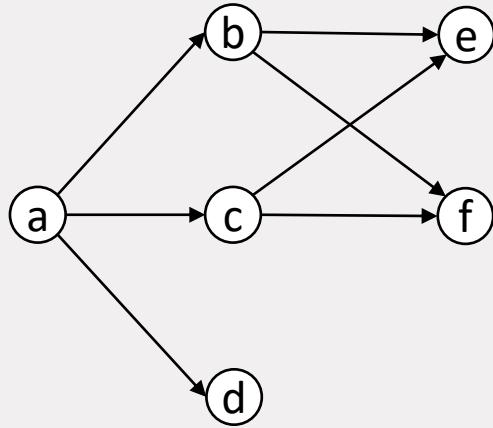
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Key points in algorithm:

1. New **decomposition** of schedules
2. Use of **look-up table**
3. Use of **Dynamic Programming** in combining results

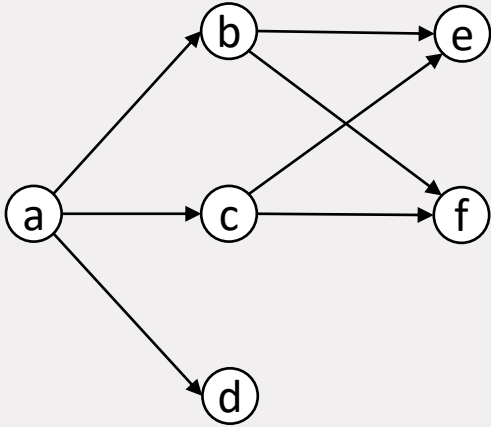
Definitions



Precedence Constraints Graph G

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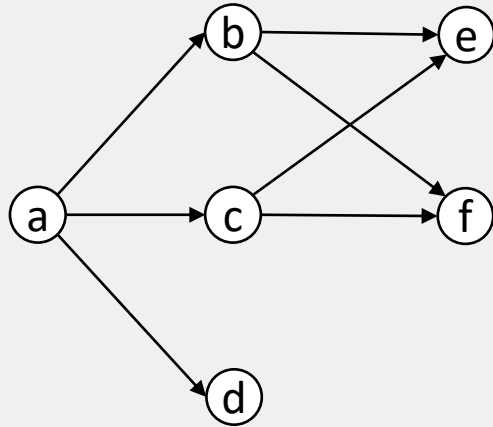
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- $i < j$ if $(i, j) \in G$



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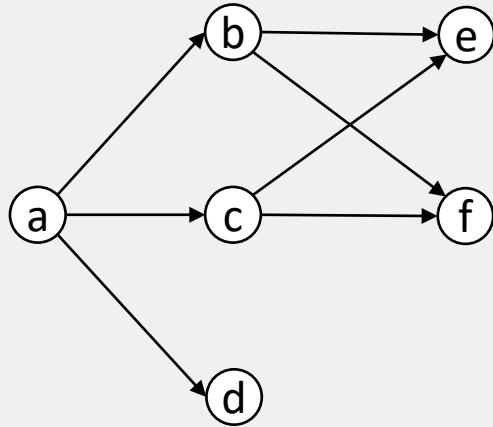
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Definitions: Let A be a set of jobs.

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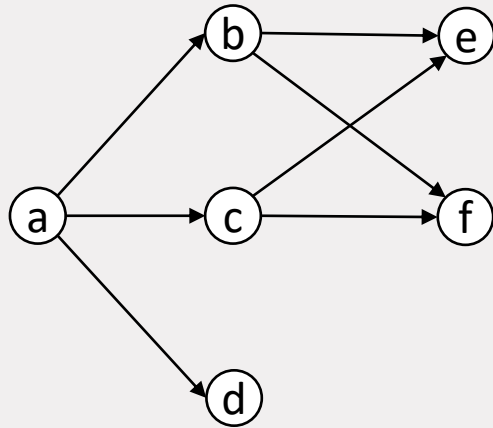
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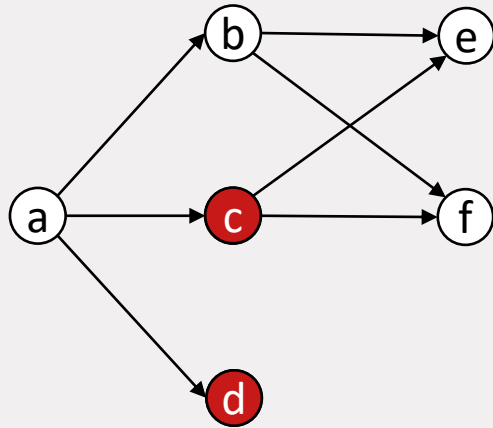
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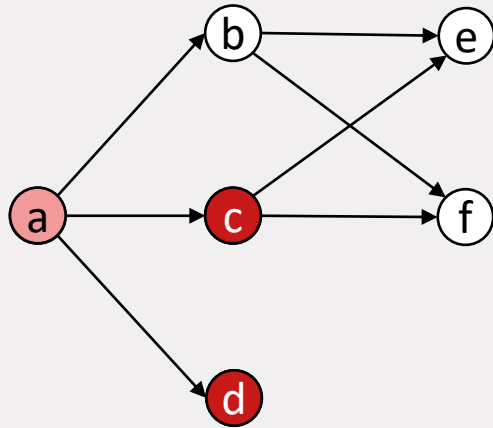
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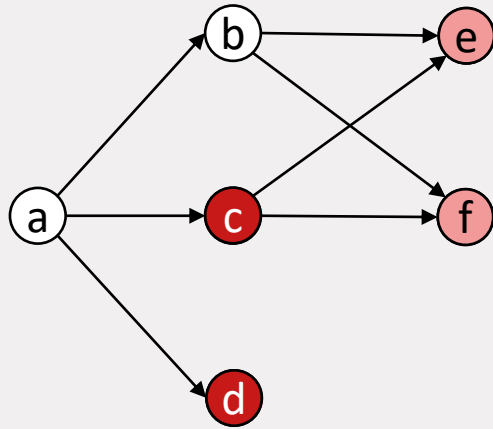
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Let $A = \{c, d\}$, then:

