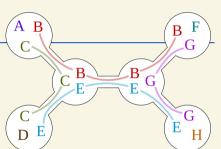
FPT via Linear Programming II A C DM898: Parameterized Algorithms Lars Rohwedder



Today's lecture

 $\ensuremath{\mathsf{FPT}}$ algorithms for classes of integer linear programs

- Parameter: number of variables
- Parameter: number of constraints

Can we design FPT algorithms for integer linear programs?

Motivation: since ILPs are very expressive, we would be able to derive many results from such an algorithm

But how to parameterize ILPs?

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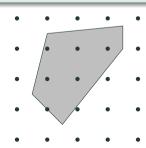
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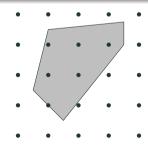
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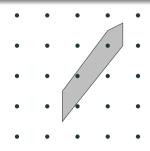
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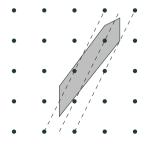
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Details are intricate and we do not cover them here



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Eisenbrand-Weismantel algorithm

Eisenbrand and Weismantel [2018] showed that integer programs with m linear equality constraints, where all coefficients lie in $\{-\Delta,\ldots,\Delta\}$, and (possibly many) variables with upper and lower bounds can be solved in time $(m\Delta)^{O(m^2)}\cdot |I|^{O(1)}$.

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One may ask whether Eisenbrand and Weismantel's algorithm can be improved to FPT time in m alone (like Lentra's algorithm that does not need parameter Δ). This is not possible, because m=1 is already NP-hard (e.g. via reduction from Subset-Sum).

We will prove Eisenbrand and Weismantel's result in the following.

Eisenbrand-Weismantel algorithm

Proximity

Problem statement. Given coefficients of the objective $c \in \mathbb{Z}^n$, a matrix $A \in \{-\Delta, \dots, \Delta\}^{m \times n}$ (encoding the coefficients of the constraints in the m rows), right-hand side $b \in \mathbb{Z}^m$ and lower and upper bounds

$$\ell_i \in \mathbb{Z} \cup \{-\infty\}, u_i \in \mathbb{Z} \cup \{\infty\}, \ i \in \{1,2,\dots,n\},$$
 solve
$$\min \ c^\mathsf{T} x$$

$$Ax = b$$

$$\ell_i \le x_i \le u_i, \quad x_i \in \mathbb{Z}$$

for all
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$$\min c^{\mathsf{T}} x$$

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for all
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Proximity theorem

Assume that the ILP above is feasible and bounded. Let x^* be a optimal solution to the LP relaxation of the ILP above with at most m non-integral variables¹. Then there exists some optimal integer solution x with

$$||x - x^*||_1 = \sum_{i=1}^n |x_i - x_i^*| \le (2m^2\Delta + 1)^m + m =: \text{prox}$$

$$\left[\quad \text{in particular:} \quad \|x - \lfloor x^* \rfloor\|_1 = \sum_{i=1}^n |x_i - \lfloor x_i^* \rfloor| \leq (2m^2 \Delta + 1)^m + 2m =: \operatorname{prox'} \quad \right]$$

We will prove this statement next lecture and first show how it can be used algorithmically.

 $^{^{1}}$ such a solution always exists and can be computed in polynomial time. More details next lecture.

Dynamic program

We proceed similar to the Knapsack dynamic program based on "dominance"

$\mathsf{ILP}(n, A, b, c, \ell, u)$

- \bullet compute optimum x^* to LP relaxation with $\leq m$ non-integral variables
- $\mathcal{T} \leftarrow \{(0,0,0)\}$ // set of undominated (objective, right-hand side, distance-to- $\lfloor x^* \rfloor$) triples obtainable
- $\bullet \ \text{ for } i \in \{1,2,\ldots,n\}$
 - $\mathcal{T}' \leftarrow \mathcal{T}, \, \mathcal{T} \leftarrow \emptyset$
 - for x_i in $\{\max\{\ell_i, \lfloor x_i^* \rfloor \mathsf{prox'}\}, \ldots, \min\{u_i, \lfloor x_i^* \rfloor + \mathsf{prox'}\}\}$
 - $\mathcal{T} \leftarrow \mathcal{T} \cup \{(C + c_i x_i, B + A_i x_i, k + |\lfloor x_i^* \rfloor x_i|) \mid (C, B, k) \in \mathcal{T}'\}$ // A_i is the ith column of A_i
 - for $(C, B, k), (C', B', k') \in \mathcal{T}$ with C < C', B = B', k = k'
 - $\bullet \ \mathcal{T} \leftarrow \mathcal{T} \setminus (C', B', k')$
 - $\bullet \ \, {\rm for} \,\, (C,B,k) \in \mathcal{T} \,\, {\rm with} \,\, k > {\rm prox'} \,\,$
 - $\mathcal{T} \leftarrow \mathcal{T} \setminus (C, B, k)$
- return $\min\{C \mid (C, B, k) \in \mathcal{T}, B = b\}$

