

# Graver Basis

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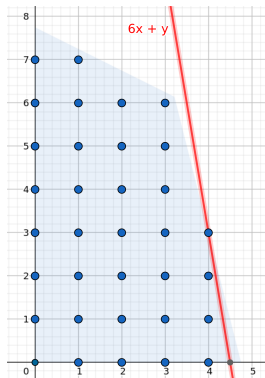
Department of discrete optimization  
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# Integer Linear Programming

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

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

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



- **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  $i \in \{1, \dots, n\}$
- **Definition:** A vector  $u \in \ker(A)$  is **indecomposable** if it is not the sum of two sign compatible and non zero elements in  $\ker(A)$ .

## 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  $A$ .

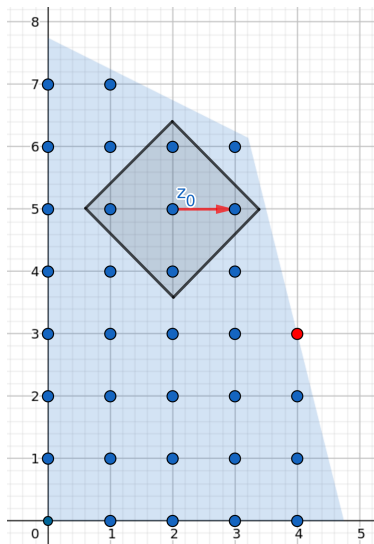
(Initially defined as *universal integral test set* in [Graver 1975])

- **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  $l \leq z + g \leq u$
- **Bounds:** Given  $A \in \mathbb{Z}^{m \times n}$  and  $\Delta$  an upper bound for the absolute value of each component of  $A$ , for every  $g \in Gr(A)$ :
  - $\|g\|_1 \leq m^{m/2} \Delta^m \cdot (n - m)$  [Onn 2010]
  - $\|g\|_1 \leq (2m\Delta + 1)^m$  [Eisenbrand, Hunkenschröder, Klein 2018]

# Graver Basis properties

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

- 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:

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

- $g^* = 0 \implies z_i$  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  $\lambda$  respecting the bounds.

[Hemmecke, Onn, Romanchuk 2013]

# N-Fold, a success example

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

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

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### N-Fold Graver basis bound

For all  $g \in Gr(N)$   $\|g\|_1 \leq L_B(2r\Delta L_B + 1)^r =: L_A$  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  $(nt)^2 \log^2(nt) \cdot \varphi(rs\Delta)^{O(r^2s+rs^2)} + LP$

[Eisenbrand, Hunkenschröder, Klein 2018]

# N-Fold, improving even more

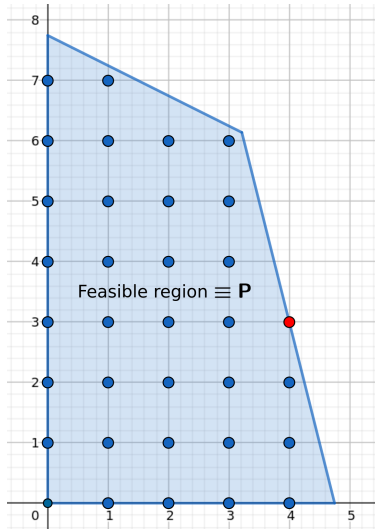


Figure 1: Linear Relaxation (LR)

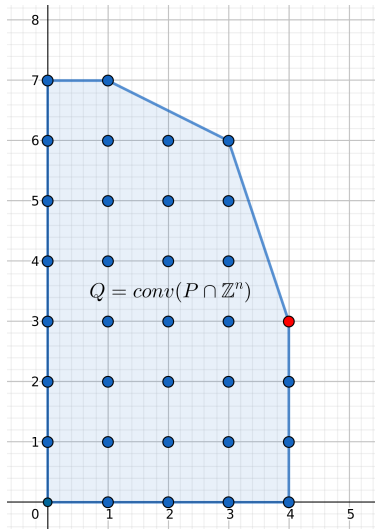


Figure 2: Restricted LR (RLR)