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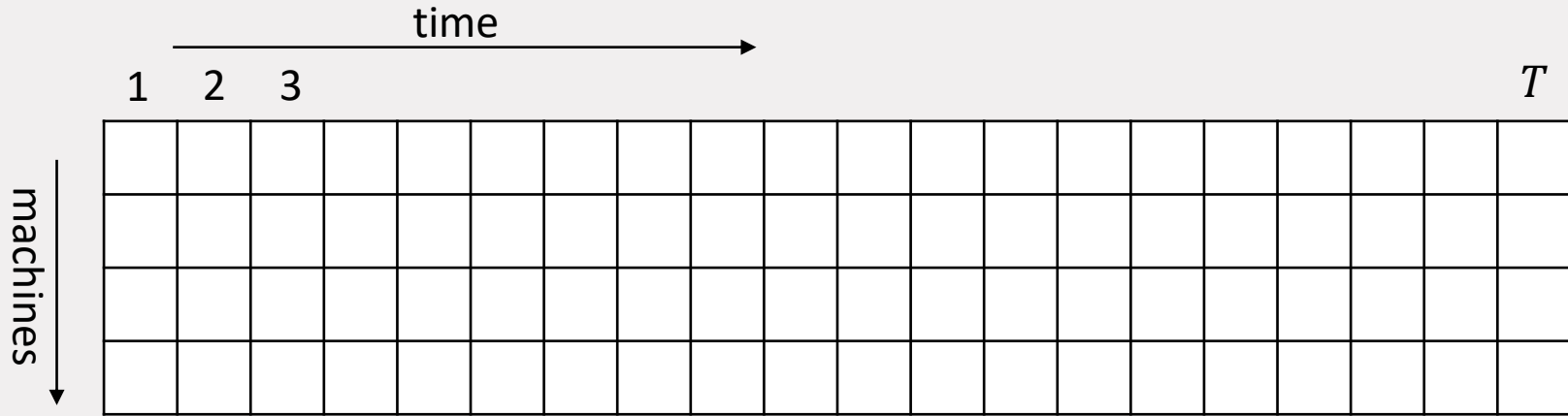
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- $\text{succ}[A] = \{c, d, e, f\}$
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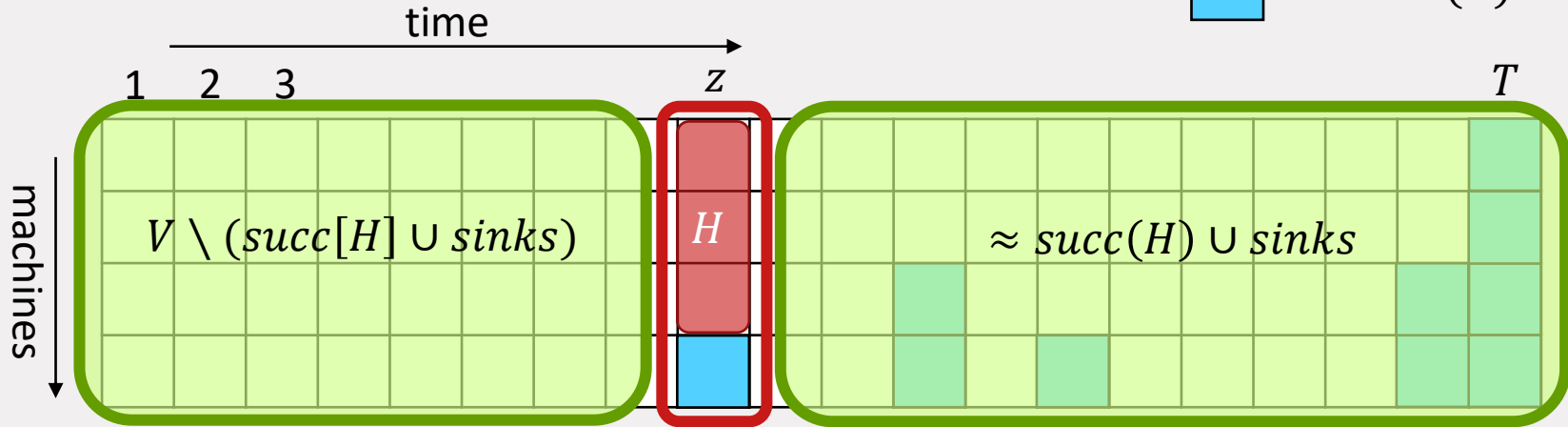
Zero-Adjusted Schedule (D&W)



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Assumption: $n = m \cdot T$

 = sinks(G)




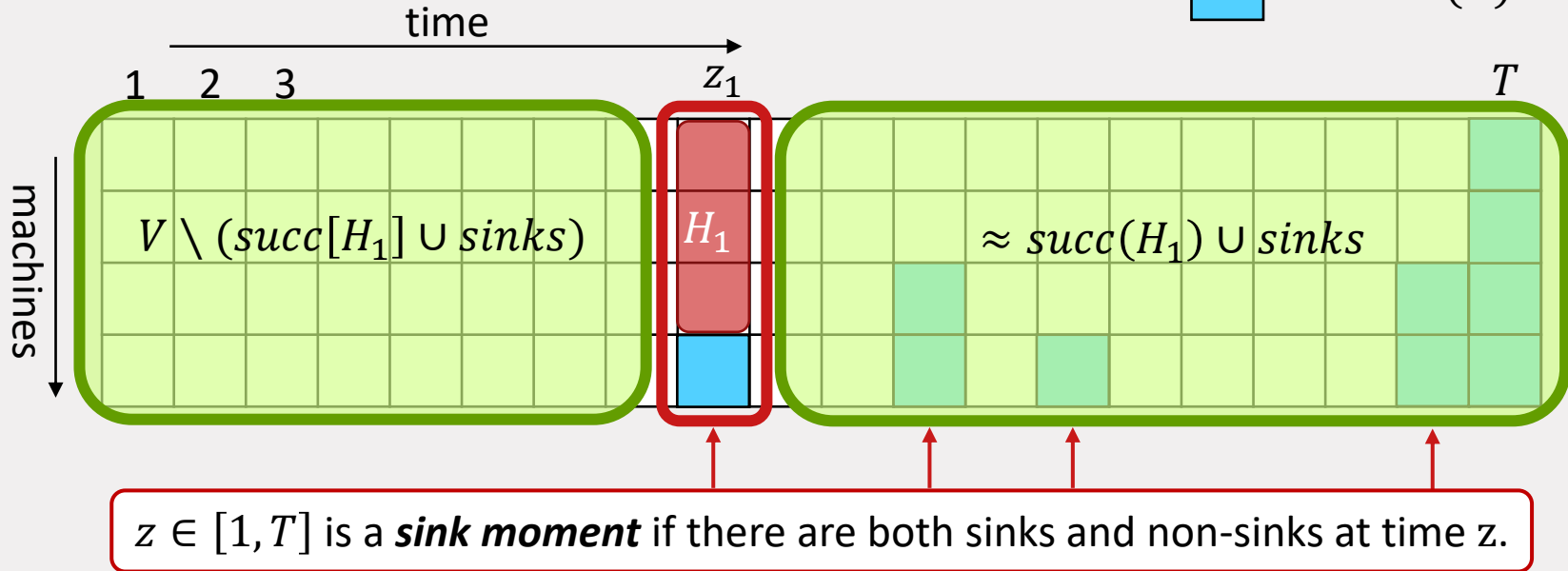
Let $z \in [1, T]$ be the first moment with a sink.

D&W: W.m.a. Each job x after z is a sink or a successor of a job at time z .

