

# A MULTI-OBJECTIVE BRANCH-AND-CUT ALGORITHM FOR THE MULTI-MODAL TRAVELING SALESMAN PROBLEM

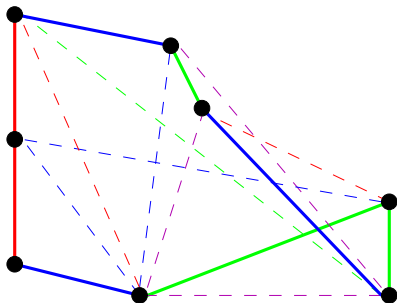
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## OUTLINES

- ▶ The multi-modal traveling salesman problem
- ▶ Multi-objective optimization
- ▶ A branch-and-cut algorithm
- ▶ Computational results
- ▶ Conclusions and perspectives

## THE MULTI-MODAL TRAVELING SALESMAN PROBLEM



### Data:

$G = (V, E)$  : an undirected valuated graph

$C$  is a set of colors

Each  $e \in E$  has a color  $k \in C$

### Goal:

Find a Hamiltonian cycle

Two objectives:

1. Minimize the total length of the cycle
2. Minimize the number of colors appearing on the cycle

## INTEGER PROGRAM

Variables

$$x_e = \begin{cases} 1 & \text{if } e \in E \text{ is used,} \\ 0 & \text{otherwise.} \end{cases}$$
$$u_k = \begin{cases} 1 & \text{if } k \in C \text{ is used,} \\ 0 & \text{otherwise.} \end{cases}$$

Constants and notations  $\forall e \in E, \delta(e) = k \in C$  the color of  $e$

$$\forall k \in C, \zeta(k) = \{e \in E \mid \delta(e) = k\}$$

$$\forall S \subset V, \omega(S) = \{e = (i, j) \in E \mid i \in S \text{ and } j \in V \setminus S\}$$

## INTEGER PROGRAM

Objective functions

$$\min \sum_{e \in E} c_e x_e$$

$$\min \sum_{k \in C} u_k$$

Constraints

$$\sum_{e \in \omega(\{i\})} x_e = 2 \quad \forall i \in V$$

$$\sum_{e \in \omega(S)} x_e \geq 2 \quad \forall S \subset V, 3 \leq |S| \leq |V| - 3$$

$$x_e \leq u_{\delta(e)} \quad \forall e \in E$$

$$x_e \in \{0, 1\} \quad \forall e \in E$$

$$u_k \in \{0, 1\} \quad \forall k \in C$$

## VALID CONSTRAINTS

$$u_k \leq \sum_{e \in \zeta(k)} x_e \quad \forall k \in C$$

$$\sum_{k \in C} \gamma_i^k u_k \geq 2 \quad \forall i \in V$$

$$\sum_{k \in C} \lambda_k(S) u_k \geq 2 \quad \forall S \in V, 3 \leq |S| \leq |V| - 3$$

with

$$\gamma_i^k = \begin{cases} 0 & \text{if } \nexists e \in \omega(\{i\}), e \in \zeta(k), \\ 1 & \text{if } \exists! e \in \omega(\{i\}), e \in \zeta(k), \\ 2 & \text{otherwise.} \end{cases} \quad \lambda_k(S) = \begin{cases} 0 & \text{if } \nexists e \in \omega(S), e \in \zeta(k), \\ 1 & \text{if } \exists! e \in \omega(S), e \in \zeta(k), \\ 2 & \text{otherwise.} \end{cases}$$

## STATE-OF-THE-ART

Minimum labelling hamiltonian cycle problem (Tabu search) [Cerulli, Dell'Olmo, Gentili, Raiconi, 2006]

Colorful traveling salesman problem (heuristic, GA) [Xiong, Golden, Wasil, 2007]

Traveling salesman problem with labels (approximation algorithm) [Gourvès, Monnot, Telelis, 2008]

Minimum labelling spanning tree problem

## MULTI-OBJECTIVE OPTIMIZATION PROBLEM

$$(PMO) = \begin{cases} \min F(x) = (f_1(x), f_2(x), \dots, f_n(x)) \\ \text{s.t. } x \in \Omega \end{cases}$$

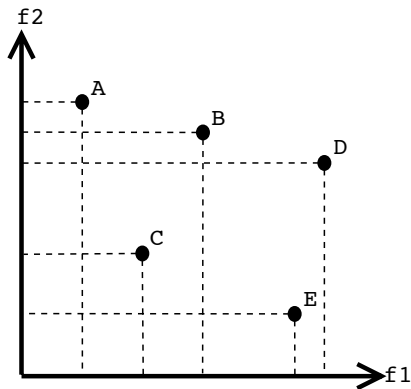
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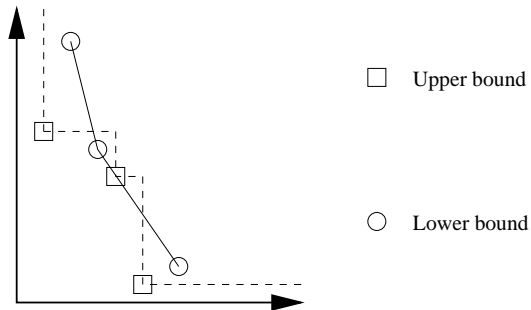
- ▶  $n \geq 2$  : number of objectives
- ▶  $F = (f_1, f_2, \dots, f_n)$  : vector of functions to optimize
- ▶  $\Omega \subseteq \mathbb{R}^m$  : set of feasible solutions
- ▶  $x = (x_1, x_2, \dots, x_m) \in \Omega$  : a feasible solution
- ▶  $\mathcal{Y} = F(\Omega)$  : objective space
- ▶  $y = (y_1, y_2, \dots, y_n) \in \mathcal{Y}$  avec  $y_i = f_i(x)$  : a point in the objective space



## PARETO DOMINANCE RELATION

A solution  $x$  dominates ( $\preceq$ ) a solution  $y$  if and only if  
 $\forall i \in \{1, \dots, n\}, f_i(x) \leq f_i(y)$  and  $\exists i \in \{1, \dots, n\}$  such that  $f_i(x) < f_i(y)$ .





Problem: it does not work if the aggregated problem is NP-hard.

