

A Heuristic Branch and Bound for the EURO/ROADEF 2018 challenge

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G-SCOP

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Methodology

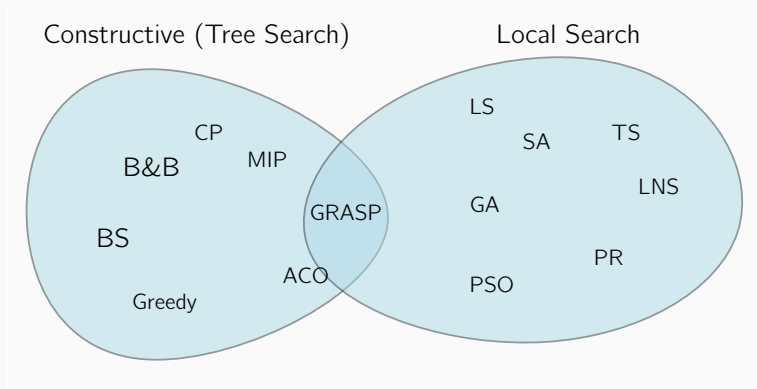
Branching Scheme

Search Strategy

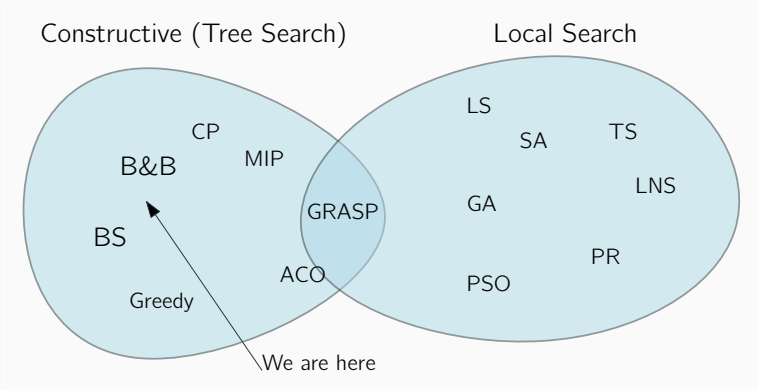
Results

Methodology

Constructive vs Local Search



Constructive vs Local Search



Our method integrates parts of *Branch and bounds* and *Beam Search*

Anatomy of a Tree Search

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 - root definition
 - children generation
 - bounds
 - *etc.*

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We developed our algorithm using this principle.

- the **Branching Scheme** (*i.e. problem specific*)
- the Search strategy (*i.e. generic*)

Branching Scheme

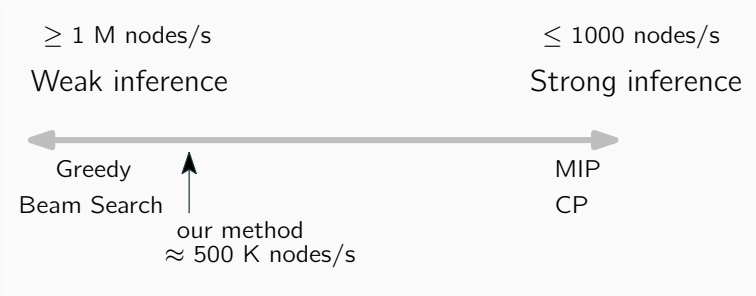
Node inference trade-off



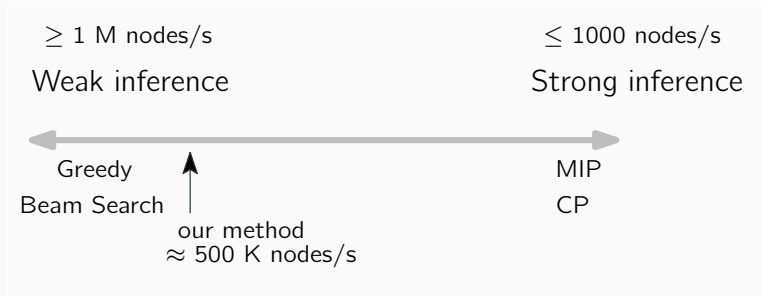
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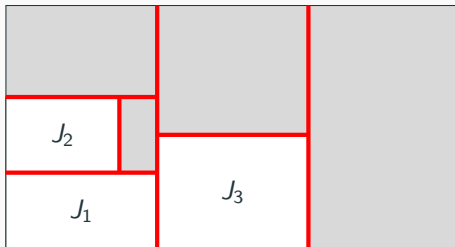