Sink-Adjusted Schedule


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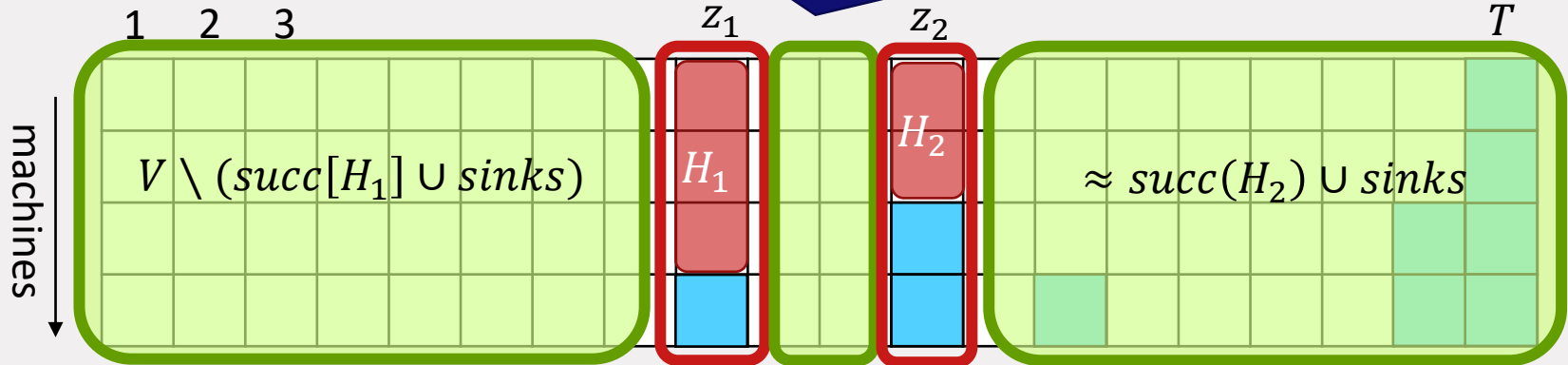


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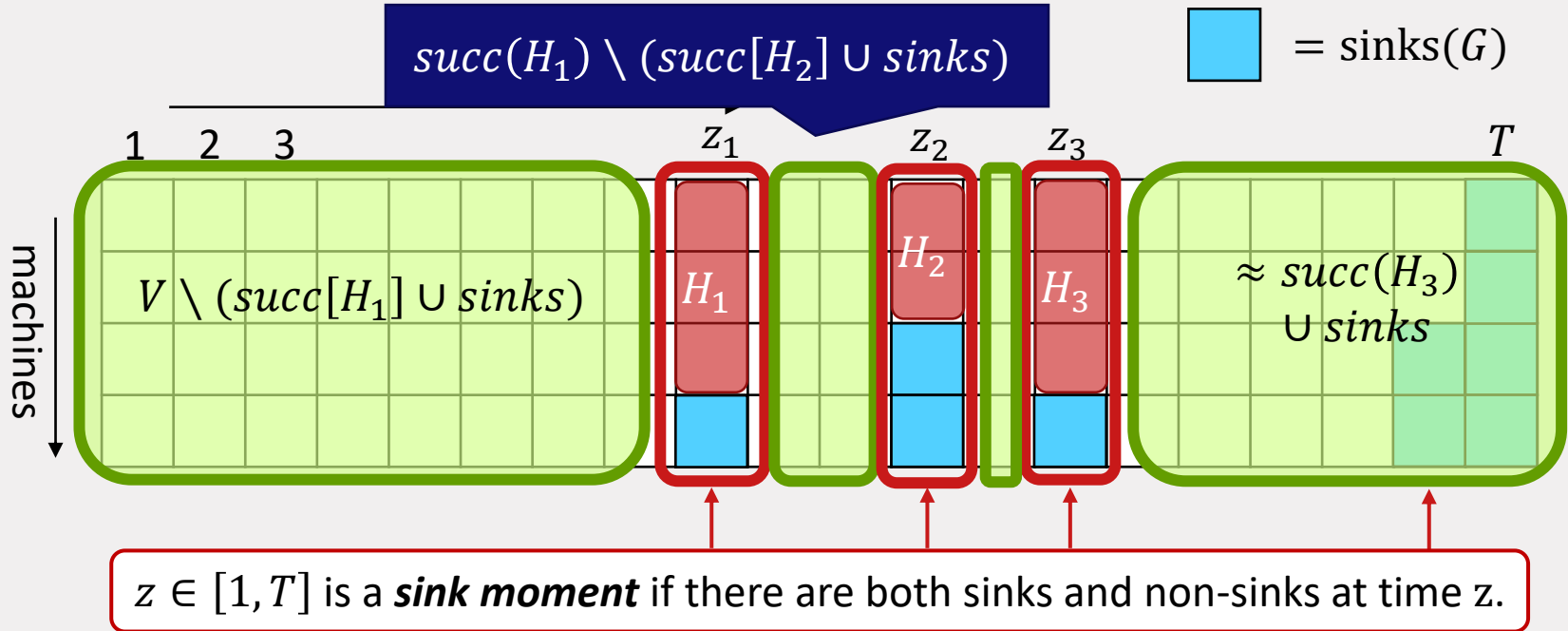
$\text{succ}(H_1) \setminus (\text{succ}[H_2] \cup \text{sinks})$



$z \in [1, T]$ is a **sink moment** if there are both sinks and non-sinks at time z .

