

Pseudo Code for Repair Algorithm

Arthur Feng

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1 Introduction

Definition 1.1. (Mixed Integer Programming)

In this report, we denote the *Mixed Integer programming* (MIP) problem in the follow form

$$(MIP) \quad \min\{c^T x : Ax \leq b, x \in \mathbb{Z}^l \times \mathbb{R}^k\}$$

where $m, n \in \mathbb{N}$, and $n = l + k$, A is a matrix for $A \in \mathbb{R}^{m \times n}$, and the vector $b \in \mathbb{R}^m$, the vector $c \in \mathbb{R}^n$.

S is a set of *feasible solution* if S satisfy all the constraints in the problem. If $s \in S$, s is called a feasible solution to the problem \mathcal{Z}_{MIP} . The MIP is called infeasible when S is empty, other wise, the MIP is called feasible. The feasible solution s^* is called optimal when $s^* \in S$ and $c^T x_{s^*} \leq c^T x_s$ for $\forall s \in S$.

Definition 1.2. (Linear Programming and LP-relaxation)

With the definition of MIP in 1.1, there is a special case of MIP when all variables are continuous, which is called *Linear Programming* (LP) problem

$$\mathcal{Z}_{LP} = \min\{c^T x : Ax \leq b, x \in \mathbb{R}^k\}$$

A LP can also be obtained by removing all integrity constraints: $x_i \in \mathbb{Z}$ where $i \in n \setminus l$, this is called *LP-relaxation*. *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 could not better than *LP-relaxation*, which is $s_{MIP}^* \geq s_{LP}^*$. This means the optimal solution found in LP could provide a lower or prime bound for MIP.

Definition 1.3. (Binary Programs)

In the definition 1.1, When all variables are integer

$$\mathcal{Z}_{IP} = \min\{c^T x : Ax \leq b, x \in \mathbb{Z}^l\}$$

The problem is call pure Integer Programming (IP) problem. And if we denote \mathbb{B} be a set of binary values where $\mathbb{B} = \{0, 1\}$ for $\forall x \in \mathbb{B}^l$, then we call this problem a Binary Programming problem. This report we focus on Binary Programs.

Question: the two terminologies *Binary programming problem* and *pseudo – Boolean problem* is same? or not?

2 Input

- A problem $\mathcal{P}^0 = \min\{c^T x \mid Ax \geq b, x_i \in \mathbb{B}^n, \mathbb{B} = \{0, 1\}\}$ with n binary variables x , objective function $O^0 = c^T x$ (min), constraint set C^0 with an optimal solution s^0 .
- A problem \mathcal{P}^1 with n binary variables x , objective function $O^1 = O^0 = c^T x$ (min), constraint set C^1 , such that $C^0 \subsetneq C^1$.

3 Output

- An optimal solution s^1 to \mathcal{P}^1 .

4 Pseudo Code

Algorithm 1: Solving Problem with Reoptimization

Data: r, \mathcal{P}^0, s^0 and \mathcal{P}^1 where $C^0 \subsetneq C^1$, $O^0 = O^1$

Result: optimal solution s^1 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      $C^+ \leftarrow C^1 - C^0$ 
6     for  $c \in C^+$  do
7       add  $v_i$  to the set  $V$  where it's coefficient  $c_i \neq 0$ 
8     end
9      $y^0 = c^T x$  for  $x \in s^0$ 
10     $y^p = c^T x$  for  $x \in V^+$ 
11     $\Delta y = |Y^0 - Y^p|$ 
12    if  $\Delta y / y^0 \geq r$  then
13      solve  $\mathcal{P}^1$  from scratch
14    else
15      solve it with reoptimization
16       $p \leftarrow (|y^0| + |y^p|)$ 
17       $O^p \leftarrow p * \sum x$  for  $x \in V^+$ 
18       $\mathcal{P}_0^* \leftarrow O^p + C1$ 
19      while  $p \geq 0$  do
20         $p = p / 1.2$ 
21         $\mathcal{P}_{i+1}^* \leftarrow$  update  $O^p$ 
22        if  $i == 0$  then
23          input  $s^i$  to find the first feasible solution
24          question: do we need to fix  $s^0$  ?
25        end
26         $s_i^* \leftarrow$  reoptimize the sub-problem in SCIP
27        if  $s_i^*$  don't change in  $n$  iterations then
28           $s^1 \leftarrow s_i^*$ 
29          return  $s_1$ 
30        else if  $p = 0$  then
31           $s^1 \leftarrow s_i^*$ 
32          return  $s_1$ 
33        end
34      end
35    end
36 end

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