GRAVER BASIS

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To my brother Joaquín, for showing me what mathematics were during dinners at home.

To my parents, for teaching me beyond the scope of mathematics.

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Introduction

The underlying problem is the clasical *Integer program* (IP) that we formulate in the following way:

$$(IP)\equiv \max\{c^tx:Ax=b,l\leq x\leq u,x\in\mathbb{Z}^n\}$$
 $A\in\mathbb{Z}^{mxn},b\in\mathbb{Z}^m,c\in\mathbb{Z}^n,l$ and u lower and upper bounds for x

Despite the simplicity of its formulation, allowing only linear constraints and a linear objective function, it's well known the importance of IP. A large number of problems in diverse fields of the mathematics and algorithms (with an infinity of applications) admit an IP formulation. Unfortunately, it's also well known that IP is NP-Complete, what means that no efficient algorithm is likely to exist for solving IP in the general case. This explains the great interest in restricted formulations of the problem and in certain resolution techniques (even when they are not useful for the general IP).

In this project we first present the concept of **Graver Basis**, its properties, and its applications for solving the IP. We then study the N-Fold IP, a restricted formulation which has won relevance in the last decades given its theoretical properties and its wide applications. We show with the help of Graver Basis that N-fold can be solved in polynomial time. In chapters four and five we go further for obtaining a better complexity and show how applying these Graver Basis techniques (and specially its bounds) we can restrict the search of an improvement direction at a feasible point or even restrict the search to an optimal solution at a solution of the linear relaxation. All together lead us to a polynomial and efficient algorithm for this kind of problems.

Graver Basis

Definition 2.1. Two vectors $u, v \in \mathbb{R}^n$ are said to be **sign compatible** if $u_i \cdot v_i \geq 0$ for all $i \in \{1, ..., n\}$.

Definition 2.2. A vector $u \in ker(A)$ is **indecomposable** if it is not the sum of two sign compatible and non zero elements in ker(A).

Definition 2.3 (Graver basis). The Graver Basis of a given matrix $A \in \mathbb{Z}^{mxn}$ is defined as the set of integral indecomposable elements in the kernel of A. (Initially defined as *universal integral test set* in [Graver 1975])

Proposition 2.4. For every matrix A, Gr(A) is a finite set.

Proposition 2.5. Every integral element in ker(A) can be expressed as positive integral linear combination of elements in Gr(A).

Proposition 2.6. Given z in the feasible region of an IP, z is not optimum if and only if there exists $g \in Gr(A)$ s.t. $c^t g > 0$ and $l \le z + g \le u$

General IP algorithm using Graver basis

- 1. From a feasible solution z_i
- 2. Find g^* optimum for the sub-problem:

$$\max\{c^t g : g \in Gr(A), l \le z_i + g \le u\}$$

• $c^t g^* \leq 0 \implies z_i$ optimal solution.

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• $c^t g^* > 0 \implies g^*$ improvement direction, loop back to 1 with $z_{i+1} = z_i + \lambda \cdot g^*$ with the biggest λ respecting the bounds.

[References??]

2.1 Graver Basis ℓ_1 -norm bounds

Proposition 2.7 (Graver basis bounds). Given $A \in \mathbb{Z}^{mxn}$ and Δ an upper bound for the absolute value of each component of A, for every $g \in Gr(A)$:

- $||g||_1 \le m^{m/2} \Delta^m \cdot (n-m)$ [Onn 2010]
- $||g||_1 \le (2m\Delta + 1)^m$ [Eisenbrand, Hunkenschröder, Klein 2018]

Bases of augmentation algorithm

- If not optimal, an element in Graver basis is an improvement direction.
- If Graver basis bounded, we can restrict our improvement direction search.

General IP algorithm using Graver basis norm bound

- 1. From a feasible solution z_i
- 2. Find g^* optimum for the sub-problem:

$$\max\{c^t g : Ag = 0, l - z_i \le g \le u - z_i, g \in \mathbb{Z}^n, ||g||_1 \le ||Gr(A)||\}$$

- $g^* = 0 \implies z_i$ optimal solution.
- $g^* \neq 0 \implies g^*$ improvement direction, loop back to 1 with $z_{i+1} = z_i + \lambda \cdot g^*$ with the biggest λ respecting the bounds.

[Hemmecke, Onn, Romanchuk 2013]

N-Fold

A generalized N-Fold IP has constriction matrix A of the form $(A_i \in \mathbb{Z}^{rxt}, B_i \in \mathbb{Z}^{sxt})$:

$$N = \begin{pmatrix} A_1 & A_2 & \cdots & A_n \\ B_1 & 0 & \cdots & 0 \\ 0 & B_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & B_n \end{pmatrix}$$

Proposition 3.1. There is a polynomial time algorithm that, given any matrix $A \in \mathbb{Z}^{m \times n}$ along with its Graver Basis G(A), and vectors $x \in \mathbb{N}^n$ and $c \in \mathbb{Z}^n$ solves the integer program $IP_A(b,c)$.