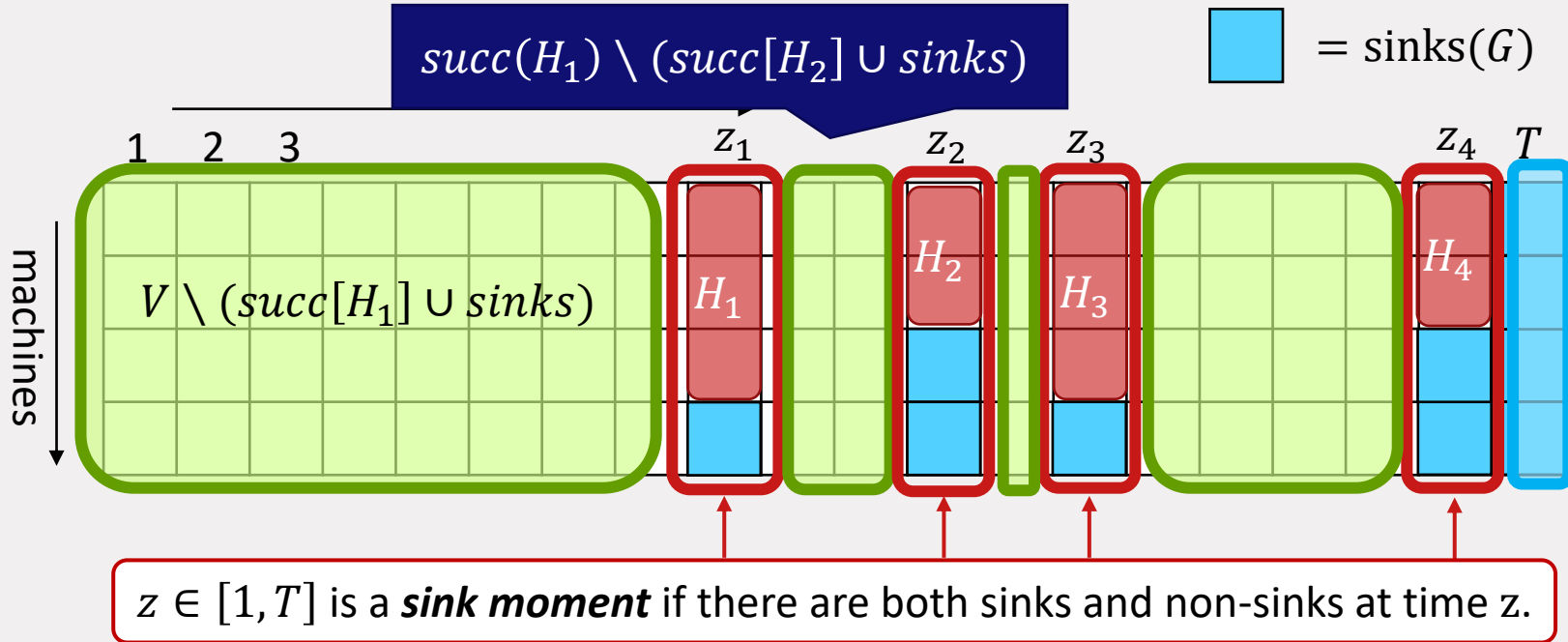
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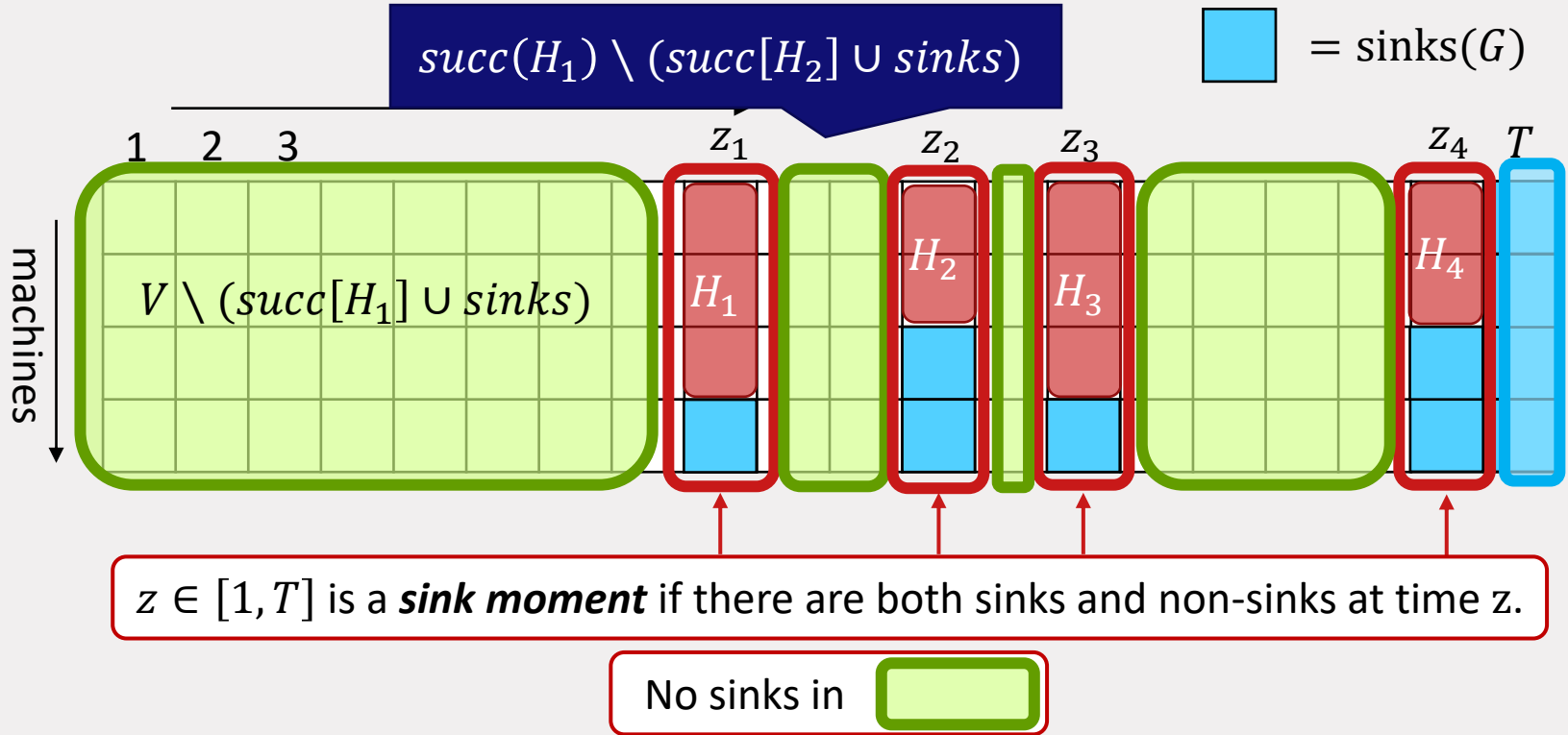
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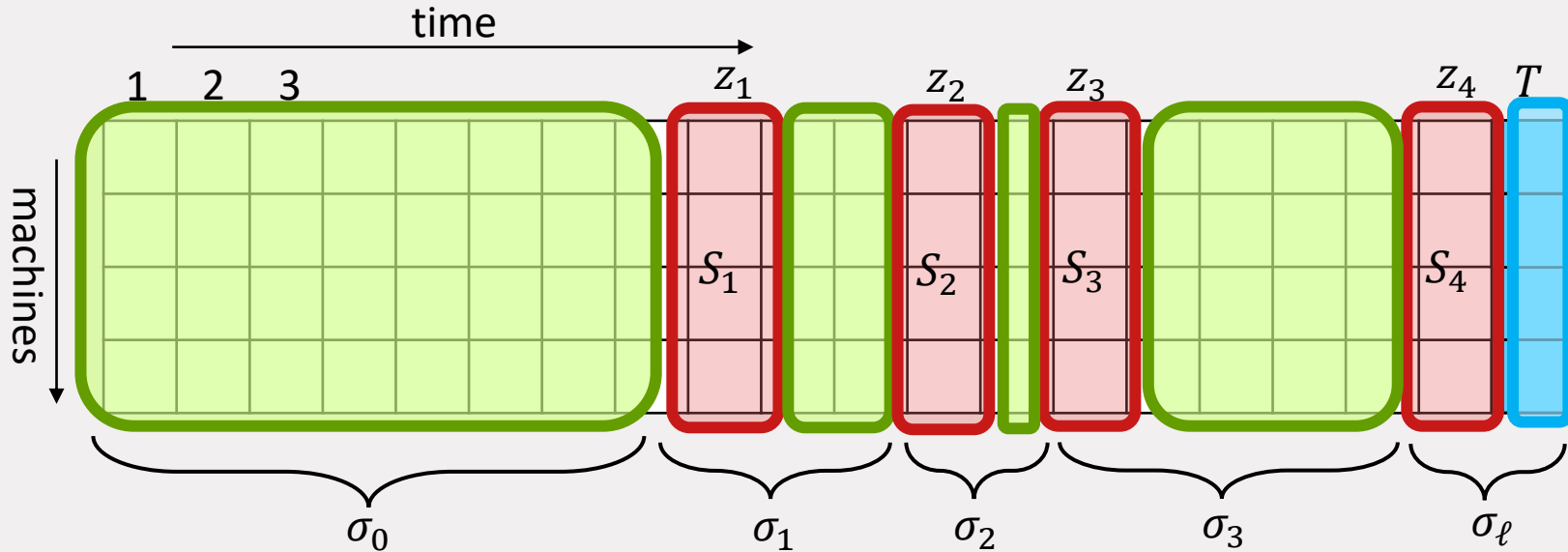
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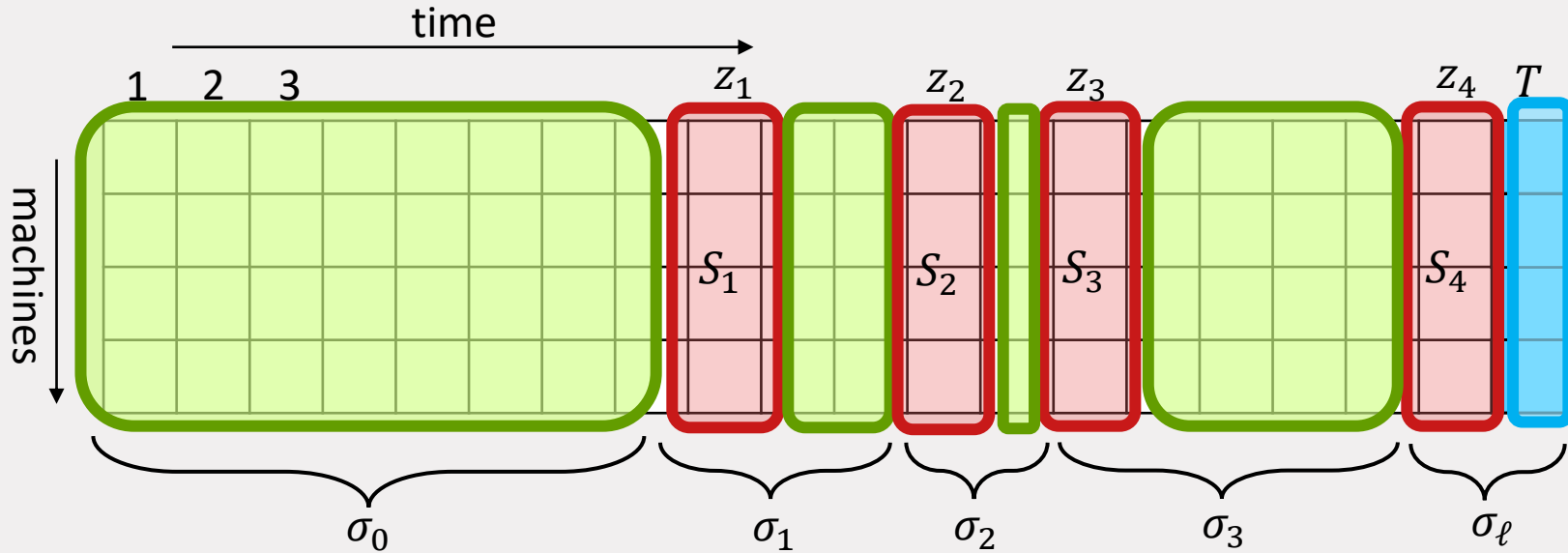
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$$\sigma_i = \sigma(S_i, S_{i+1})$$