**Solution: linear programming**

## A MULTI-OBJECTIVE BRANCH-AND-CUT ALGORITHM

### STEP 1 (Root of the tree)

Generate an initial upper bound  $ub$

Define a first sub-problem

Insert the sub-problem in a list  $L$

### STEP 2 (Stopping criterion)

If  $L = \emptyset$  then STOP, else choose a sub-problem from  $L$  and remove it from  $L$

### STEP 3 (Sub-problem solution)

Solve the sub-problem to obtain the lower bound  $lb$

### STEP 4 (Constraint generation)

If some integer solutions have been found, try to insert them in  $ub$

**if**  $ub \preceq lb$  **then**

    Go to STEP 2.

**else**

**if** violated constraints are identified **then**

        Add them to the model and go to STEP 3.

**else**

        Go to STEP 5.

**end if**

**end if**

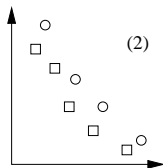
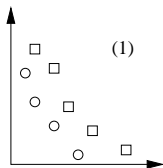
### STEP 5 (Branching)

Branch on variable and introduce 2 new sub-problems in  $L$ . Go to STEP 2.

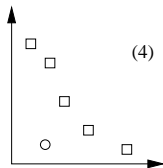
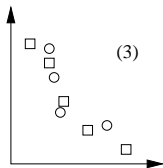
## ADAPTATIONS TO A MULTI-OBJECTIVE PROBLEM

**Upper bound** = set of non-dominated solutions found during the search

**Lower bound** = set of non-dominated points in the objective space such that all feasible solutions are dominated by these points



□ Upper bound



○ Lower bound

## COMPUTATION OF THE LOWER BOUND

Initial sub-problem :

$$\min \quad \sum_{e \in E} c_e x_e$$

$$\min \quad \sum_{k \in C} u_k$$

$$\sum_{e \in \omega(\{i\})} x_e = 2 \quad \forall i \in V$$

$$x_e \leq u_{\delta(e)} \quad \forall e \in E$$

$$u_k \leq \sum_{e \in \zeta(k)} x_e \quad \forall k \in C$$

$$\sum_{k \in C} \gamma_i^k u_k \geq 2 \quad \forall i \in V$$

$$0 \leq x_e \leq 1 \quad \forall e \in E$$

$$0 \leq u_k \leq 1 \quad \forall k \in C$$

## COMPUTATION OF THE LOWER BOUND

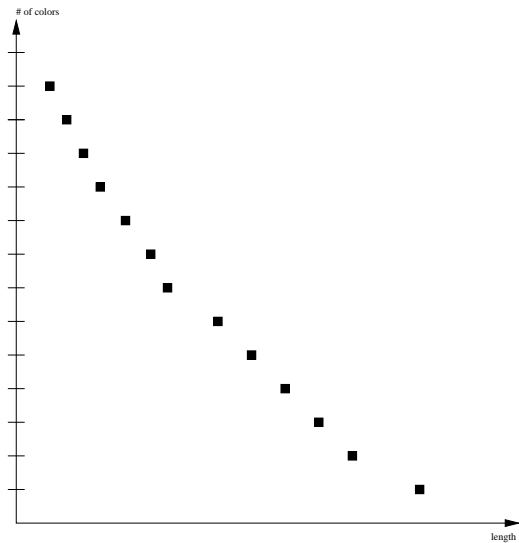
Solve the following problem for different values of  $\epsilon$

$$\begin{aligned} \min \quad & \sum_{e \in E} c_e x_e + m \sum_{k \in C} u_k \\ & \sum_{e \in \omega(\{i\})} x_e = 2 && \forall i \in V \\ & x_e \leq u_{\delta(e)} && \forall e \in E \\ & u_k \leq \sum_{e \in \zeta(k)} x_e && \forall k \in C \\ & \sum_{k \in C} \gamma_i^k u_k \geq 2 && \forall i \in V \\ & \sum_{k \in C} u_k \leq \epsilon \\ & 0 \leq x_e \leq 1 && \forall e \in E \\ & 0 \leq u_k \leq 1 && \forall k \in C \end{aligned}$$

After founding non-dominated solution for a given  $\epsilon$ , identify violated constraints and add them

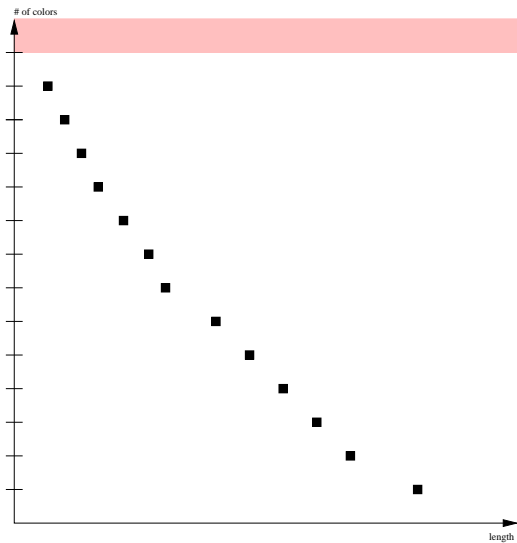
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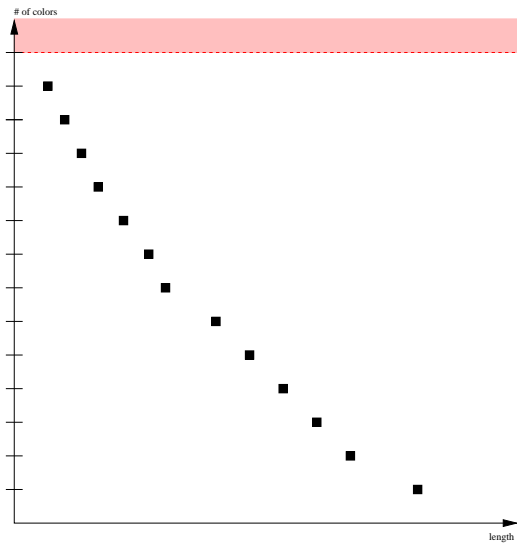




