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418 FINAL PROJECT

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**A review of mixed integer knapsack problems**

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# Contents

<b>1</b>	<b>Introduction</b>	<b>2</b>
<b>2</b>	<b>Algorithms for solving MIKP</b>	<b>2</b>
2.1	Infeasible, unbounded, and trivial instances of MIKP . . . . .	2
2.2	Preprocessing an instance of MIKP . . . . .	4
2.3	Solving the LP relaxation of PP-MIKP . . . . .	6

# 1 Introduction

Consider a positive integer  $n$ , letting  $b \in \mathbb{Q}$ ,  $a \in \mathbb{Q}^n$ ,  $l \in \{\mathbb{Q} \cup \{-\infty\}\}^n$ ,  $u \in \{\mathbb{Q} \cup \{+\infty\}\}^n$  and  $I \subset [n] := \{1, \dots, n\}$ . The Mixed Integer Knapsack Set is defined as

$$K = \{x \in \mathbb{R}^n : a^T x \leq b, l \leq x \leq u, x_i \in \mathbb{Z}, \forall i \in I\}. \quad (1)$$

Furthermore, if we have  $c \in \mathbb{Q}^n$  and assume  $l_i$  is finite for each  $i \in [n]$ , then the Mixed Integer Knapsack Problem (MIKP) can be described as:

$$\max\{c^T x : x \in K\}. \quad (2)$$

We assume that for each  $k = 1, 2, \dots, n$  either  $a_k \neq 0$  or  $c_k \neq 0$ , otherwise, we could remove variable  $x_k$  without affecting the problem.

In this report, we present a new branch-and-bound algorithm for MIKP. The methodology that we propose is a linear-programming-based algorithm which exploits dominance conditions. We further make use of lexicographic-domination conditions to eliminate problems with symmetry. One interesting aspect of this approach is that it differs from traditional linear-programming based algorithms by allowing feasible solutions to be pruned during the branching phase.

It might be very difficult to solve MIKP, even by using very effective mixed integer programming solvers such as CPLEX [1]. However, the proposed algorithm is shown to be very effective in solving instances of MIKP, much more effective than CPLEX in fact, both in the amount of time taken to solve problems as by the size of the branch and bound tree explored to find the optimal solution.

In the following content, we will state procedures aiming to solve MIKP

1. An easy way of identifying unbounded solutions.
2. A way of pre-processing instances of MIKP.
3. The issue of quickly solving the LP-relaxation of MIKP.
4. A simple branch-and-bound algorithm for MIKP.
5. A enhancement of the branch-and-bound algorithm by introducing domination-criteria.

And finally, we will analyze the computational results of our algorithm and compare it with the general mixed-integer-programing solver CPLEX.

## 2 Algorithms for solving MIKP

### 2.1 Infeasible, unbounded, and trivial instances of MIKP

We aim to use a simple procedure to verify that our MIKP instances are either infeasible, unbounded or trivial, where 'trivial' stands for that our instances are very easy to solve.

Actually, it is very easy to detect the infeasibility of a problem.

**Lemma 1.** *If there is any variable  $x_i$  with  $i \in [n]$  such that  $a_i > 0$  and  $l_i = \infty$ , or such that  $a_i < 0$  and  $u_i = \infty$ , then the problem is feasible.*

We define the concept of 'efficiency' which can tell us how valuable it is relative to the amount of capacity it uses up in the knapsack constraints.

potentiator	$(a_k \leq 0, c_k > 0, u_k = +\infty)$ or $(a_k \geq 0, c_k < 0, l_k = -\infty)$
accumulator	$(a_k < 0, c_k = 0, u_k = +\infty)$ or $(a_k > 0, c_k = 0, l_k = -\infty)$
incrementor	$(a_k > 0, c_k > 0, u_k = +\infty)$ or $(a_k < 0, c_k < 0, l_k = -\infty)$
decrementor	$(a_k > 0, c_k \geq 0, l_k = -\infty)$ or $(a_k < 0, c_k \leq 0, u_k = -\infty)$

Table 1: Status of a MIKP

**Definition 1.** Consider  $k \in [n]$  and define

$$e_k = \begin{cases} c_k/a_k & \text{if } a_k \neq 0, \\ +\infty & \text{if } a_k = 0 \text{ and } c_k > 0, \\ -\infty & \text{if } a_k = 0 \text{ and } c_k < 0. \end{cases} \quad (3)$$

We say that  $e_k$  is the efficiency of variable  $x_k$ .