Summary of Algorithm

Input: X a set of jobs, m

Output: minimal makespan of X

1. For each $S_1, S_2 \in \binom{X}{\leq m}$ do:
 - a) Compute recursively makespan of $X \cap \sigma(S_1, S_2)$
2. Find a minimal combination of these subschedules
3. Return found schedule

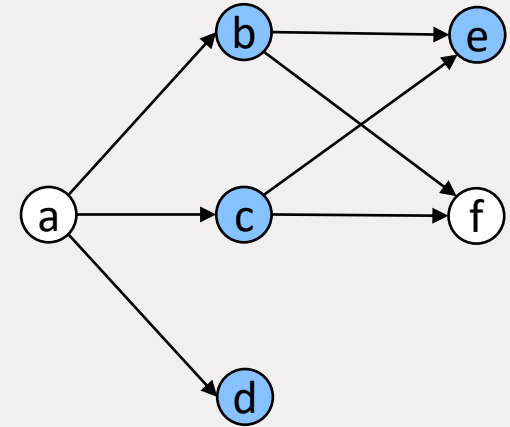
Simple analysis: $2^n \text{poly}(n)$ time...?

Only n^{2m} subproblems

Use of DP in $\approx n^{2m}$ time

Describing subschedules

Let σ' be a *subschedule*.



time

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|---|---|---|---|---|
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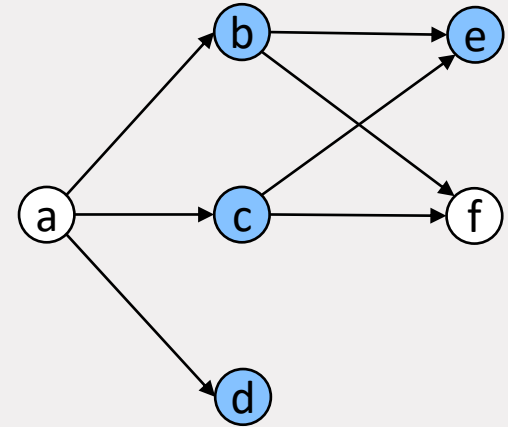
Jobs $V(\sigma')$ can be described as

$$V(\sigma') = \text{succ}[A] \cap \text{pred}[B]$$

where

A = minimal elements = sources of σ'

B = maximal elements = sinks of σ'