Proposition 3.2. Fix any pair of integer matrices $A \in \mathbb{Z}^{r \times q}$ and $B \in \mathbb{Z}^{s \times q}$. Then there is a polynomial time algorithm that, given n, computes the Graver basis of the N-Fold matrix $[A,B]^{(n)}$. In particular, the cardinality and the bit size of $G([A,B]^{(n)})$ are bounded by a polynomial function of n.

Proposition 3.3. Fix any pair of integer matrices $A \in \mathbb{Z}^{r \times q}$ and $B \in \mathbb{Z}^{s \times q}$. Then there is a polynomial time algorithm that, given n, objective vector $c \in \mathbb{Z}^{nq}$, and non-negative integer vector $x \in \mathbb{Z}^{nq}$, solves the generalized N-Fold integer programming problem in which x is feasible.

Proposition 3.4. Fix any pair of integer matrices $A \in \mathbb{Z}^{r \times q}$ and $B \in \mathbb{Z}^{s \times q}$. Then there is a polynomial time algorithm that, given n and demand vector $b \in \mathbb{Z}^{s+nr}$, either finds a feasible point $x \in \mathbb{N}^{nq}$ to the N-Fold IP of order n, or asserts that no feasible solution exists.

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Theorem 3.5 (N-Fold IP is polynomially solvable). Fix any pair of integer matrices A, B of compatible sizes. Then there is a polynomial time algorithm that solves the generalized n-fold integer programming problem on any input n, b, c.

N-Fold augmentation algorithm

Lemma 4.1 (Steinitz Lemma). Let $v_1, ..., v_n$ be vectors with $||v_i|| \le \Delta$ for i = 1, ..., n. If $\sum_{i=1}^n v_i = 0$, then there is a reordering $\pi \in S_n$ such that for each $k \in \{1, ..., n\}$ the partial sum $p_k := \sum_{i=1}^k v_{\pi(i)}$ satisfies $||p_k|| \le n\Delta$.

It's possible (using Steinitz Lemma) to obtain a much tighter bound for the norm of the elements in the Graver basis than the ones mentioned before. This implies a restriction in the space of search for the improvement direction in the augmentation algorithm making it much faster.

Lemma 4.2 (N-Fold Graver basis bound). For all $g \in Gr(N) ||g||_1 \le L_B (2r\Delta L_B + 1)^r =: L_A \text{ where } L_B = (2s\Delta + 1)^s$

Lemma 4.3 (N-Fold augmentation algorithm complexity). The N-Fold IP can be solved in time $(nt)^2log^2(nt)\cdot \varphi(rs\Delta)^{O(r^2s+rs^2)}+LP$

[Eisenbrand, Hunkenschröder, Klein 2018]

N-Fold via LP rounding

N-Fold resolution via RLR

• N-Fold RLR complexity

The N-Fold IP restricted linear relaxation problem can be solved in time

$$O(nt \cdot log^2(nt) \cdot \varphi p(r)(s\Delta)^{O(s^2)})$$

N-Fold RLR to optimum complexity

Given an optimal vertex of an N-Fold RLR, the N-Fold IP can be solved in time

$$O(nt \cdot (rs\Delta)^{O(r^2s+s^2)})$$

[Cslovjecsek, Eisenbrand, Weismantel 2020]

N-Fold from RLR to optimum

• N-Fold proximity to RLR

Let x^* be an optimal vertex solution of a N-Fold RLR, then there exists an optimal solution z^* for the N-Fold IP verifying:

$$||z^* - x^*||_1 \le (rs\Delta)^{O(rs)}$$

[Cslovjecsek, Eisenbrand, Weismantel 2020]

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Facts for N-Fold complexity

- $|S_l| \le (rs\Delta)^{O(r^2s)}$
- $|V| + |E| \le O(n(rs\Delta)^{O(r^2s)})$
- The edge IP can be computed in time $t((r+s)\Delta)^{O(r+s)^2}$
- Longest path problem in a acyclic digraph can be solved in linear time.

N-Fold complexity

• N-Fold complexity

The N-Fold IP can be solved in time $nt(rs\Delta)^{O(r^2s+s^2)}+RLR$

[Cslovjecsek, Eisenbrand, Weismantel 2020]

Appendix A

Graver basis computation with 4ti2

Appendix B

IP resolution with Graver Basis example

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