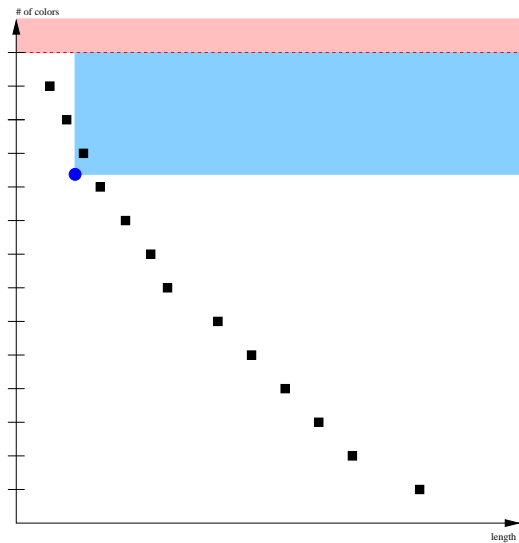
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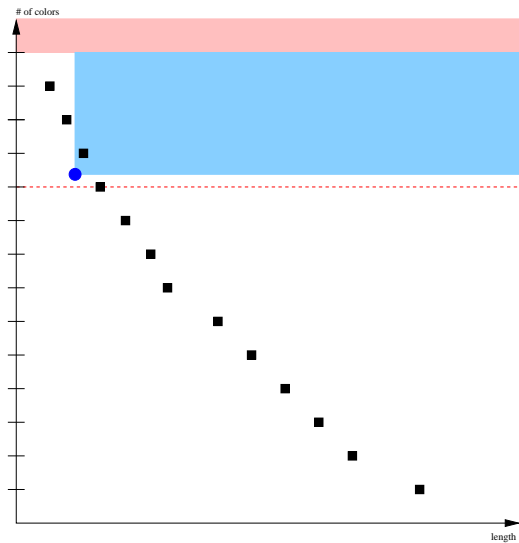
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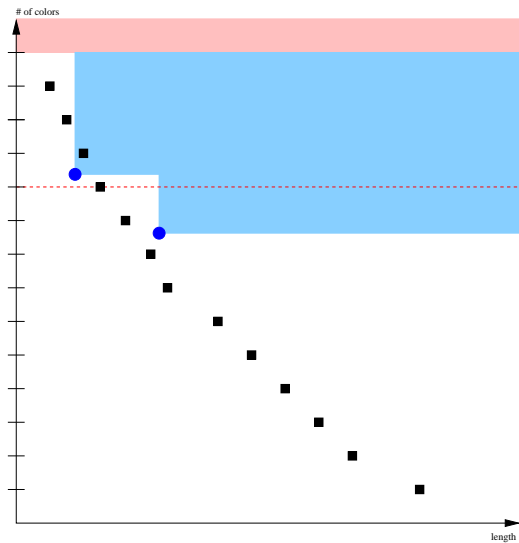
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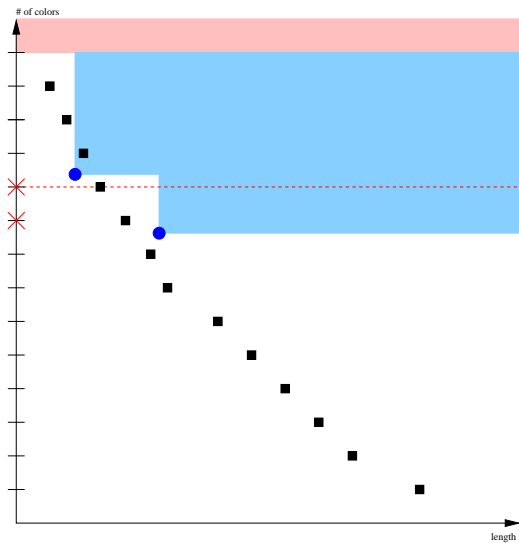
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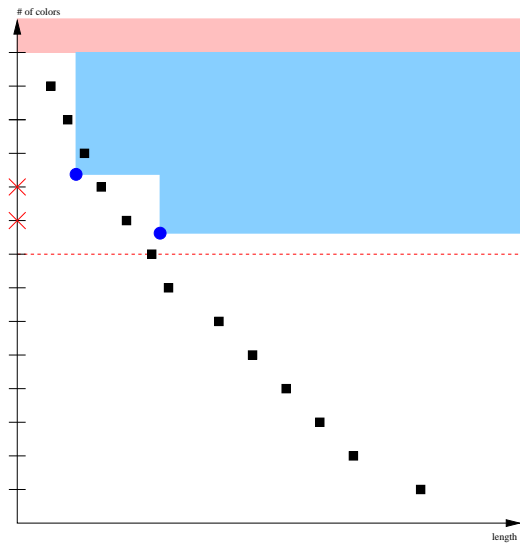
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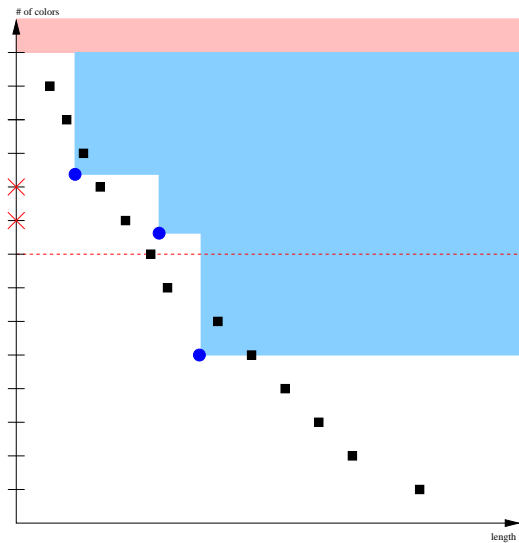
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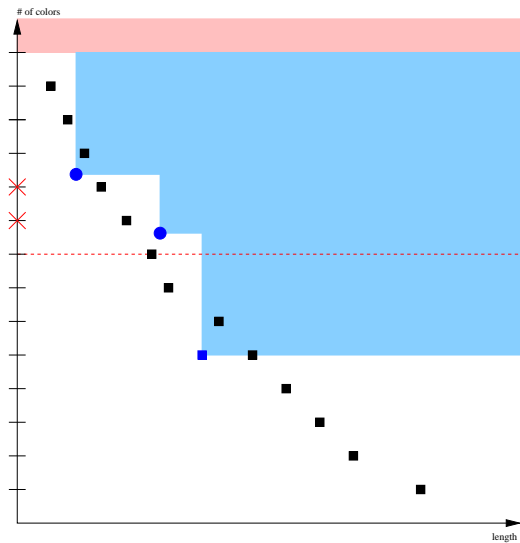


