

# ALGORITHMIC METHODS FOR MATHEMATICAL MODELS

## Course Project Report

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December 2019

## 1 Formal statement problems A, B and C

### 1.1 Base problem

Having a company that produces cars, the problem can be formally stated as follows.

*Given:*

- The set  $O$  of options that a car can have.
- The set  $C$  of classes a car belongs. Each class  $c$  is a set of options.
- A Boolean matrix  $ClassOption_{c,o}$  whether a  $c \in C$  has option  $o \in O$
- An integer array  $carsOfClass_c$  representing the numbers of cars of class  $c \in C$  that will be produced
- A sequence  $P$  of positions to fill in the production line

*Find* a production schedule of cars that satisfies:

- Each position  $p \in P$  has one and only one car of class  $c \in C$  assigned
- For each  $c \in C$ , the corresponding number  $carsOfClass_c$  cars need to be produced
- Each class  $c \in C$  has the corresponding options  $o \in O$  specified in  $ClassOption_{c,o}$

### 1.2 Problem A

*Given:*

- The statement defined in the **base problem**.
- Two integers  $k_o$  and  $m_o$  for every option  $o \in O$

*Find* a production schedule of cars that satisfies:

- That for every window of  $k_o$  consecutive cars, at most  $m_o$  can have option  $o$ .

### 1.3 Problem B

*Given:*

- The statement defined in the **problem A**.
- That a position  $p \in P$  has a change if  $p - 1$  and  $p + 1$  exist, the classes at positions  $p - 1$  and  $p$  are different from each other and the classes at positions  $p$  and  $p + 1$  are also different from each other.

The *objective* is to minimize the number of positions where there is a change.

## 1.4 Problem C

*Given:*

- The statement defined in the **base problem**.
- Two integers  $k_o$  and  $m_o$  for every option  $o \in O$
- That for every window of  $k_o$  consecutive cars in  $P$ , for each option  $o \in O$ , there is a violation if there are more than  $m_o$  cars with option  $o$ .

The *objective* is to minimize the number of violations.

## 2 Integer linear programming formulation

### 2.1 Base problem

$\mathcal{O}$  : Set of options, index  $o$

$\mathcal{C}$  : Set of classes, index  $c$

$classOption_{c,o}$  : whether class  $c \in \mathcal{C}$  has option  $o \in \mathcal{O}$

$carsOfClass_c$  : number of cars of class  $c \in \mathcal{C}$  to be produced

$\mathcal{P} = \{1, 2, \dots, nPositions\}$  : Set of positions to fill,  $nPosition = \sum_{c \in \mathcal{C}} carsOfClass_c$

We model the problem as a Integer Linear Program (ILP) with the following decision variables:

$pc_{p,c}$  : whether a car of class  $c$  is assigned to position  $p$

$po_{p,o}$  : whether a class with option  $o$  is assigned to position  $p$

Taking into account that each variable has to fulfill the following constraints:

$$\begin{aligned} \sum_{c \in \mathcal{C}} pc_{p,c} &= 1 && \text{for all } p \in P \text{ (all positions have one car)} \\ \sum_{p \in P} pc_{p,c} &= carsOfClass_c && \text{for all } c \in C \text{ (carsOfClass is fulfilled)} \\ pc_{p,c} = 1 &\implies po_{p,o} = classOption_{c,o} && \text{for all } p \in P, c \in C, o \in O \text{ (ClassOptions is fulfilled)} \end{aligned}$$

### 2.2 Problem A

Taking into account the variables and constraints defined in the **base problem**, we will model the following statement:

minimize 1 (try to find a satisfiable model)

subject to:

$$\sum_{i=p}^{p+k_o-1} po_{i,o} \leq m_o \quad \text{for all } o \in O, 1 \leq p \leq nPositions - k_o + 1 \text{ (window violations)}$$

### 2.3 Problem B

Taking into account the variables and constraints defined in the **problem A**, and the following decision variable:

$z$  : Positive integer with the number of changes

We model the following statement:

minimize  $z$  (minimize the number of changes)

subject to:

$$z \geq \sum_{c \in \mathcal{C}} \sum_{p=2}^{nPosition-1} pc_{p-1,c} \neq pc_{p,c} \wedge pc_{p,c} \neq pc_{p+1,c} \text{ (number of changes)}$$

## 2.4 Problem C

Taking into account the variables and constraints defined in the **base problem**, and the following boolean decision variable:

$zopt_{p,o}$  : Whether a option  $o$  is violated in position  $p$ .

We model the following statement:

$$\begin{aligned}
 & \text{minimize } \sum_{p \in P, o \in O} zopt_{p,o} \text{ (minimize the number of violations)} \\
 & \text{subject to:} \\
 & zopt_{p,o} = \left( \sum_{i=p}^{p+k_o-1} po_{i,o} \right) > m_o \text{ for all } o \in O, 1 \leq p \leq nPositions - k_o + 1 \text{ (number of violations)}
 \end{aligned}$$

### 3 Greedy and Grasp pseudocode

#### 3.1 Greedy

Our  $q(\cdot)$  FUNCTION differences two states:

- If is the **first car**, we choose the class that has lower number of cars in carsOfClass. The ideal case is to put a class that only have 1 car.
- If is **not the first car**...
  - We will try to put any car that does not lead to a change.
  - If its not possible, we will prioritize the classes that will allow the assignation of 2 or more cars of that class. In order to do this, we check that, for each option present in the class, the number of 1's in the last  $k_o - 2$  positions plus the value of the option of the car that we are going to assign has to be smaller than  $m_o$ .  
 If there's no class that for all options the ratio is smaller than 1 we will choose those that are equal to 1, but knowing that in the following step this will lead us to another change. We will add 1 to be sure that if all the classes do not have any option, will take the previous condition rather than this one.
  - If a class has less than 1 car left to assign and forces a change, will be the last to be chosen.

$$q(p, o) = \begin{cases} \frac{-1}{carsOfClass_c} & \text{if } p = 0 \\ 0 & \text{if } S_{p-1} = c \\ \max \left\{ \frac{\sum_{i=\max(0, p-k_o+2)}^p po_{i,o} \times classOption_{c,o} + classOption_{c,o}}{m_o} \middle| o \in O \right\} + 1 & \text{if } S_{p-1} \neq c \wedge carsOfClass_c > 1 \\ 3 & \text{if } S_{p-1} \neq c \wedge carsOfClass_c \leq 1 \end{cases}$$

$$r(p, o) = \sum_{i=\max(0, p-k_o)}^p po_{i,o}$$

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**Algorithm 1** GREEDY

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```
function GETFEASIBLECLASS(p):  
    feasibleClass  $\leftarrow \emptyset$   
    for c in C do:  
        if  $|\{o \in O | r(p, o) > m(o)\}| = 0$  then  
            feasibleClass  $\leftarrow$  feasibleClass  $\cup$  c  
        end if  
    end for  
    return feasibleClass  
end function  
  
function GREEDY:  
    S  $\leftarrow \emptyset$   
    po  $\leftarrow$  Matrix(p,o)  
    p = 1  
    while | S |  $\leq$  nPositions do:  
        feasibleClass =: getFeasibleClass(p)  
        if feasibleClass =  $\emptyset$  then  
            return INFEASIBLE  
        end if  
        S  $\leftarrow$  S  $\cup$  argmin{q(p,c) | c  $\in$  feasibleClass  $\wedge$  carsOfClass(c)  $\neq$  0}  
        carsOfClass(c)  $\leftarrow$  carsOfClass(c) - 1  
        Update pop  
        p  $\leftarrow$  p + 1  
    end while  
    return S  
end function
```

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### 3.2 GRASP

$$q(p, o) = \begin{cases} \frac{-1}{carsOfClass_c} & \text{if } p = 0 \\ 0 & \text{if } S_{p-1} = c \\ \max \left\{ \frac{\sum_{i=\max(0, p-k_o+2)}^p po_{i,o} \times classOption_{c,o} + classOption_{c,o}}{m_o} \middle| o \in O \right\} + 1 & \text{if } S_{p-1} \neq c \wedge carsOfClass_c > 1 \\ 3 & \text{if } S_{p-1} \neq c \wedge carsOfClass_c \leq 1 \end{cases}$$

$$r(p, o) = \sum_{i=\max(0, p-k_o)}^p po_{i,o}$$

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#### Algorithm 2 GRASP

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```

function GETFEASIBLECLASS(p):
    feasibleClass  $\leftarrow \emptyset$ 
    for c in C do:
        if  $|\{o \in O | r(p, o) > m(o)\}| = 0$  then
             $feasibleClass \leftarrow feasibleClass \cup c$ 
        end if
    end for
    return feasibleClass
end function

```

```

function GRASP:
    S  $\leftarrow \emptyset$ 
    po  $\leftarrow$  Matrix(p, o)
    p = 1
    while  $|S| \leq nPositions$  do:
        feasibleClass = getFeasibleClass(p)
        if  $feasibleClass = \emptyset$  then
            return INFEASIBLE
        end if
         $q_{min} \leftarrow \min\{q(p, c) | c \in feasibleClass \wedge carsOfClass(c) \neq 0\}$ 
         $q_{max} \leftarrow \max\{q(p, c) | c \in feasibleClass \wedge carsOfClass(c) \neq 0\}$ 
         $RCL \leftarrow \{c \in feasibleClass \wedge carsOfClass(c) \neq 0 | q(p, c) \leq q_{min} + \alpha(q_{max} - q_{min})\}$ 
         $c \leftarrow$  select  $c \in RCL$  at random
         $S \leftarrow S \cup c$ 
        carsOfClass(c)  $\leftarrow$  carsOfClass(c) - 1
        Update  $po_p$ 
        p  $\leftarrow$  p + 1
    end while
    return S
end function

```

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### 3.3 Code execution

Position	1	2	3	4	5
Class	1	1	2	4	4

#### 3.3.1 Greedy results

The following table shows the result of the  $q(\cdot)$  function for each candidate in both iterations:

Candidate	$q(\cdot)$ Iteration 1	$q(\cdot)$ Iteration 2
1	3	3
2	Infeasible	Infeasible
3	2	2
4	1.5	0
5	3	3
6	2	2
7	Infeasible	Infeasible

#### 3.3.2 GRASP results

We used the same  $q(\cdot)$  of GRASP so the costs are the same as in the Greedy results.

Having  $q_{min} = 1.5$  and  $q_{max} = 3$ , the RCL contains the candidates with cost between  $[1.5, 2.25]$ . The RCL for the first iteration with an  $\alpha$  of 0.5 is the following one:  $RCL_1 = \{1, 3, 4, 6\}$ .

We can't know the results of the second iteration because they depend on the candidate chosen in the first iteration.