| | | time | | | |
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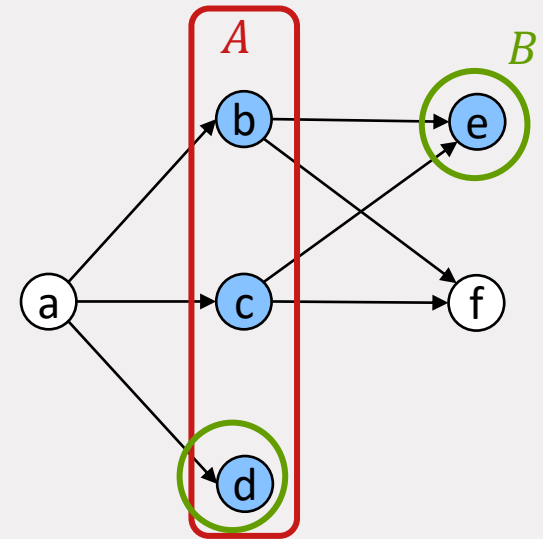
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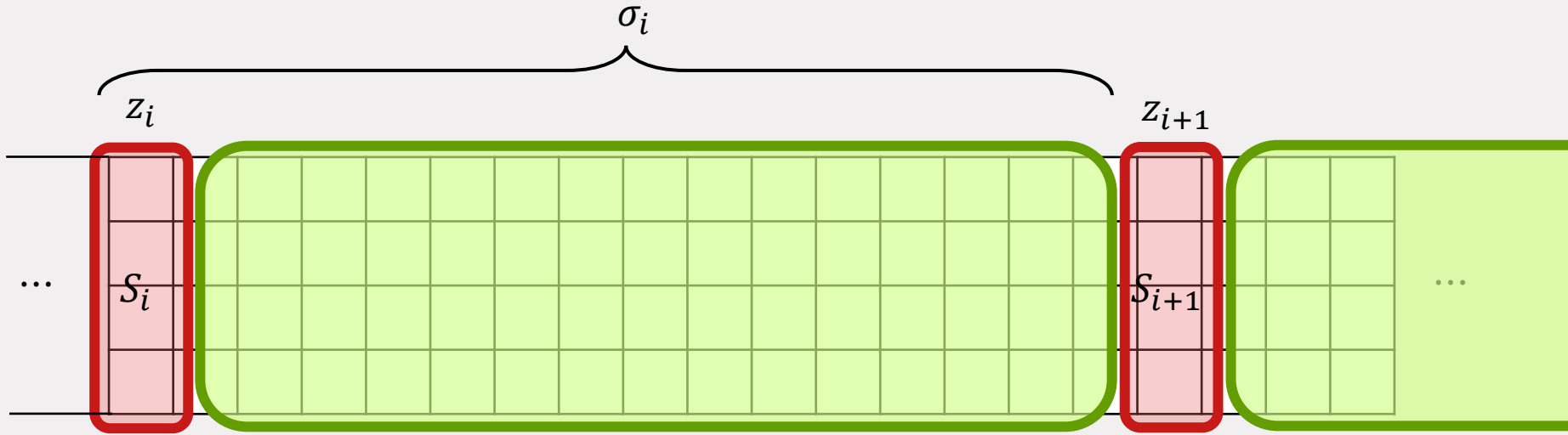
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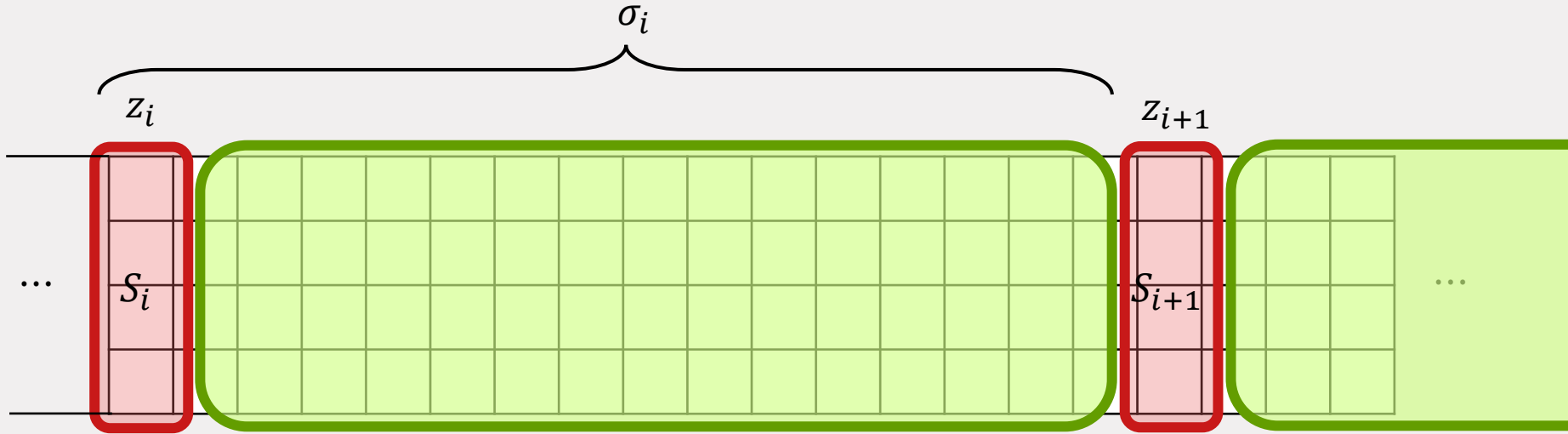


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Sink-Adjusted Schedule



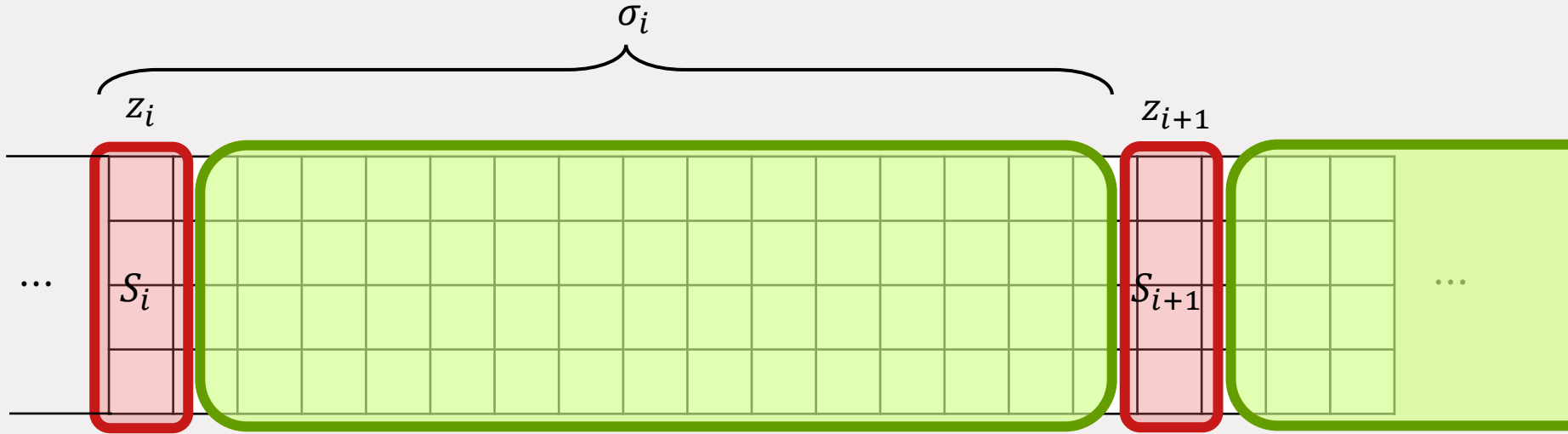
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Lemma:

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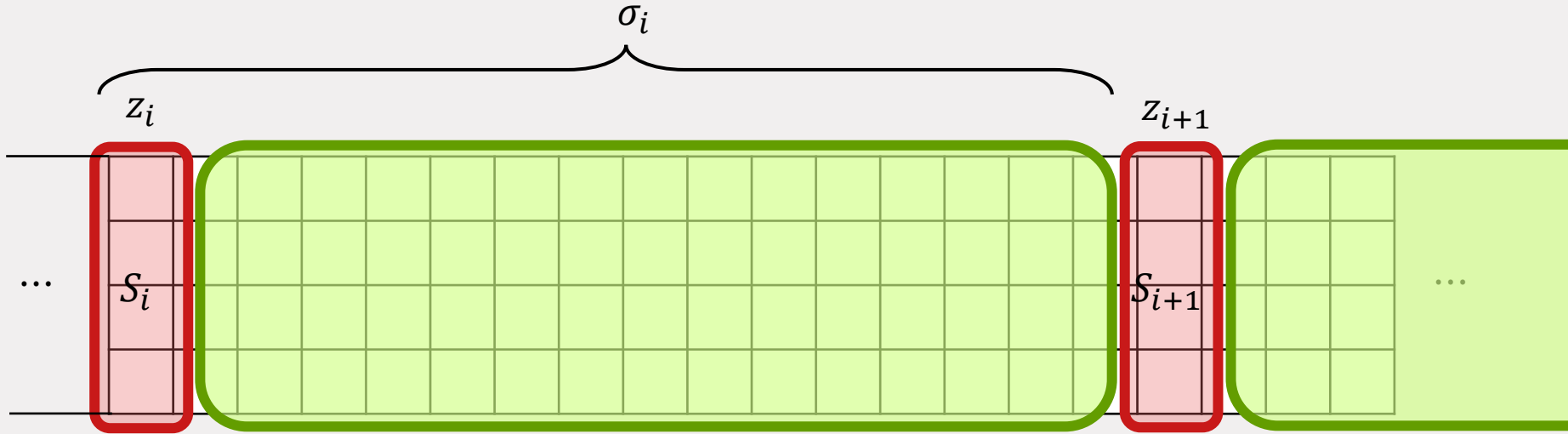
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n^m

win-win
strategy

Algorithm

Store subproblem results in **lookup table**:

$$T(A, B) := \min \text{makespan of } \text{succ}[A] \cap \text{pred}[B]$$

Use of **lookup table**, how many *different* subproblems?