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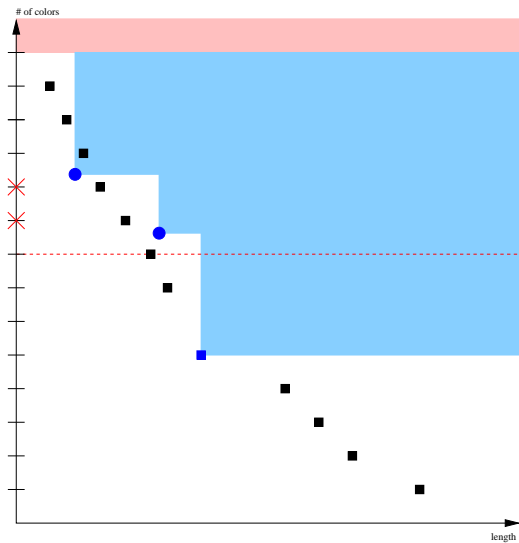




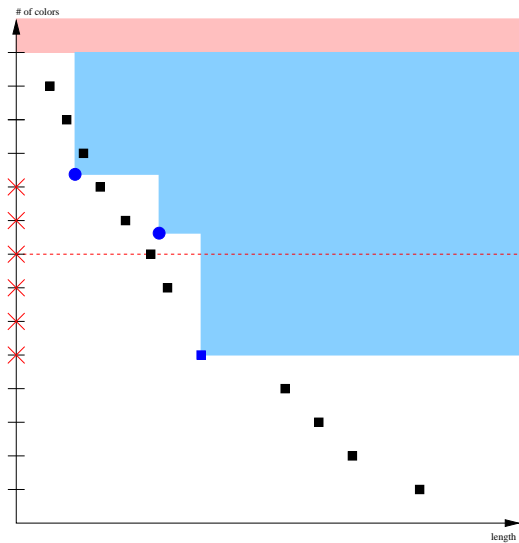
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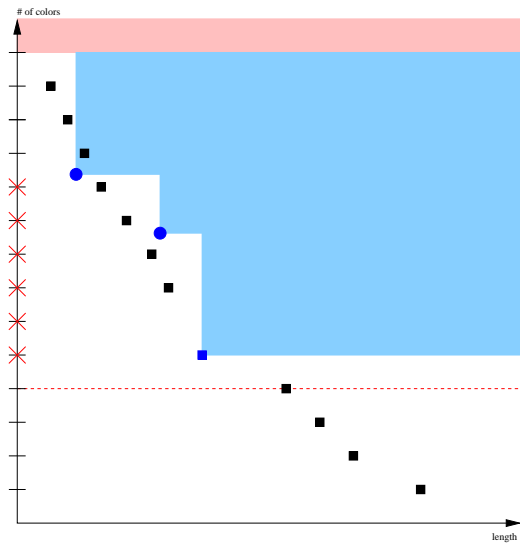
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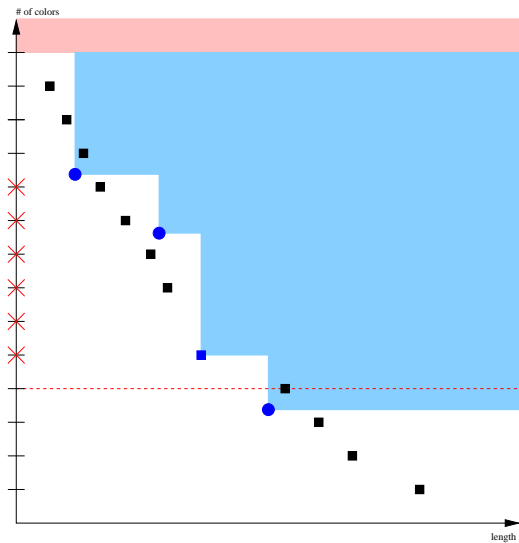
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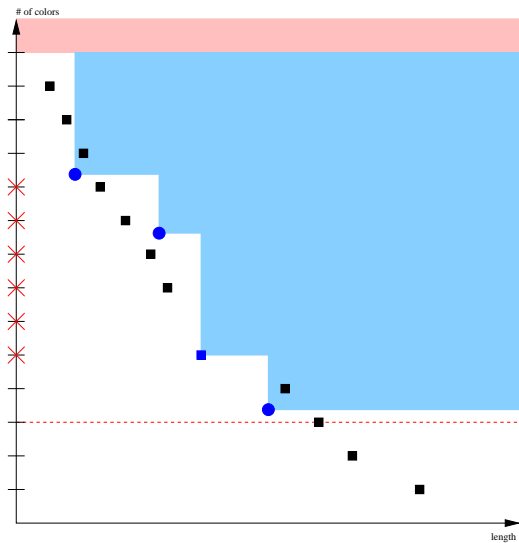
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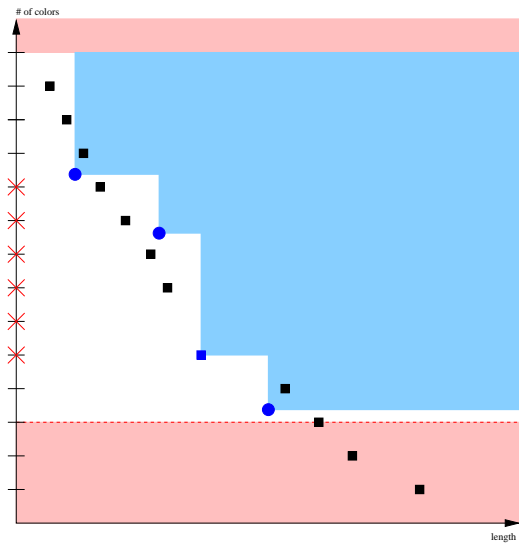
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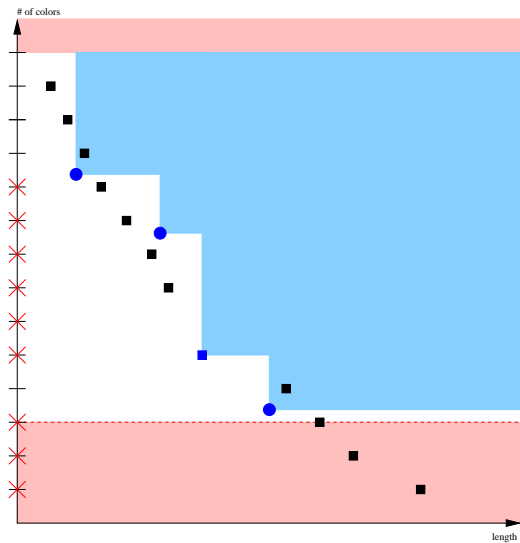
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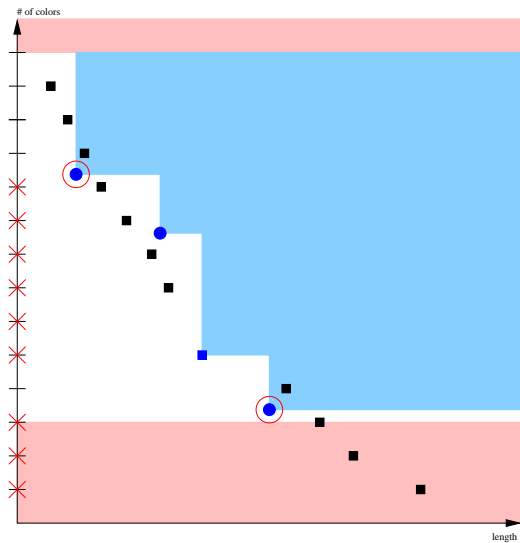