# N-Fold, resolution steps

## N-Fold RLR complexity

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

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

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## 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:

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

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Let  $\gamma := (rs\Delta)^{O(rs)}$ . For every  $1 \leq \ell \leq n$ :

$$\begin{aligned}\|x^* - z\|_1 \leq \gamma &\implies \sum_{i=1}^{\ell} \|x_i^* - z_i\|_1 \leq \gamma \\ &\implies \sum_{i=1}^{\ell} \|A_i(x_i^* - z_i)\|_{\infty} \leq \Delta\gamma = \gamma\end{aligned}$$

Therefore, for every feasible point  $z$  verifying the proximity bound we have:

$$\sum_{i=1}^{\ell} A_i x_i^* - \gamma \leq \sum_{i=1}^{\ell} A_i z_i \leq \sum_{i=1}^{\ell} A_i x_i^* + \gamma$$

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We now define the set  $S_{\ell}$  as the set of  $y \in \mathbb{Z}^r$  such that:

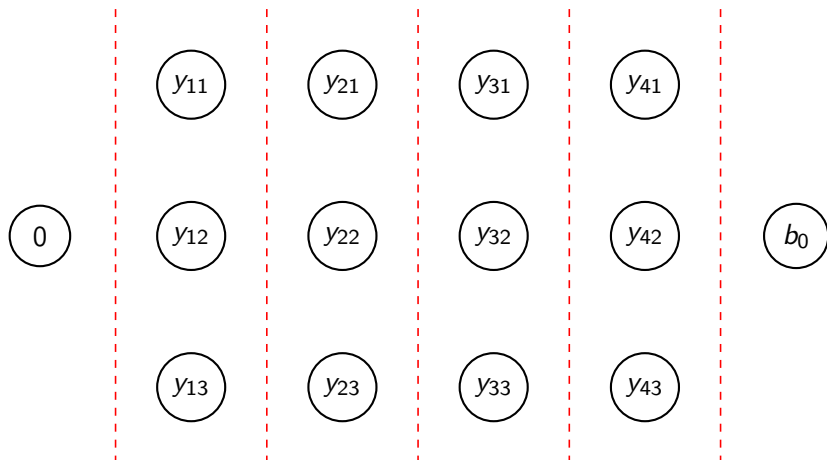
$$\sum_{i=1}^{\ell} A_i x_i^* - \gamma \leq y \leq \sum_{i=1}^{\ell} A_i x_i^* + \gamma$$



# N-Fold, from RLR to optimum

We can construct a weighted directed acyclic graph  $G(V,E)$  with vertices:

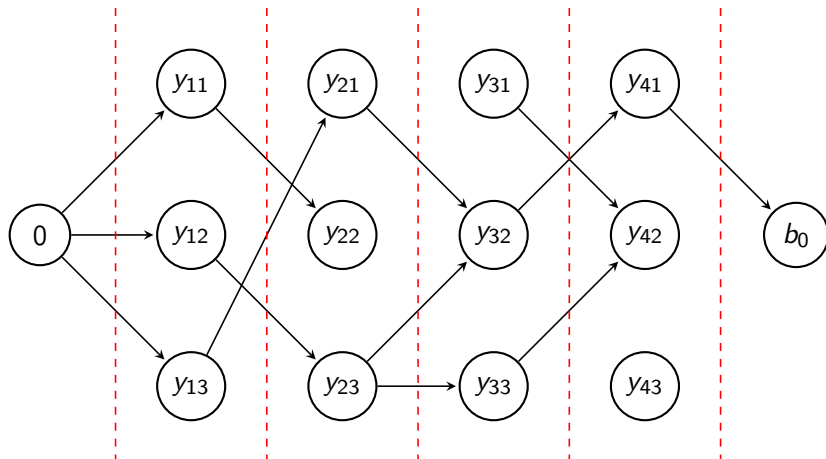
$$V = \{(\ell, y)/y \in S_\ell\} \cup \{(0, 0), (n, b_0)\}$$