Furthermore, we also define some status that use them to claim the situation of our problem. Actually, we say that

These four definition are very useful, actually, we can obtain the following lemma

**Lemma 2.** *If MIKB is feasible and admits a potentiator, then MIKP is unbounded.*

**Lemma 3.** *If MIKB is feasible, and admits an incrementor  $x_i$  and a decrementor  $x_j$  such that  $e_i > e_j$ , then MIKP is unbounded.*

**Proposition 1.** *MIKP is unbounded if and only one of the following conditions hold,*

- *MIKP is feasible and admits a potentiator  $x_j$ .*
- *MIKP is feasible and admits an incrementor  $x_i$  and a decrementor  $x_j$  such that  $e_i > e_j$ .*

Note that even if MIKP is bounded, it may still admit an accumulator. We further define the 'triviality' of MIKP.

**Definition 2.** Consider an instance of MIKP which is feasible and not unbounded. If MIKP has an accumulator, we say that MIKP is trivial.

Actually, a trivial MIKP can be easily solved by considering the coefficients of the problem.

**Proposition 2.** *Assume that MIKP is feasible and not unbounded. In addition, let  $j$  correspond to an accumulator of MIKP. For each  $k \in [n]$  such that  $k \neq j$  define:*

- $U_k = \begin{cases} \lfloor u_k \rfloor & \text{if } k \in I, \\ u_k & \text{otherwise.} \end{cases}$
- $L_k = \begin{cases} \lceil l_k \rceil & \text{if } k \in I, \\ l_k & \text{otherwise.} \end{cases}$
- $x_k = \begin{cases} U_k & \text{if } (c_k > 0) \text{ or } (c_k = 0 \text{ and } u_k < \infty), \\ L_k & \text{if } (c_k < 0) \text{ or } (c_k = 0 \text{ and } l_k > -\infty), \\ 0 & \text{if } c_k = 0 \text{ and } x_k \text{ is free.} \end{cases}$

In addition, with respect to  $k = j$ , we claim that

$$x_j = \begin{cases} \max\{\lceil -\frac{\sum_{k \neq j} a_k x_k - b}{a_j} \rceil, l_k\} & \text{if } a_j < 0, \\ \min\{\lfloor -\frac{\sum_{k \neq j} a_k x_k - b}{a_j} \rfloor, u_k\} & \text{if } a_j > 0. \end{cases} \quad (4)$$

Then, we derive that  $x$  is well-defined and corresponds to an optimal solution of MIKP.

Actually, we can build an algorithm to detect infeasibility, unbounded and find trivial solutions.

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**Algorithm 1** Detecting infeasibility unbounded and finding trivial solutions

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**Input:**  $c, a, b, e, l, u$

**Output:**  $status$

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1:  $e^+ \leftarrow -\infty; e^- \leftarrow +\infty$ 
2: for  $i = 1$  to  $n$  do
3:   if  $a_i > 0$  and  $l_i = -\infty$  or  $a_i < 0$  and  $u_i = \infty$  then
4:      $status \leftarrow$  infeasible
5:   end if
6:   if  $x_i$  is a potentiator then
7:      $status \leftarrow$  potentiator
8:   return
9:   else if  $x_i$  is an incrementor and  $e_i > e^+$  then
10:     $e^+ \leftarrow e_i$ 
11:   else if  $x_i$  is a decrementor and  $e_i < e^-$  then
12:     $e^- \leftarrow e_i$ 
13:   end if
14: end for
15: if  $e^+ > e^-$  then
16:    $status \leftarrow$  incrementor/decrementor pair
17:   return
18: else if  $e^- = 0$  then
19:    $status \leftarrow$  accumulator
20:   A solution  $x$  is given.
21: end if
22: return

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## 2.2 Preprocessing an instance of MIKP

In this section we are concerned with reducing an instance of MIKP to another, equivalent instance of MIKP which is easier to solve. A series of procedures for pre-processing an instance of MIKP are now presented. For a thorough introduction to preprocessing see [2].

**Test if MIKP is infeasible, trivial or unbounded.** Using Algorithm 1 to test if MIKP is infeasible, trivial or unbounded. If MIKP is feasible, not trivial and not unbounded, which means it has no potentiators and no accumulators. In addition if variable  $x_i$  is an incrementor, and  $x_j$  a decrementor, then  $e_i e_j$ .