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Case $|B| \leq \sqrt{n}$

⇒ only $\binom{n}{\sqrt{n}} = 2^{O(\sqrt{n} \cdot \log n)}$ **different** B 's

Case $|B| > \sqrt{n}$

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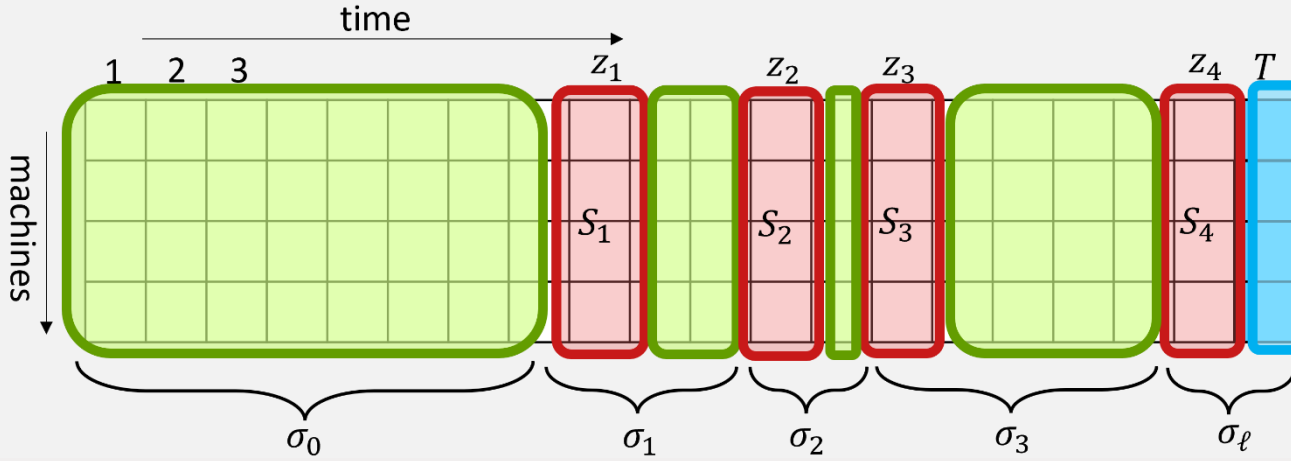
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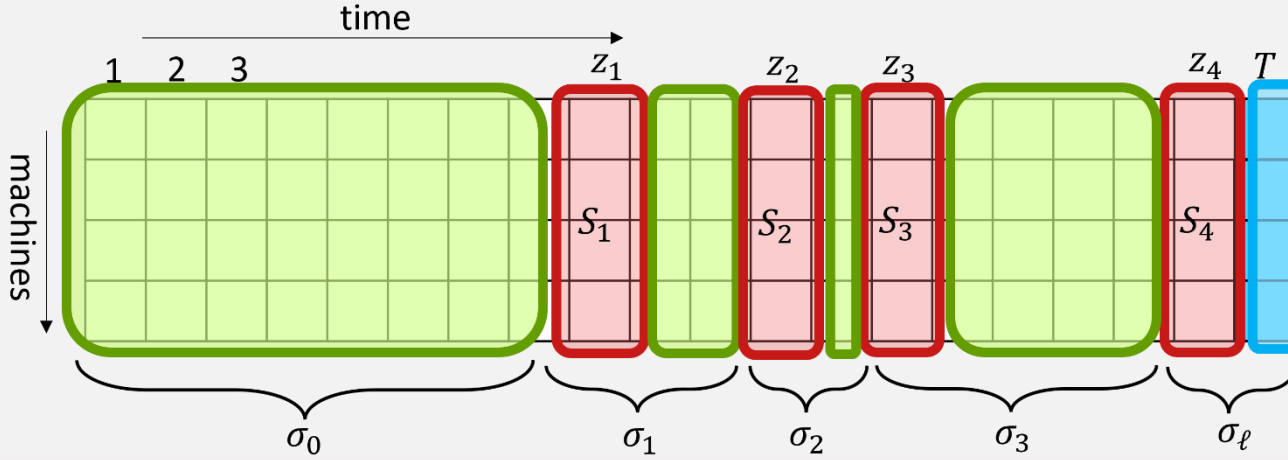
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Each σ_i contains at most m sinks (and σ_ℓ is extremely easy)

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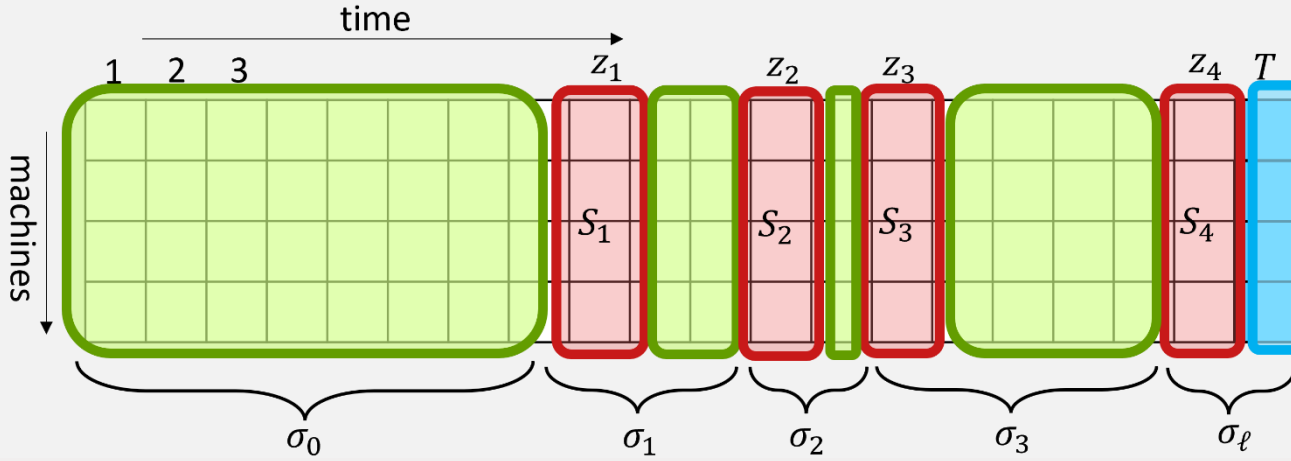
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$$V(\sigma_i) \approx \text{succ}[S_i] \setminus (\text{succ}[S_{i+1}] \cup \text{sinks}(\sigma)) \\ = \text{succ}[S_i] \cap \text{pred}(\text{sinks}(V(\sigma_i)))$$



