

# Branch and bound - Approximations

## Advanced Algorithms

Master CPS2/DSC/MLDM

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## Question 1 **Task Assignment**

# Q1 cost and search tree

Similarly as in the lecture, imagine that we have currently affected a task for  $k$  agents out of  $n$ , for a corresponding solution vector  $\mathbf{v}$ , we can define the following quantities:

- $g^*(\mathbf{v}) = \sum_{i=1}^k c[i, \mathbf{v}[i]]$   
the sum of the costs of the tasks affected to the first  $k$  agents

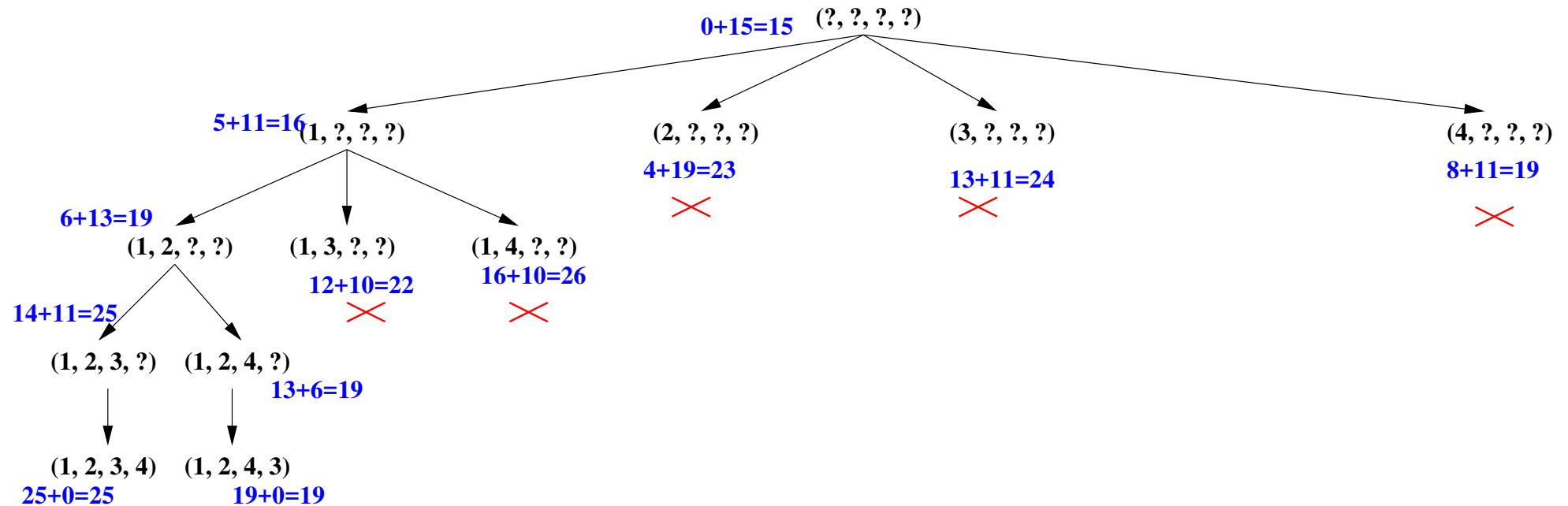


$$f(\mathbf{v}) = \sum_{i=k+1}^n \min_{\substack{1 \leq j \leq n \\ j \neq \mathbf{v}[l] \text{ for } 1 \leq l \leq k}} c[i, j]$$

we take the minimum for each agent (line) among the available quantities

# Q1: Search Tree

Here is the new search tree obtained:



Similarly, each node defines a (partial) solution written in black. We

associate the quantity  $g^* + h = f$  in blue next (or below) to each node (first value before = is  $g^*$ , second is  $h$ ). The red crosses indicate when the search can be cut. You can see that more branches can cut when a finer evaluation function is used.

## Q2: example

Consider the following cost table with 2 agents and 2 functions:

Agent \ Task	1	2
1	2	1
2	8	2

The strategy implies to assign task 2 to agent 1, then we need to associate task 1 to agent 2. The final cost is  $8+1=9$ .

However, the optimal solution consists in assigning task 1 to agent 1 and task 2 to agent 2, leading to a final cost of  $2+2=4$ .

This strategy is clearly not optimal. In particular, with this example you can see that  $h = 8$  because the other values are removed leading to an evaluation function that can be higher than the optimal result.

**$\Rightarrow$  your evaluation function must always provide a smaller result than the optimal solution.**