# N-Fold, from RLR to optimum

Weighted edges between  $(\ell - 1, y)$  and  $(\ell, y')$  if the problem is feasible:

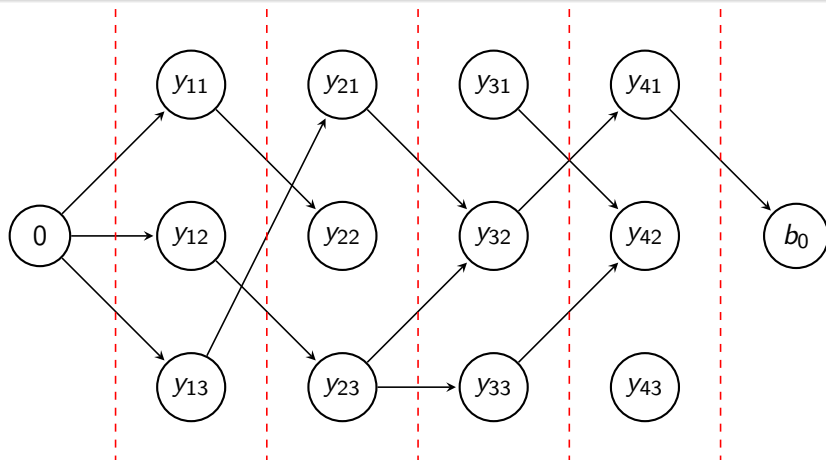
$$\max\{c_\ell^t x : A_\ell x = (y' - y), B_\ell x = b_\ell, l_\ell \leq x \leq u_\ell\}$$



# N-Fold, from RLR to optimum

## Correspondence

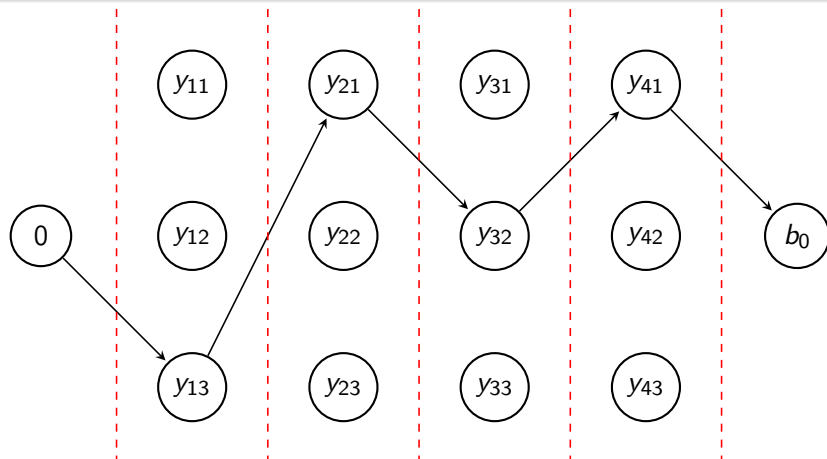
A longest path from  $(0,0)$  to  $(n, b_0)$  in  $G(V,E)$  corresponds to an optimal solution of the N-Fold integer program.



