Report

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Introduction

This is the Report for the first module of the Combinatorial Decision Making and Optimization exam for July 2021.

1 Setting

Since the project specification required to take data from a text file - and also to write the results on it - we wrapped Python code around the MiniZinc model. We used Python 3.6 mainly due to compatibility issues, the file Module1.py uses the os library to read the data from the standard instance txt files, then runs the external MiniZinc model (CP_base.mzn) on the gathered input. Finally the results of the CP model are written in an output text file as specified in the specification. The Python code also produces a visual representation of the model's solution with matplotlib, this was used mainly for developing purposes as advised.

2 7 points

As advised in the paper describing the project, we are going to described the seven points that brought us to the final implementation of the Constraint Programming model.

2.1 Point 1: Variables, Main constraints and objective function

Starting from the variables, we take the values from the input file(one of the instance) and we save them in the corresponding variables. In paricular width correspond to the first parameter of the txt file, n_rets identifies the number of rectangles to place and sizes is an array containing the sizes of each rectangle.

```
int: width;
int: n_rets;
set of int: RETS = 1..n_rets;
array[RETS, 1..2] of int: sizes;
```

The array *positions* is a variable use to store the starting position of each rectangle inside the optimal solution found.

```
array[RETS, 1..2] of var 0..sum(
   [sizes[i,2]| i in RETS]): positions;
```

We then proceed by defining the main constraints.

A predicate has been defined to help us avoid that two rectangle could be place one on top of the other.

These two constrains are our first try to find an optimal solution.

The objective functions consist in the minimization of the variable that indicates the height of our workspace, in this case l.

```
var 0..sum([sizes[i,2] | i in RETS]): 1 =
   max([positions[i,2] + sizes[i,2] | i in RETS]);
solve minimize 1;
```

2.2 Point 2: implied constraints

These constraints guarantee that any rectangle is not going to be wider than the width in input and it cannot be placed outside the boundary imposed (totally or partially).

```
constraint forall(i in RETS)
        (sizes[i,1] < width);

constraint forall(i in RETS)
        (positions[i,1]+sizes[i,1] <= width);</pre>
```

Given that the height is defined as the highest value reach by any rectangles, it is implied that no rectangle can exceed the value set.

2.3 Results part 1

Even though the results that we found so far were correct, the solutions were not optimal. With the time limit set to 5 minutes, without further optimization, we were not able to solve instances after the 10^{th} .

2.4 Point 3: global constraints

The first global constraint that we added was the *cumulative* constraint. Where size1 and size2 correspond respectively to the base and height size of each rectangle, which summed doesn't have to be bigger than the width constraint.

Another global constraint that we tried to introduce is diffn, by replacing the already present one for checking that two rectangles would not overlap in one of the two dimensions. Even though the results for the instances we were able to solve remained the same, the time taken to find the solution was increased. We rolled back to the previous constraints.

```
constraint diffn([positions[i,1] | i in RETS],
        [positions[i,2] | i in RETS],
        size1,
        size2);
```