Node inference trade-off

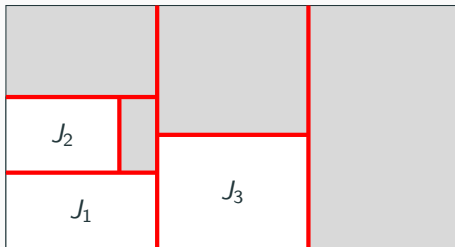


- We integrate quick bounds, symmetry breaking, dominance checking
- The idea of integrating Branch and Bound parts into Beam Searches can be found in [STDC18]

Packing in the bottom left corner



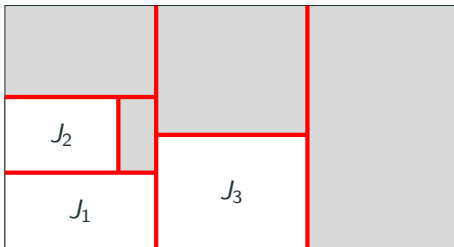
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We prove that it is optimal if:

- guillotine and defects and precedence only
- guillotine and min waste only

Packing in the bottom left corner



We prove that it is optimal if:

- guillotine and defects and precedence only
- guillotine and min waste only

We prove it is not if:

- guillotine and min waste and precedences
- guillotine and min waste and defects

Not dominant in the challenge

Since guillotine and min waste and precedences and defects constraints.



Good news - It still works very well !

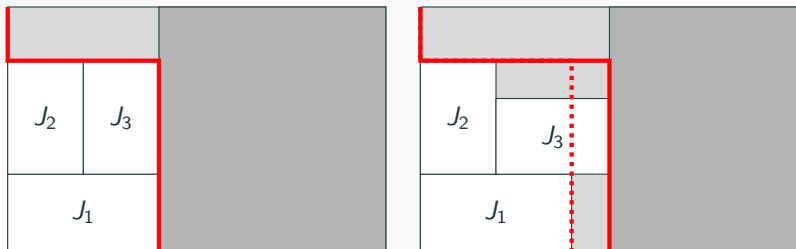
We only need good solutions, so we make a heuristic Branch and Bound.



How to construct children

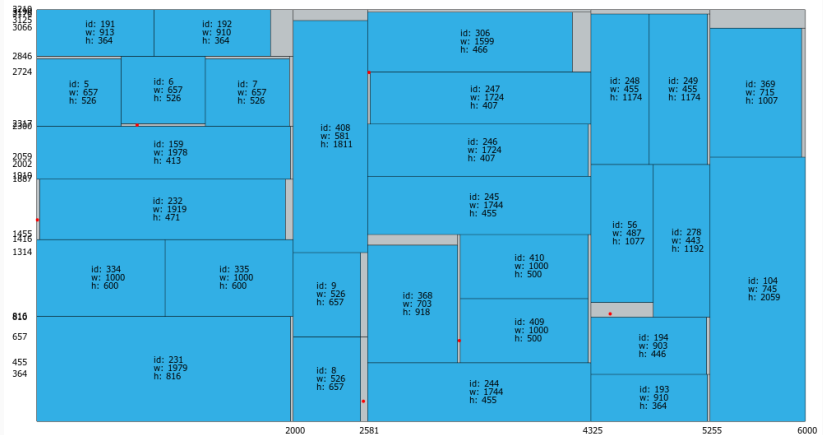
- Root node: empty solution
- Children: all possible items in all possible positions (*i.e.* new plate, new 1-cut, new 2-cut, new 3-cut or new 4-cut, rotations, defect avoidance)

Pseudo dominance



Symmetry breaking

- Symmetry breaking strategy: for two consecutive blocks, the one with the smallest minimum item id comes before.