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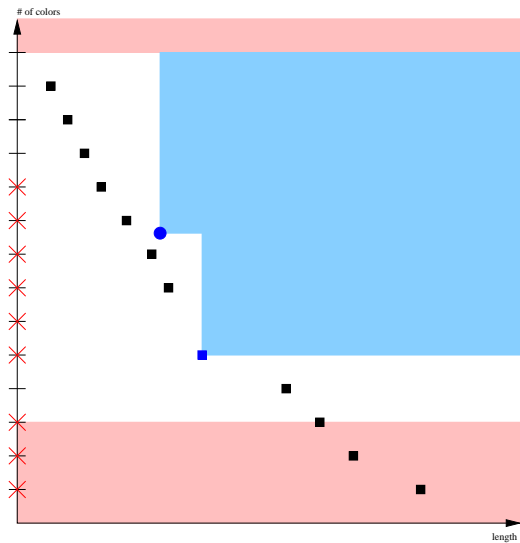




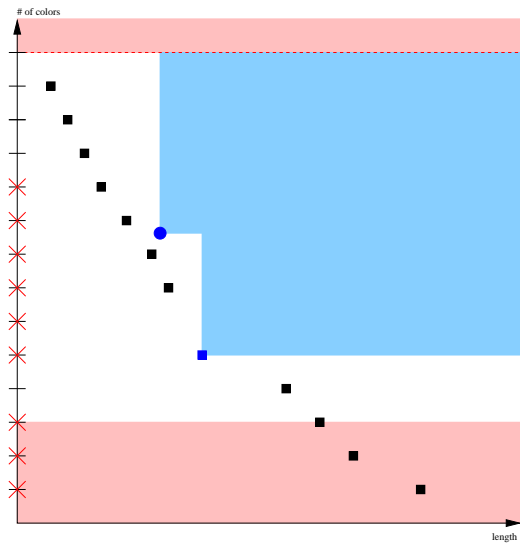
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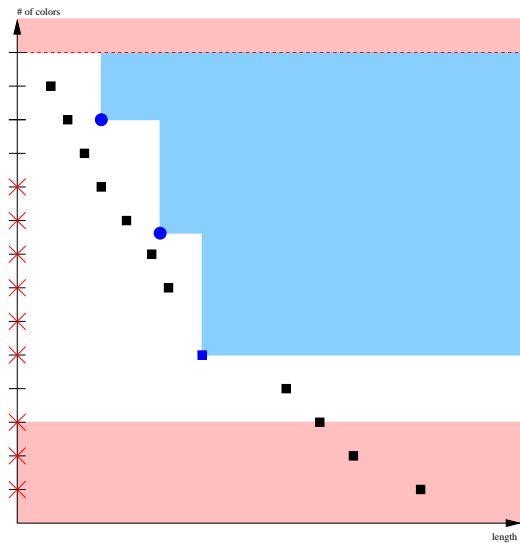
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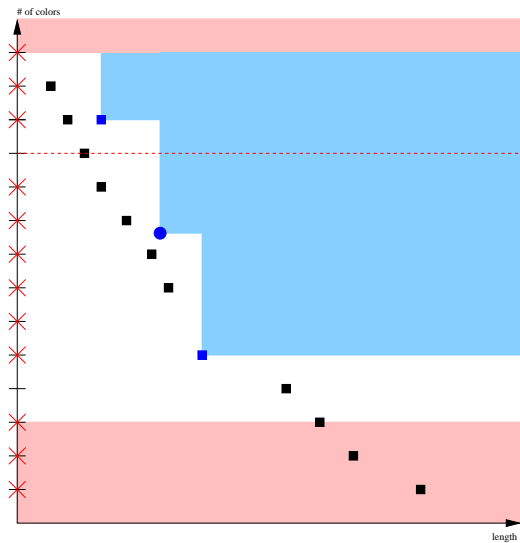
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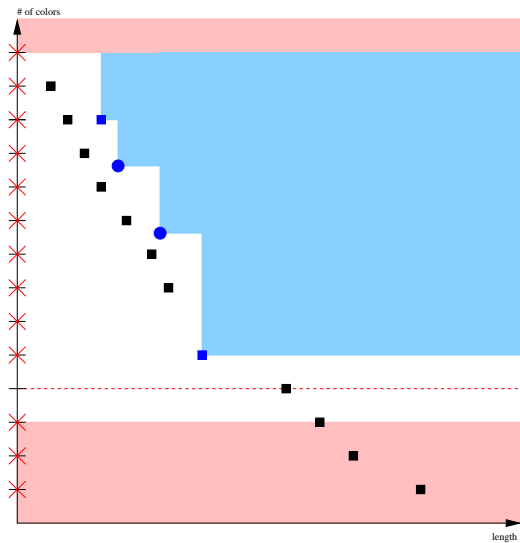