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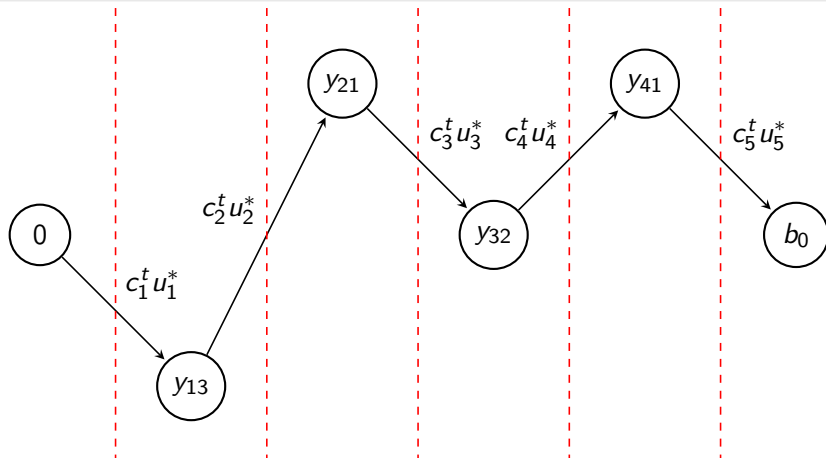
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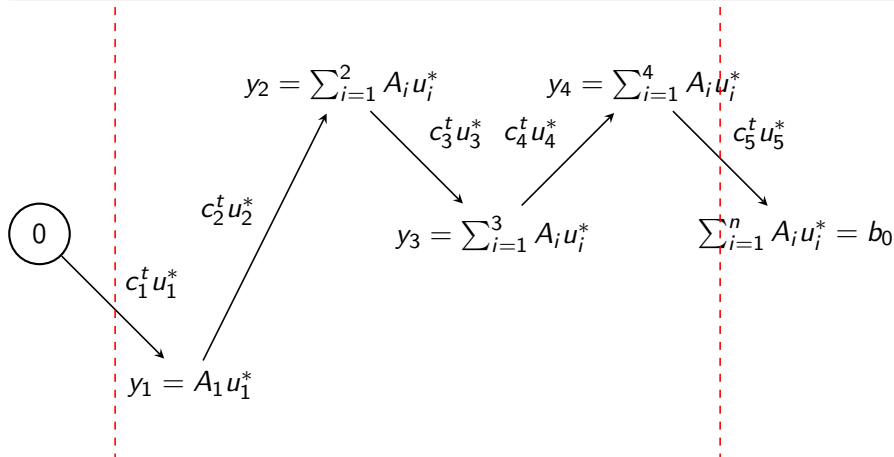
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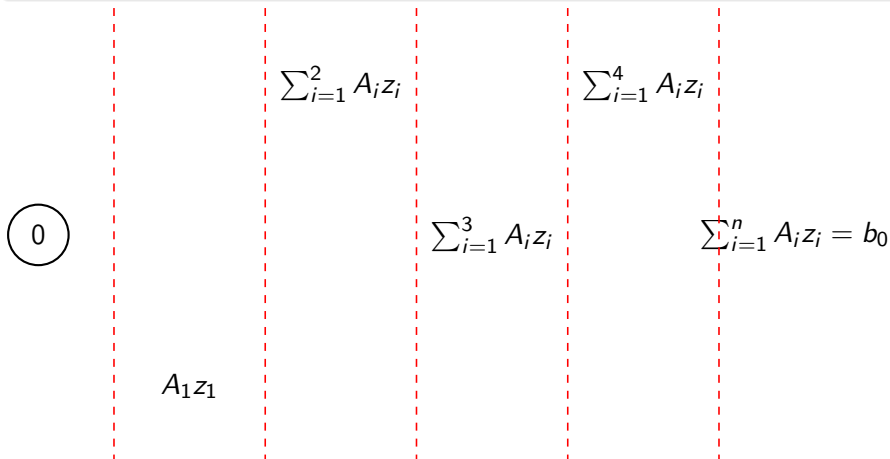
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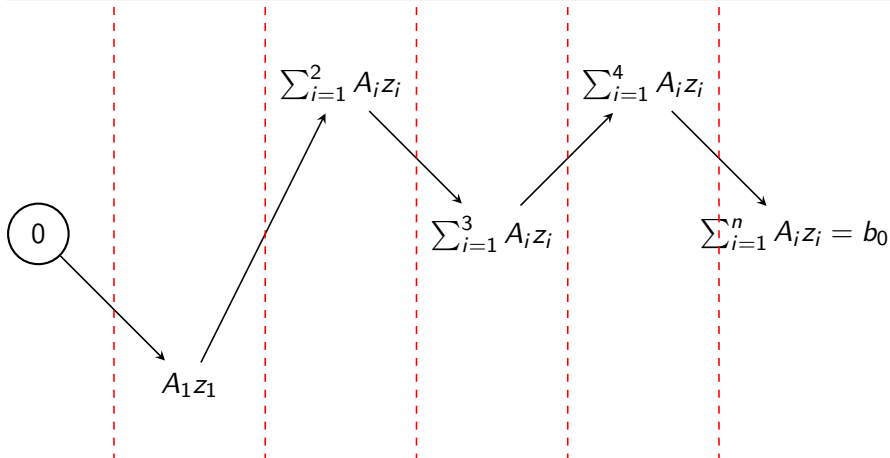
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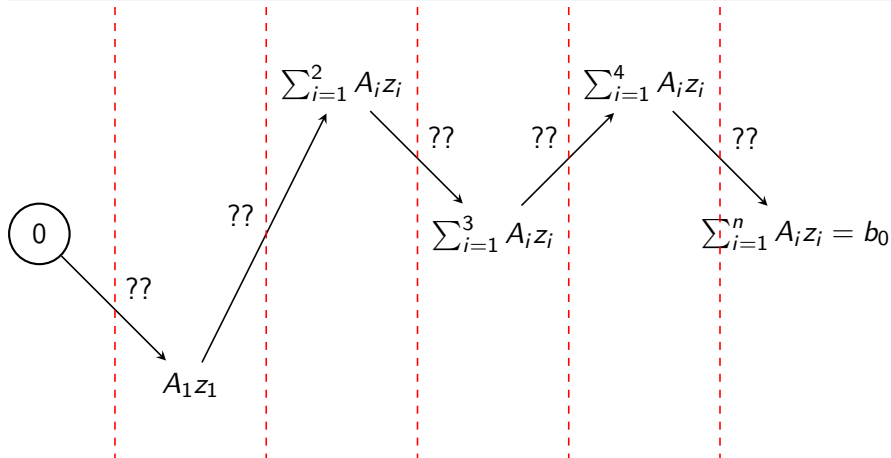