Each σ_i contains at most m sinks (and σ_ℓ is extremely easy)

$$|\sigma_i| \leq n - \sqrt{n} + m$$

Case $|B| \leq \sqrt{n}$

\Rightarrow only $\binom{n}{\sqrt{n}} = 2^{O(\sqrt{n} \cdot \log n)}$ different B 's ✓

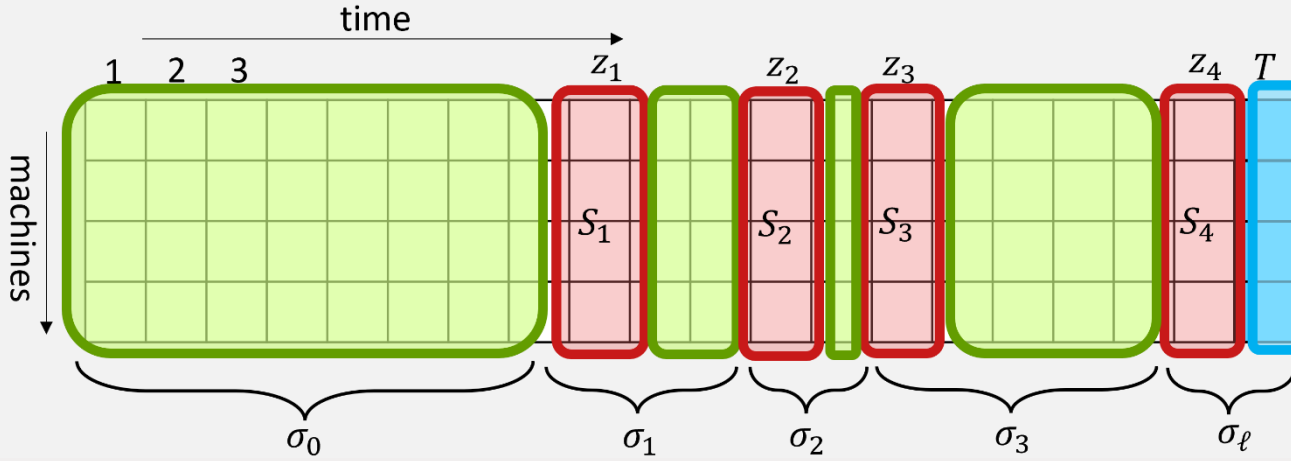
Case $|B| > \sqrt{n}$

In next step: make \sqrt{n} jobs progress!

Algorithm

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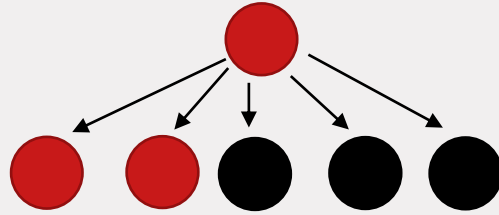
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Algorithm

$\leq n^{2m} = n^{O(1)}$
children



● = $|B| \leq \sqrt{n}$
● = $|B| > \sqrt{n}$

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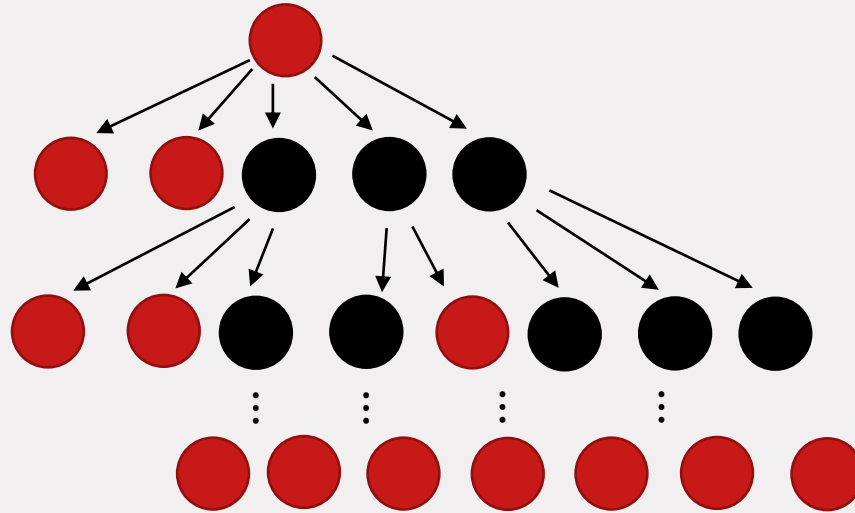
In next step: make \sqrt{n} jobs progress! ✓

Algorithm

$\leq n^{2m} = n^{O(1)}$
children

Height $\leq \sqrt{n}$

$\leq (n^{O(1)})^{\sqrt{n}}$
 $= 2^{O(\sqrt{n} \cdot \log n)}$
nodes in tree!



● = $|B| \leq \sqrt{n}$
● = $|B| > \sqrt{n}$

Case $|B| \leq \sqrt{n}$

\Rightarrow only $\binom{n}{\sqrt{n}} = 2^{O(\sqrt{n} \cdot \log n)}$ different B ✓

Case $|B| > \sqrt{n}$

In next step: make \sqrt{n} jobs progress! ✓

Algorithm

Store subproblem results in **lookup table**:

$$T(A, B) := \min \text{makespan of } \text{succ}[A] \cap \text{pred}[B]$$

Use of **lookup table**, how many *different* subproblems?

⇒ at most $2^{O(\sqrt{n} \cdot \log n)}$ trees with $2^{O(\sqrt{n} \cdot \log n)}$ nodes

⇒ $2^{O(\sqrt{n} \cdot \log n)}$ *different* subproblem!

| Case $ B \leq \sqrt{n}$ | Case $ B > \sqrt{n}$ |
|-----------------------------------------------------------------------------------------------|------------------------------------------------|
| ⇒ only $\binom{n}{\sqrt{n}} = 2^{O(\sqrt{n} \cdot \log n)}$ different B ✓ | In next step: make \sqrt{n} jobs progress! ✓ |

Summary of Algorithm

Input: A, B, m

Output: minimal makespan of $\text{succ}[A] \cap \text{pred}[B] = \text{Jobs}$

1. For each $S_1, S_2 \in \binom{\text{Jobs}}{\leq m}$ do:
 - a) Compute recursively makespan of $\text{Jobs} \cap (\text{succ}[S_1] \setminus (\text{succ}[S_2] \cup B))$
2. Find a minimal combination of these subschedules
3. Return found schedule

Conclusion

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Main result:

$Pm|prec, p_j = 1|C_{\max}$ in $2^{O(\sqrt{n} \cdot \log n)}$ time.

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Key idea's:

- New [decomposition](#) of schedules
- Use of [look-up table](#)
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Future Research:

$P3|prec, p_j = 1|C_{\max}$ in [quasi-polynomial](#) time?

Conclusion

Main result:

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Key idea's:

- New **decomposition** of schedules
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Future Research:

$P_3 | prec, p_j = 1 | C_{\max}$ in **quasi-polynomial** time?

*Thanks for your
attention!*