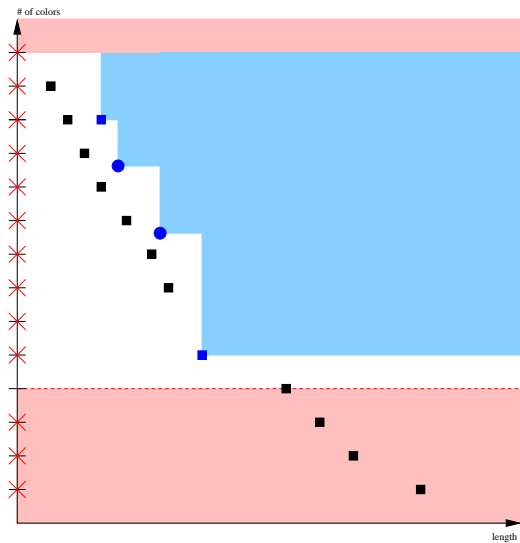




## COMPUTATION OF THE LOWER BOUND



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## CONSTRAINT GENERATION, CUTTING, AND BRANCHING

### Constraint generation

Connectivity constraints  $\rightarrow$  min-cut problem

Call to a CONCORDE function [Padberg & Rinaldi, 1990]

Test if the color connectivity constraint is also violated

### Cutting

$\forall \epsilon$ , the sub-problem is unfeasible

$\forall \epsilon$ , the solution is either feasible or dominated by  $ub$

### Branching

First on the  $u_k$  variables then on the  $x_e$

Priority on the variable that is non integral for the most values of  $\epsilon$

## COMPUTATION OF THE INITIAL UPPER BOUND

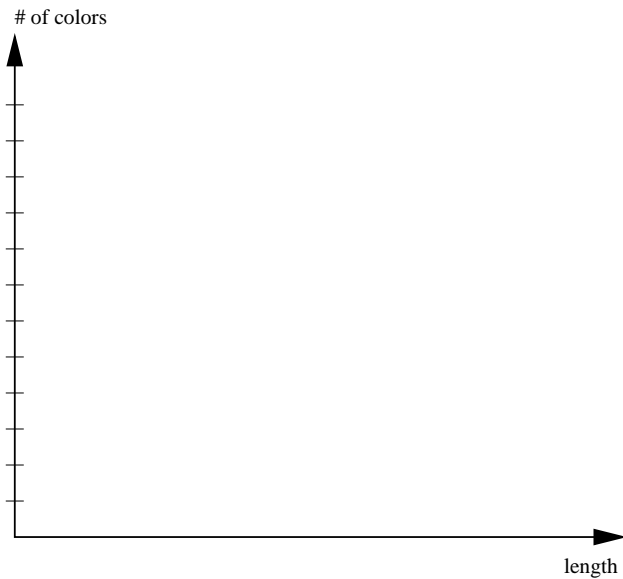
Repeat the following sequence for different values of  $\epsilon$

**STEP 1:** Solve the following mixed integer program :

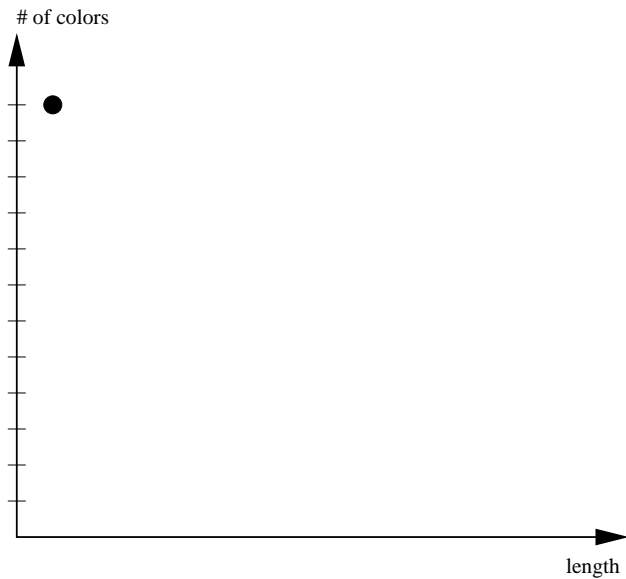
$$\begin{aligned} \min \quad & \sum_{e \in E} c_e x_e + m \sum_{k \in C} u_k \\ & \sum_{e \in \omega(\{i\})} x_e = 2 & \forall i \in V \\ & x_e \leq u_{\delta(e)} & \forall e \in E \\ & u_k \leq \sum_{e \in \zeta(k)} x_e & \forall k \in C \\ & \sum_{k \in C} \gamma_i^k u_k \geq 2 & \forall i \in V \\ & \sum_{k \in C} u_k \leq \epsilon \\ & 0 \leq x_e \leq 1 & \forall e \in E \\ & u_k \in \{0, 1\} & \forall k \in C \end{aligned}$$