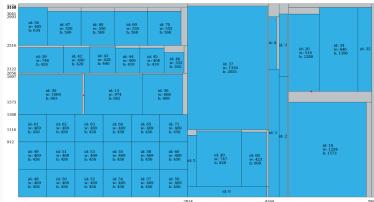


Node Goodness measure

Waste accumulated so far

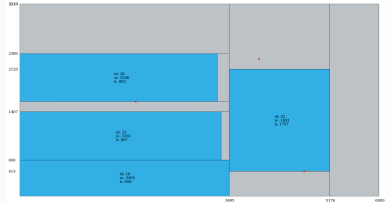
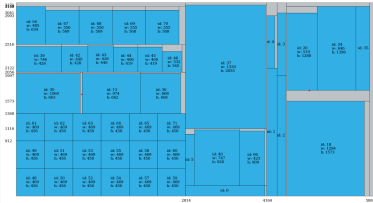
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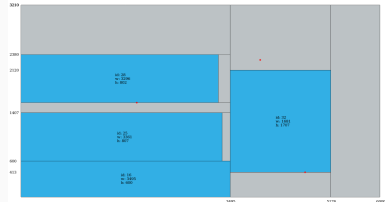
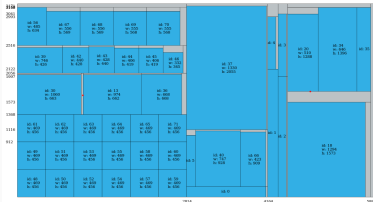
Node Goodness measure

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Problem with waste:

- Small items at the beginning and big items at the end

A better node goodness measure

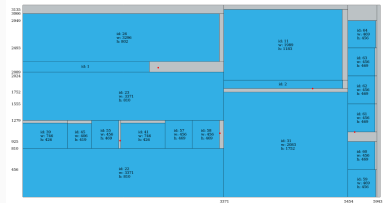
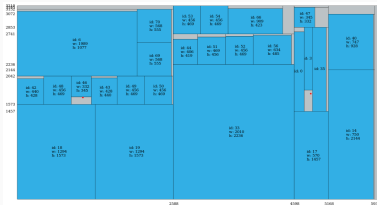
waste percentage

An even better node goodness measure

$$\frac{\text{waste}}{\text{total area} \cdot \text{mean area}}$$

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Search Strategy

- the Branching Scheme (*i.e. problem specific*)
- **the Search strategy** (*i.e. generic , DFS, Best First, Beam Search, ...*)

Inspired from *Beam Search* and *SMA**

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- Best First strategy
- Delete some bad nodes if too many at the same time
- If finished, Restart with a bigger node limit D ($D_{n+1} \leftarrow D_n \times 2$)

Inspired from *Beam Search* and *SMA**

- Best First strategy
 - Delete some bad nodes if too many at the same time
 - If finished, Restart with a bigger node limit D ($D_{n+1} \leftarrow D_n \times 2$)
-
- at the beginning ($D = 1$), it behaves like a greedy algorithm
 - at the end ($D \approx \infty$), it behaves like a Best First Search

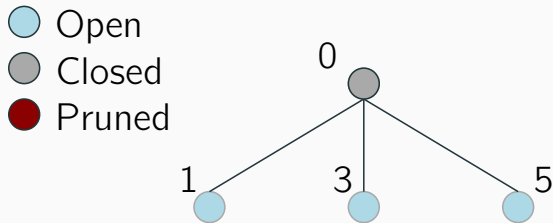
MBA* - an example

- Open
- Closed
- Pruned

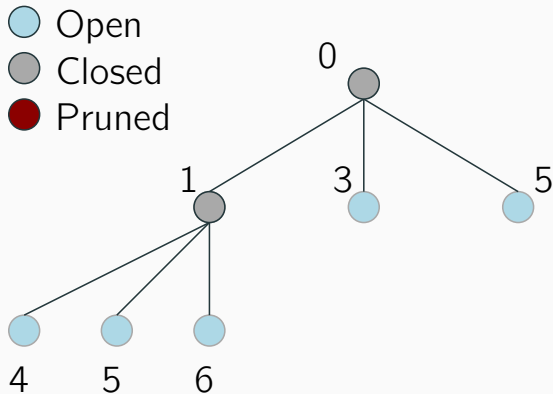
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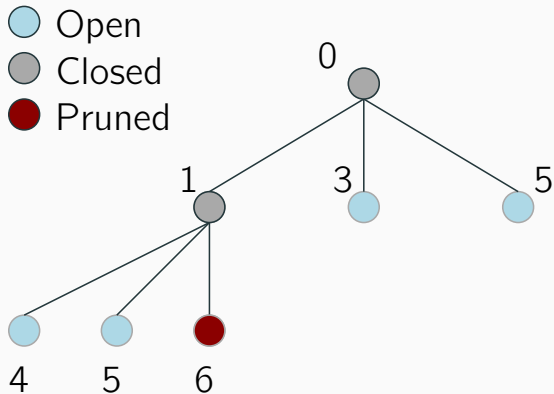
MBA* - an example