**Strength bound.** We give a strength bound for our problem. First, we define

$$\bullet U_k = \begin{cases} +\infty & \text{if } a_k \leq 0, \\ (b - \sum_{i \neq k, a_i > 0} a_i l_i - \sum_{i \neq k, a_i < 0} a_i u_i) / a_k & \text{otherwise.} \end{cases}$$

$$\bullet L_k = \begin{cases} -\infty & \text{if } a_k \geq 0 \\ (b - \sum_{i \neq k, a_i > 0} a_i u_i - \sum_{i \neq k, a_i < 0} a_i l_i) / a_k & \text{otherwise.} \end{cases}$$

Then, we redefine the stronger bounds of our problem

$$\begin{cases} u_k = \min\{u_k, U_k\}, l_k = \max\{l_k, L_k\} & \text{if } k \notin I, \\ u_k = \min\{\lfloor u_k \rfloor, \lfloor U_k \rfloor\}, l_k = \max\{\lceil l_k \rceil, \lceil L_k \rceil\} & \text{if } k \notin I. \end{cases} \quad (5)$$

**Fix values of variable.** For a given variable  $x_k$ , we claim

$$x_k = \begin{cases} u_k & \text{if } a_k \leq 0 \text{ and } c_k \geq 0, \\ l_k & \text{if } a_k \geq 0 \text{ and } c_k \leq 0. \end{cases} \quad (6)$$

After fixing variables as described above, we can substitute out the values in MIKP and obtain a smaller problem with a new right-hand side. In the smaller problem, each variable  $x_k$  satisfies either  $(a_k > 0, c_k > 0)$  or  $a_k < 0, c_k < 0$ .

**Complement variables.** For simplicity, we introduce new variable to make sure the lower-bound is always non-negative. Consider a variable  $x_k$ , and then we set

$$x_k := \begin{cases} x_k - l_k & \text{if } -\infty < l_k < 0, \\ u_k - x_k & \text{if } l_k = -\infty. \end{cases} \quad (7)$$

**Sort data.** Sort the variables in order of decreasing efficiency. Break first ties if variables are of integer type or not. Break second ties by value of  $a_k$ .

**Aggregate variables.** For any given two variables  $x_i$  and  $x_j$ ,  $i, j \in I$ , if  $a_i = a_j$ ,  $c_i = c_j$ . We aggregate these two variables into a single variable  $x_k$  such that  $a_k = a_i$ ,  $c_k = c_i$ ,  $l_k = l_i + l_j$ ,  $u_k = u_i + u_j$  and  $k \in I$ . This procedure will be very helpful later in spending up the branch and bound algorithm.

After the several steps, our finally propose is to reformulate the original MIKP to the following PP-MIKP

$$\max \sum_{k \in P \cup N} c_k x_k \quad (8)$$

$$s.t. \sum_{k \in P \cup N} a_k x_k \leq b \quad (9)$$

$$l_k \leq x_k \leq u_k, \forall k \in P \cup N \quad (10)$$

$$x_k \in \mathbb{Z}, \forall k \in I. \quad (11)$$

where  $P = \{k : c_k > 0 \text{ and } a_k > 0\}$  and  $N = \{k : c_k < 0 \text{ and } a_k < 0\}$ . The PP-MIKP satisfies the following conditions:

- PP-MIKP is feasible.
- PP-MIKP is not unbounded, and is not trivial.
- The variable indices are sorted by efficiency.
- All variables  $x_k$  are such that  $(a_k > 0 \text{ and } c_k > 0)$  or  $(a_k < 0 \text{ and } c_k < 0)$ .
- For each  $k \in P \cup N$ , we have  $l_k > 0$ .
- For each  $k \in P \cup N$ , all finite bounds are tight; that is, there exists a feasible solution to MIKP which achieves the bound.
- There are no two identical variables.

### 2.3 Solving the LP relaxation of PP-MIKP

#### References

- [1] ILOG CPLEX. High-performance software for mathematical programming and optimization, 2005.
- [2] Martin WP Savelsbergh. Preprocessing and probing techniques for mixed integer programming problems. *ORSA Journal on Computing*, 6(4):445–454, 1994.