**STEP 2:** Solve a TSP on  $G' = (V, E')$  with  $E' = \{e \in E \mid u_{\delta(e)} = 1\}$

## COMPUTATION OF THE INITIAL UPPER BOUND

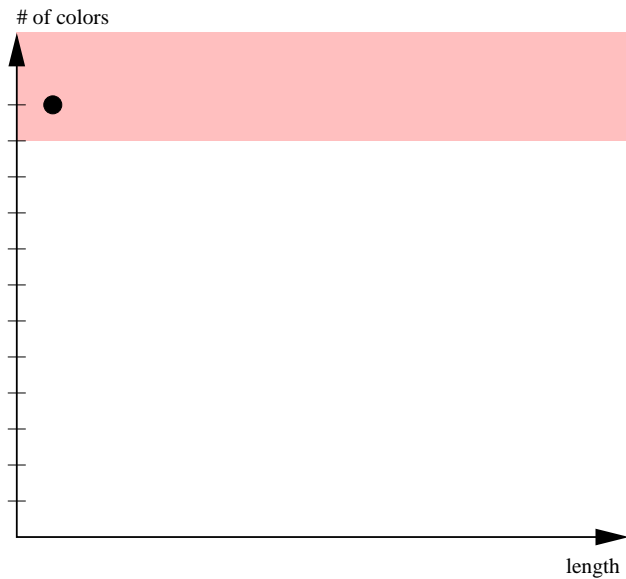


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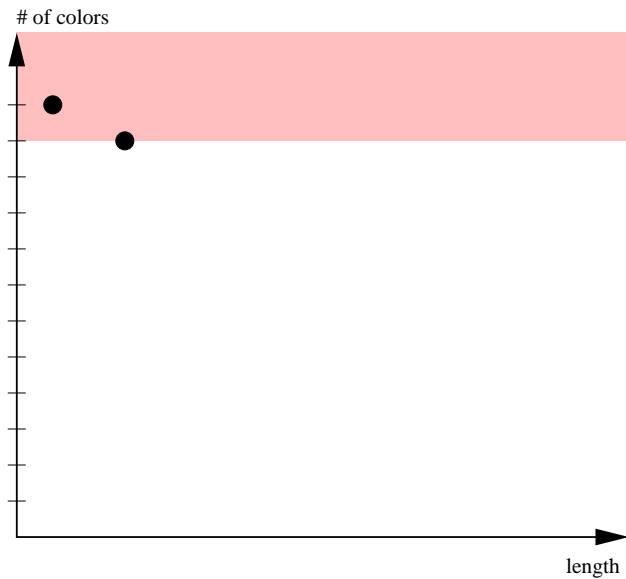




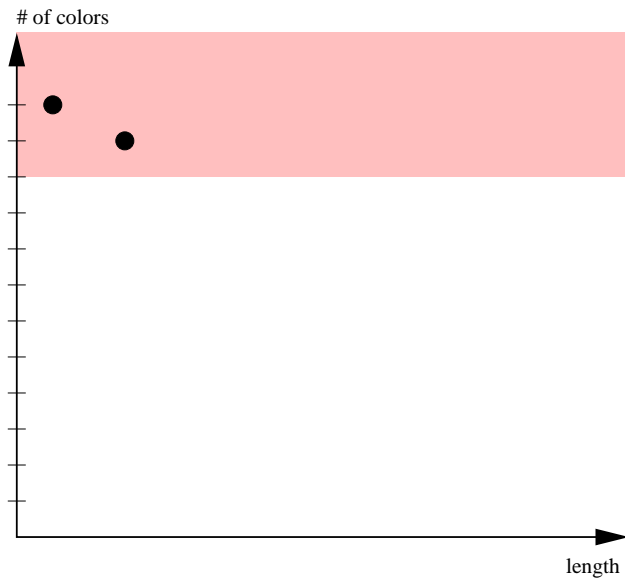
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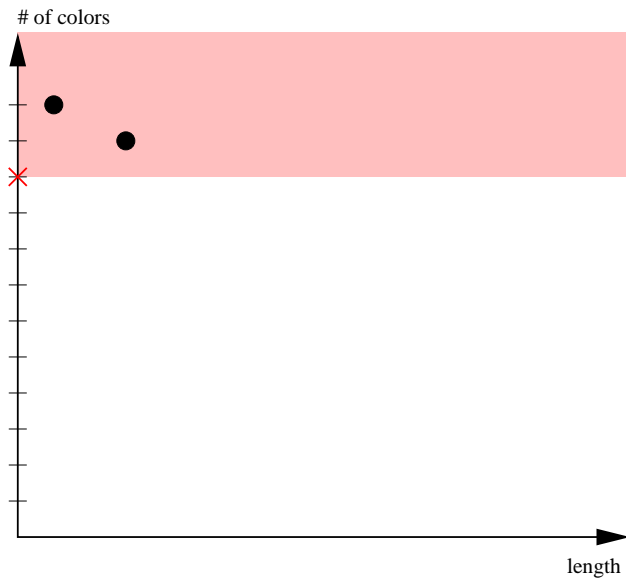
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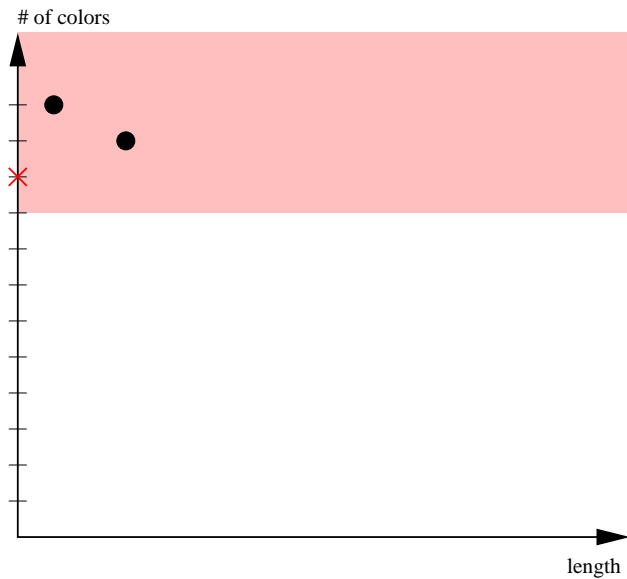
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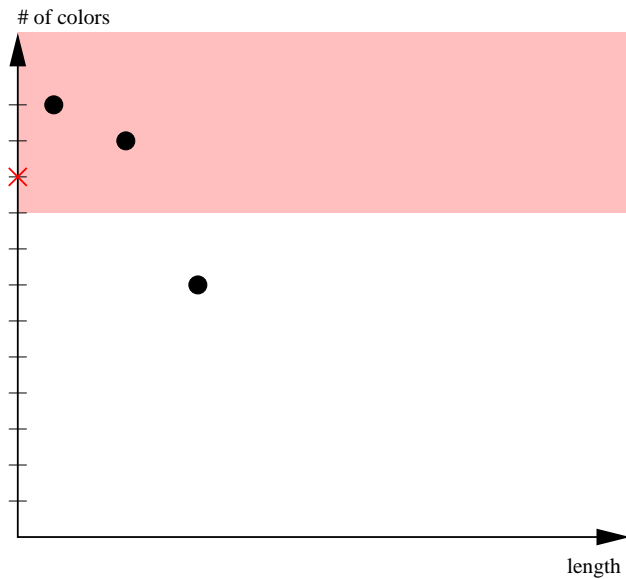
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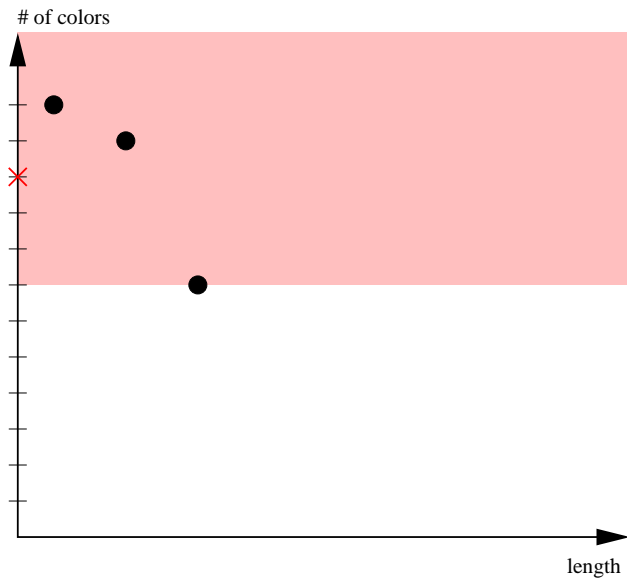
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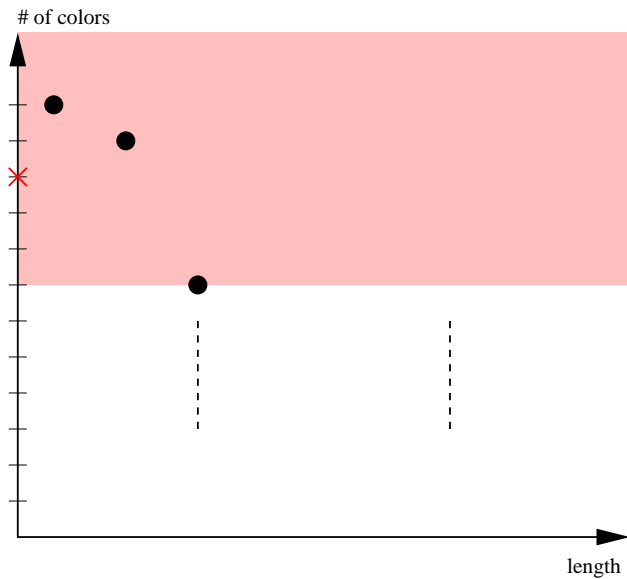
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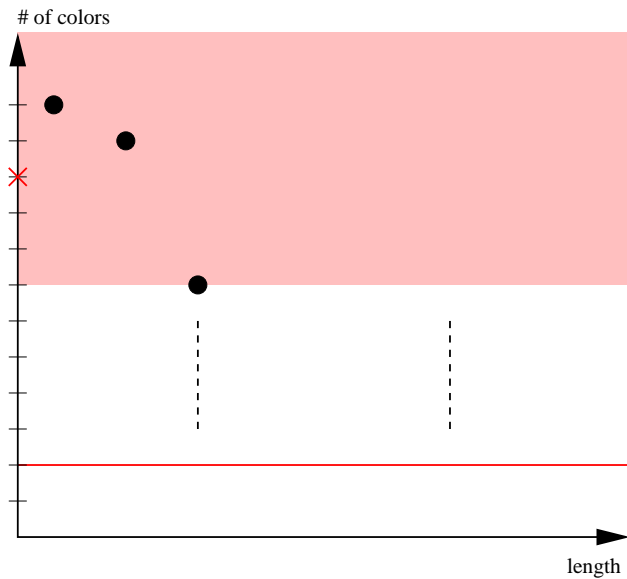