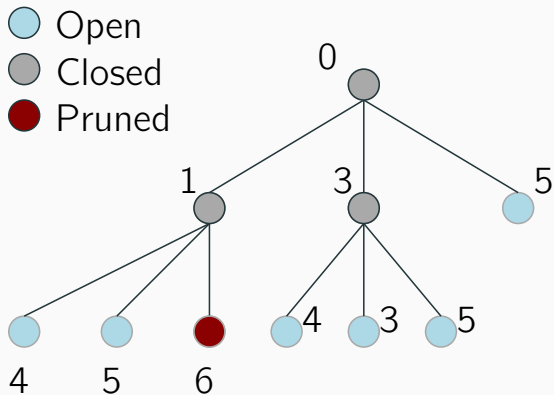
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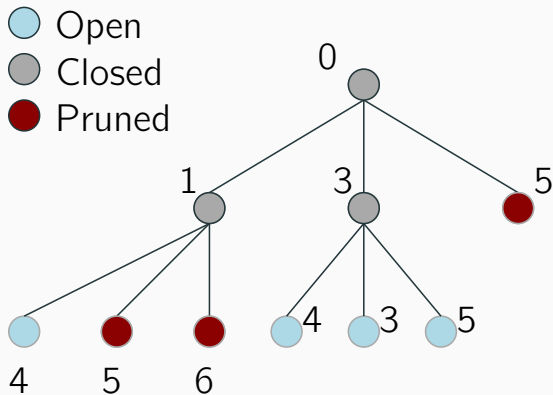
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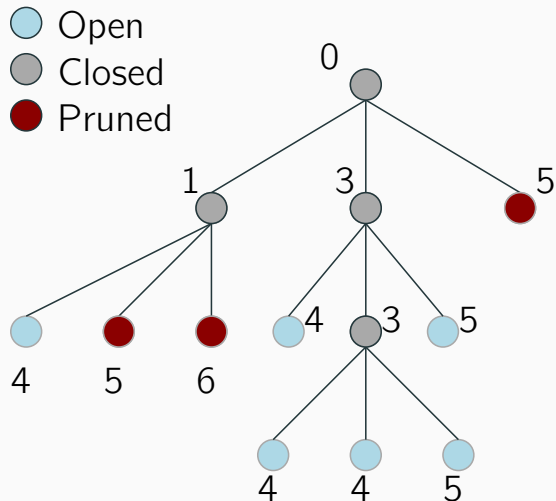
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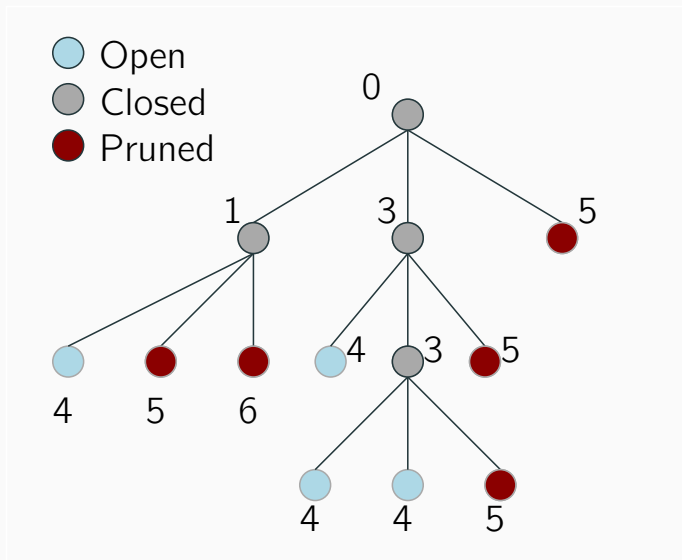
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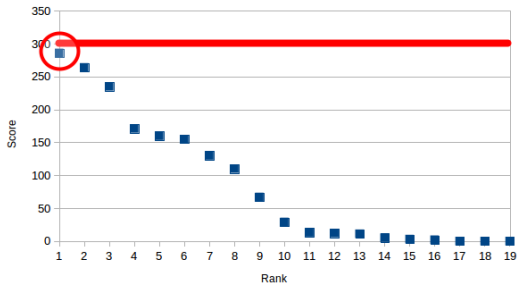
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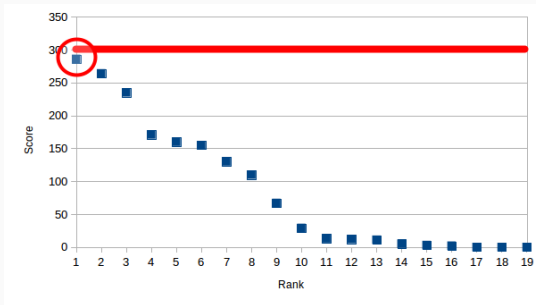