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Correctness. By induction we have that for each $B \in \mathbb{Z}^m$, $k \in \mathbb{Z}_{\geq 0}$, \mathcal{T} contains (C, B, k) after iteration i if and only if $k \leq \text{prox}'$ and the following minimum exists with

$$C = \min \left\{ \sum_{j=1}^{i} c_j x_j \mid x_i \in \{\ell_i, \ell_i + 1, \dots, u_i\} \forall j \in \{1, 2 \dots, i\}, \sum_{j=1}^{i} A_j x_j = B, \sum_{j=1}^{i} |x_j - \lfloor x_j^* \rfloor| = k \right\}$$

One iteration of main loop

- $\mathcal{T}' \leftarrow \mathcal{T}, \, \mathcal{T} \leftarrow \emptyset$
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$$||B - (\underbrace{A_1 \lfloor x_1^* \rfloor + \dots + A_i \lfloor x_i^* \rfloor})||_{\infty} \le k\Delta.$$

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• Thus, at the beginning of an iteration $|\mathcal{T}| \leq (2\text{prox}'\Delta + 1)^m \cdot \text{prox}' \leq (m\Delta)^{O(m^2)}$

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• Thus, at the beginning of an iteration $|\mathcal{T}| \leq (2\mathsf{prox}'\Delta + 1)^m \cdot \mathsf{prox}' < (m\Delta)^{O(m^2)}$

Total running time: #iterations $\cdot (m\Delta)^{O(m^2)} \le n \cdot (m\Delta)^{O(m^2)}$

Other FPT results for ILP

Unbounded variables

Consider the Eisenbrand-Weismantel setting, but with all variables being non-negative integers:

$$\min c^\mathsf{T} x$$

$$Ax = b$$

$$x_i \in \mathbb{Z}_{\geq 0}$$
 for all $i = 1, 2, \dots, n$

The Eisenbrand-Weismantel algorithm from before has a running time of $n(m\Delta)^{O(m^2)}$. In this case, faster algorithms are known.

Fastest known algorithm runs in time $(\sqrt{m}\Delta)^{2m}+O(nm)$, due to Jansen and Rohwedder [2019].

Block structures

Consider $A_1, \ldots, A_n, B_1, \ldots, B_n \in \{-\Delta, \ldots, \Delta\}^{k \times k}$ that form one of the following two block structures.

$$\min c^{\mathsf{T}} x \qquad \qquad \min c^{\mathsf{T}} x$$

$$\begin{pmatrix} A_1 & A_2 & \cdots & A_n \\ B_1 & & & & \\ & B_2 & & & \\ & & \ddots & & \\ & & B_n \end{pmatrix} x = b \qquad \qquad \begin{pmatrix} A_1 & B_1 \\ A_2 & & B_2 \\ \vdots \\ A_n & & & B_n \end{pmatrix} x = b$$

$$\ell_i \leq x_i \leq u_i \quad \text{for all } i = 1, 2, \dots, kn$$

$$x_i \in \mathbb{Z}$$

$$\min c^{\mathsf{T}} x \qquad \qquad \min c^{\mathsf{T}} x$$

$$\begin{pmatrix} A_1 & B_1 \\ A_2 & & B_2 \\ \vdots \\ A_n & & & B_n \end{pmatrix} x = b$$

$$\ell_i \leq x_i \leq u_i \quad \text{for all } i = 1, 2, \dots, k(n+1)$$

Both classes of ILPs have FPT algorithms in parameters k and Δ . Currently fastest are due to Cslovjecsek, Eisenbrand, Hunkenschröder, Rohwedder, Weismantel [2021] and Klein [2020].

In practice, these classes are also well solvable via decomposition methods (Dantzig-Wolfe decomposition and Bender's decomposition), which we do not detail here.

Summary

A number of FPT results for integer linear programs are known, which:

- Explain to some extend the good empirical behavior and utility of LP relaxations we can see in commercial ILP solvers
- Have some interesting applications for concrete problems (see e.g. exercises)
- Disclaimer: Many (perhaps most) FPT results in literature are based on entirely different techniques and cannot be derived from generic ILP results

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PARAMLP



- ERC-funded project PARAMLP: Parameterized Algorithms and Polyhedra
- Funded with $\approx 11\,000\,000$ DKK
- Awarded to me in 2025, hiring PhD students and PostDocs soon
- Possibly also thesis projects in this field