## COMPUTATION OF THE INITIAL UPPER BOUND





## COMPUTATION OF THE INITIAL UPPER BOUND



## COMPUTATIONAL RESULTS

$ C $	$ V $	#nodes	#u	#x	#cut	#Pareto	#Parub	#time
10	20	77.1	32.1	5.9	31.5	7.9	4.7	0.6
10	30	132.0	50.1	15.4	70.2	8.5	5.5	4.3
10	40	220.4	60.8	48.9	109.4	9.2	5.5	12.0
10	50	411.6	70.8	134.4	148.6	10.0	6.3	37.0
15	20	166.5	77.0	5.8	48.4	9.1	4.4	2.1
15	30	393.9	170.0	26.4	133.4	12.2	4.9	17.9
15	40	551.6	209.3	66.0	208.5	12.6	5.2	52.0
15	50	1880.8	362.0	577.9	340.1	13.5	5.7	207.2
20	20	350.5	169.5	5.3	78.4	10.3	4.5	6.1
20	30	781.9	357.9	32.6	194.8	13.9	4.7	45.9
20	40	1394.5	603.9	92.9	348.8	15.9	5.3	191.9
20	50	2327.0	855.2	307.8	575.8	17.4	5.8	780.0
25	20	429.5	207.2	7.0	84.4	11.1	4.5	8.1
25	30	1596.5	769.5	28.2	304.1	15.2	4.3	147.7
25	40	3200.8	1433.9	166.0	611.0	17.6	4.8	940.9
25	50	5634.5	2376.0	440.7	962.5	20.9	5.5	3915.9
30	20	792.5	391.1	4.6	137.6	12.4	4.5	20.8
30	30	2232.0	1062.5	53.0	364.2	16.4	4.6	259.8
30	40	4866.0	2247.8	184.7	757.2	18.8	4.3	2170.6
30	50	10169.1	4219.0	865.0	1370.0	21.7	5.1	9705.7

## CONCLUSIONS AND PERSPECTIVES

- ▶ Branch-and-cut algorithm able to solve a multi-objective problem in one run
- ▶ Identify new valid constraints  $\rightarrow$  variables  $u_k$
- ▶ Rules to choose on which variables to branch
- ▶ Progressive partition of the objective space