MBA* - an example

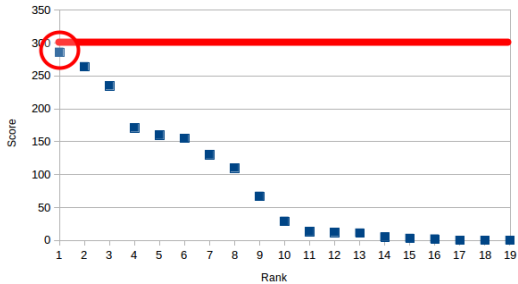


Results

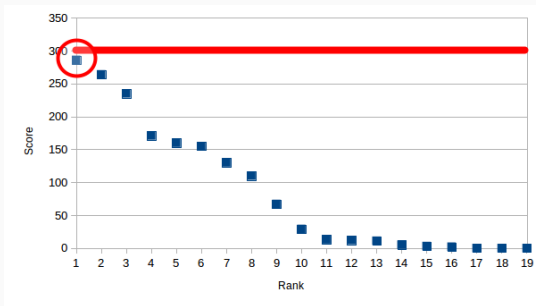




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- Total waste 2nd team: 506M
- Total waste: 493M (13M less)



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- Total waste 2nd team: 506M
- Total waste: 493M (13M less)
- Total waste new version: 469M (24M less than our submission)

Conclusions

- Simple and effective algorithm
- Tree searches can be competitive with other methods
- Decomposing the algorithm helps to identify good (and bad) parts

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- Simple and effective algorithm
- Tree searches can be competitive with other methods
- Decomposing the algorithm helps to identify good (and bad) parts
- We tried
 - several search strategies (DFS, Beam Search, LDS, and MBA*)
 - several guides
- Chose best combination

Questions or remarks ?



Lei Shang, Vincent T'Kindt, and Federico Della Croce.

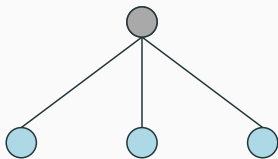
The memorization paradigm: Branch & memorize algorithms for the efficient solution of sequencing problems.

2018.

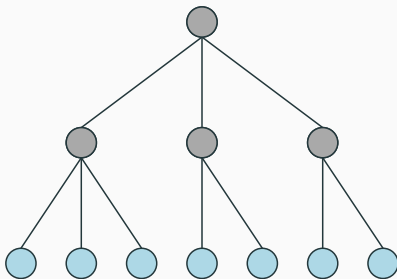
Breadth First Search



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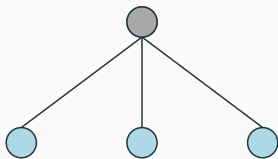
Breadth First Search



