

# Pseudo Code for Repair Algorithm

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## 1 Introduction

In this section, we define some important definitions and algorithms....

**Definition 1.1.** (Mixed Integer Programming)

In this report we consider a generic mixed-integer programming problem (MIP) in the following form

$$\begin{aligned} \text{(MIP)} \quad & \min \quad c^T x \\ \text{s.t.} \quad & Ax \geq b \\ & x_j \in \mathbb{Z} \quad \forall j \in \mathcal{I} \\ & x_j \in \mathbb{R} \quad \forall j \in \mathcal{N} \setminus \mathcal{I} \end{aligned}$$

where the vector  $b \in \mathbb{R}^m$  and the vector  $c \in \mathbb{R}^n$  are input vectors.  $A$  is a input matrix of size  $m \times n$ , the variable input set  $\mathcal{I} \subseteq \mathcal{N} = \{1, 2, \dots, n\}$ . We denote  $\mathcal{P}$  for this problem, which called a mixed-integer programming problem (MIP) with minimize objective function  $c^T x$  subject to the constraints  $Ax \geq b$ . Besides, some variables are restricted to integer values while the else of are restricted to real value.  $S$  is a set of feasible solution if  $S$  satisfy all the constraints in the problem. A vector  $s^*$  with  $s^* \in S$  is called *optimal solution* when  $c^T x_{s^*} \leq c^T x_s$  for  $\forall s \in S$ . When all of the variables are restricted to integer, the problem is called *pureintegerlinearprogram*(IP) for  $\mathcal{I} = \mathcal{N}$ . If there is no integrality constraint, the program is called *linearprogram*

$$\begin{aligned} \text{(MIP)} \quad & \min \quad c^T x \\ \text{s.t.} \quad & Ax \geq b \\ & x_j \in \mathbb{R} \quad \forall j \in \mathcal{N} \end{aligned}$$

**Definition 1.2.** (LP-relaxation)

Lp *relaxation* is obtained by removing all integrity constraints  $\mathcal{I} \leftarrow \emptyset$ . LP-*relaxation* is the foundation of LP-based branch-and-bound technology. As the searching space is increase by removing integrity restrictions, the optimal solution in MIP problem could not better than LP-*relaxation*, which is  $s_{MIP}^* \geq s_{LP}^*$ . This means the optimal solution found in LP problem could provide a lower or prime bound for MIP problem.

## 2 Input

- A MIP problem  $\mathcal{P}^0$  with  $n$  variables  $x$ , constraint set  $C^0$  with an optimal solution  $s^0$ , where  $s^0$  is a  $n$ -vector.
- A MIP problem  $\mathcal{P}^1$  with  $n$  variables  $x$ , constraint set  $C^1$ , such that  $C^0 \subsetneq C^1$ .

## 3 Output

- An optimal solution  $s^1$  to  $\mathcal{P}^1$ , where  $s^1$  is a  $n$ -vector too.

## 4 Pseudo Code

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### Algorithm 1: Solving Problem with Reoptimization

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**Input:**  $k, \mathcal{P}^0, s^0$  and  $\mathcal{P}^1$  where  $C^0 \subsetneq C^1$

**Output:** optimal solution  $s^*$  to  $\mathcal{P}^1$

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1 begin
2   if  $s^0$  is feasible to  $\mathcal{P}^1$  then
3     return  $s^0$ 
4   else
5      $\alpha \leftarrow k \times c^T s^0$ 
6     create new variables  $y$  where  $y_j \in \mathbb{Z}, \forall j \in \mathcal{I}$            // add new variables
7     for  $j$  in  $\mathcal{I}$  do
8        $\mathcal{P}^2 \leftarrow$  add constraints to  $\mathcal{P}^1 : y_j \geq x_j - s_j^0$            // add new constraints
9        $\mathcal{P}^2 \leftarrow$  add constraints to  $\mathcal{P}^2 : y_j \geq s_j^0 - x_j$ 
10    end
11    add penalty function to  $\mathcal{P}^2 : \alpha \times \sum_j y_j, j \in \mathcal{I}$            // add penalty function
12     $r \leftarrow 1$ 
13    for  $\alpha \geq 0$  do
14       $s^r \leftarrow$  solve  $\mathcal{P}_r^2$  ( $r$ -th iteration) to optimal or stop criterion with reoptimization
15       $\alpha \leftarrow \alpha - \frac{\alpha}{k}$ 
16    end
17     $s^* \leftarrow s^r$ 
18    return  $s^*$ 
19  end
20 end

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