Graver Basis

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Department of discrete optimization November 2020

$$(IP) \equiv \max\{c^t x : Ax = b, I \leq x \leq u, x \in \mathbb{Z}^n\}$$

$$A \in \mathbb{Z}^{m \times n}, \ b \in \mathbb{Z}^m, \ c \in \mathbb{Z}^n, \ I \ \text{and} \ u \ \text{lower and upper bounds for } x$$

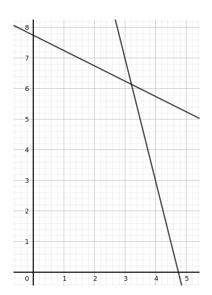
NP-Hard

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 $A \in \mathbb{Z}^{m \times n}$, $b \in \mathbb{Z}^m$, $c \in \mathbb{Z}^n$, I and u lower and upper bounds for x

- NP-Hard
- Cutting plane methods
- Lattice-basis reduction
- Dynamic programming
- Graver basis techniques

$$(P1) \equiv \left(egin{array}{c} \max 6x + y \ s.t: 4x + y \leq 19 \ x + 2y \leq 31 \ x, y \geq 0 \ x, y \in \mathbb{Z} \end{array}
ight)$$



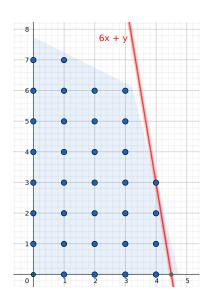
$$(P1) \equiv \begin{pmatrix} \max 6x + y \\ s.t : 4x + y \le 19 \\ x + 2y \le 31 \\ x, y \ge 0 \\ x \ne \infty \end{pmatrix}$$

- Feasible region
- Extreme points
- Optimum in extreme points



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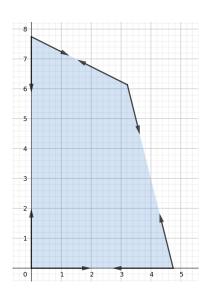
- Feasible points
- Optimum can be far from the linear relaxation solution.



$$(P1) \equiv \begin{pmatrix} \max 6x + y \\ s.t : 4x + y \le 19 \\ x + 2y \le 31 \\ x, y \ge 0 \\ x \ne \mathbb{Z} \end{pmatrix}$$

Extreme directions:
 Allow Simplex fast optimality test and finding a direction of improvement in LP.

Can we do the same for IP?



Graver Basis

- **Definition**: 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:** 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).

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Graver Basis $\equiv Gr(A)$

The Graver Basis of a given matrix A is defined as the set of integral indecomposable elements in the kernel of A.

(Initially defined as universal integral test set in [Graver 1975])

Graver Basis properties

- **Spanning**: Every integral element in ker(A) can be expressed as positive integral linear combination of elements in Gr(A).
- Optimality: 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 $l \le z + g \le u$

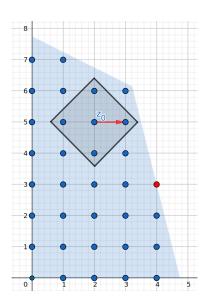
Graver Basis properties

- **Spanning**: Every integral element in ker(A) can be expressed as positive integral linear combination of elements in Gr(A).
- Optimality: 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 $l \le z + g \le u$
- Bounds: Given $A \in \mathbb{Z}^{m \times n}$ 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]

Graver Basis properties

$$(P1) \equiv \begin{pmatrix} \max 6x + y \\ s.t : 4x + y \le 19 \\ x + 2y \le 31 \\ x, y \ge 0 \\ x, y \in \mathbb{Z} \end{pmatrix}$$

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



Augmentation algorithm

General IP algorithm using Graver basis norm bound

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

$$\textit{max}\{c^tg: \textit{Ag} = 0, \textit{I} - \textit{z}_i \leq g \leq \textit{u} - \textit{z}_i, g \in \mathbb{Z}^n, ||g||_1 \leq ||\textit{Gr}(\textit{A})||\}$$

- $g^* = 0 \implies z_i$ 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 success example

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

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

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N-Fold Graver basis bound

For all $g \in \mathit{Gr}(\mathsf{N}) \ ||g||_1 \leq L_B (2r\Delta L_B + 1)^r =: L_A \ \mathsf{where} \ L_B = (2s\Delta + 1)^s$

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N-Fold augmentation algorithm complexity

The N-Fold IP can be solved in time $n^2t^2\varphi log^2nt \cdot (rs\Delta)^{O(r^2s+rs^2)} + LP$

[Eisenbrand, Hunkenschröder, Klein 2018]

N-Fold, improving even more

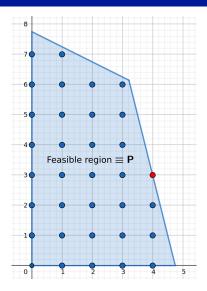


Figure 1: Linear Relaxation (LR)

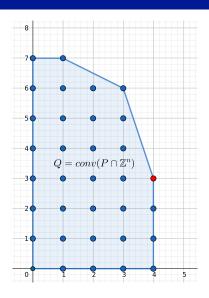


Figure 2: Restricted LR

N-Fold, improving even more

N-Fold proximity to restricted LR

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

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

N-Fold, improving even more

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N-Fold complexity

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[Cslovjecsek, Eisenbrand, Weismantel 2020]

References

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