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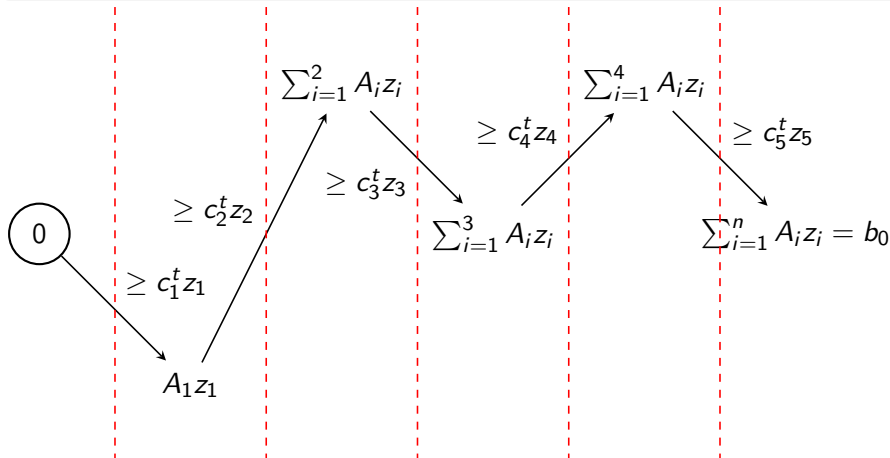
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# N-Fold, from RLR to optimum

- $|S_I| \leq (rs\Delta)^{O(r^2s)}$
- $|V| + |E| \leq O(n(rs\Delta)^{O(r^2s)})$
- 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.

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## 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]

- [1] Jack E. Graver. “On the foundations of linear and integer linear programming I”. In: Syracuse University, New York, U.S.A., 1975.
- [2] Kim-Manuel Klein Friedrich Eisenbrand Christoph Hunkenschröder. “Faster Algorithms for Integer Programs with Block Structure”. In: École polytechnique fédérale de Lausanne, Switzerland, 2018.
- [3] Robert Weismantel Jana Cslovjecsek Friedrich Eisenbrand. “N-fold integer programming via LP rounding”. In: École polytechnique fédérale de Lausanne, Switzerland, 2020.