



## Optimization (Exercices)

## 1 Solving COP

Let P be the following COP instance:

```
from pycsp3 import *

x = Var(0, 1, 2)
y = Var(0, 1, 2)
z = Var(0, 1, 2)

satisfy(
    x > y,
    y != z
)

maximize(
    x + y + z
)
```

Assuming that AC is maintained during search (while considering a non-binary branching scheme), and lexico is used as variable ordering heuristic, draw the search tree that is built when solving P:

- with values selected in increasing order;
- with values selected in decreasing order.

## 2 Depth-First Branch and Bound

Depth-first branch and bound (DFBB) performs a depth-first traversal of the search tree, when solving an optimization (minimization) problem. We have that:

- at any time, the upper bound ub is known, which is the maximum cost (-1) that we are willing to accept;
- at any node corresponding to an instantiation I, the lower bound lb(I) is an underestimation of the cost of any complete assignment (extending I) below that node.

When  $ub \leq lb(I)$ , the subtree rooted by I can be pruned because it contains no solution with a cost lower than ub. If it finds a complete assignment (i.e., we have |I| = n, with n being the number of variables in the problem) with a cost lower than ub, this cost becomes the new ub; after exhausting the tree, ub is the optimal cost. The temporal complexity of DFBB is  $O(d^n)$ , while its spatial complexity is O(nd).

Give the pseudo-code of a recursive algorithm that implements DFBB. The initial call is  $dfbb(\emptyset)$ .

## 3 Solving MaxCSP

We assume binary constraints only, and we are interested in finding the instantiation that satisfies the higher number of constraints (MaxCSP): the cost of a violated constraint is 1 whereas it is 0 if the constraint is satisfied. We propose to use DFBB for solving MaxCSP, with two different ways of computing lb(I) where I is the current instantiation.

- $lb_1(I)$  is the sum of the costs of all constraints covered by the current instantiation (i.e., the constraints only involving fixed variables)
- $lb_2(I)$  is  $lb_1(I)$  plus the sum of the minimum possible costs of all constraints almost covered by the current instantiation (i.e., the constraints involving a fixed variable and an unfixed variable)

Let P be the following CN:

```
• vars(P) = \{shoes, shirt, slacks\} with

- dom(shoes) = \{cordovans, sneakers\}
- dom(slacks) = \{jeans, dress, skirt\}
- dom(shirt) = \{green, white\}
• ctrs(P) = \{
\langle shoes, slacks \rangle \in \{(cordovans, skirt), (sneakers, jeans)\},
\langle shoes, shirt \rangle \in \{(cordovans, white)\},
\langle slacks, shirt \rangle \in \{(jeans, white), (dress, white), (skirt, green)\}
}
```

While considering the static variable ordering shoes < slacks < shirt and a lexicographic value ordering,

- 1. draw the search tree built by DFBB for MaxCSP solving of P, when using  $lb_1$ ,
- 2. draw the search tree built by DFBB for MaxCSP solving of P, when using  $lb_2$ .