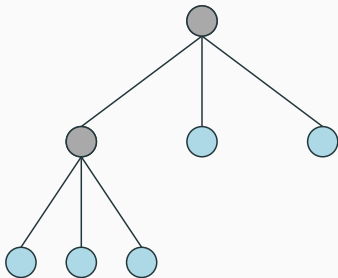
Depth First Search



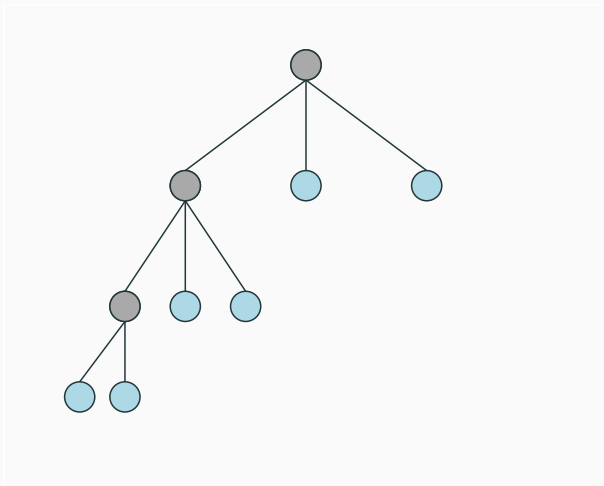
Depth First Search



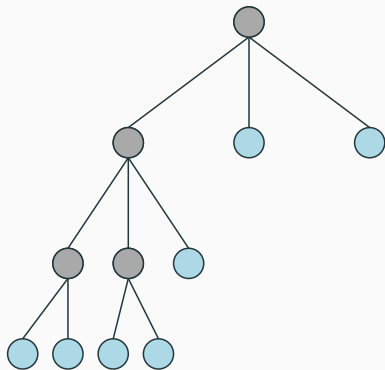
Depth First Search



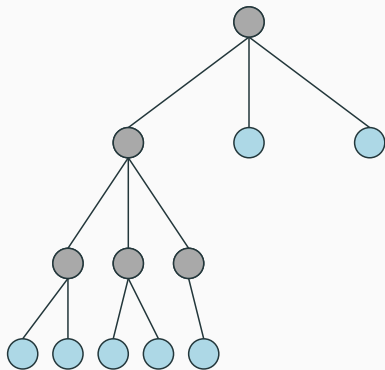
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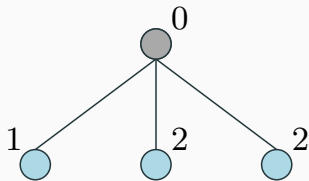


Beam Search ($D = 3$)

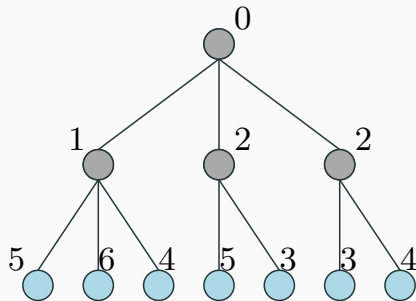


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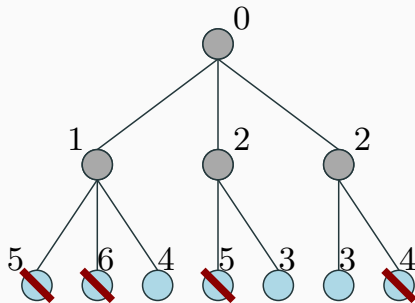
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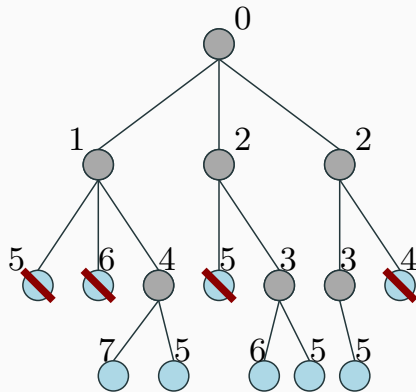
Beam Search ($D = 3$)



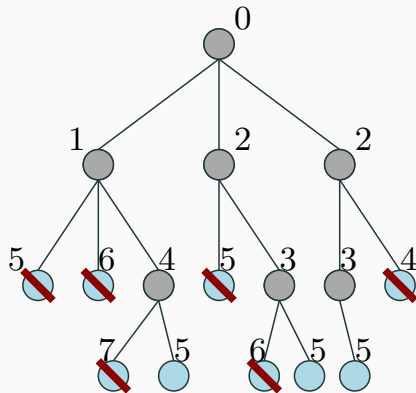
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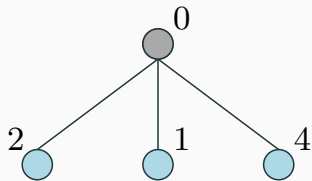


Best First

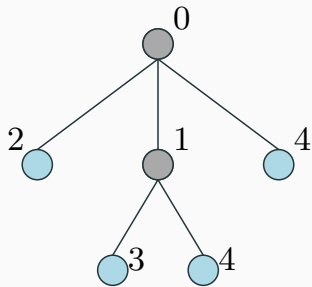


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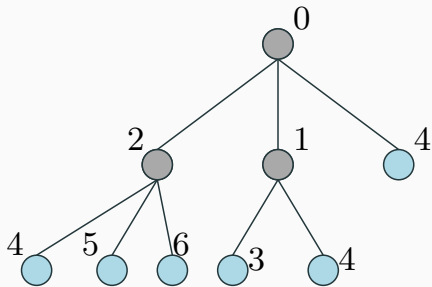
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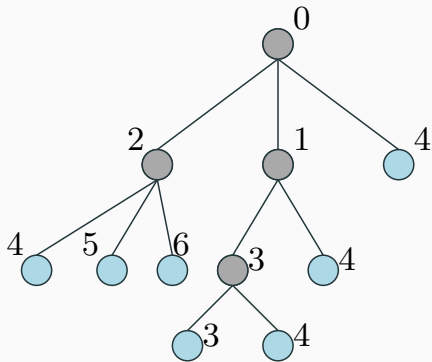
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