

Integer Linear-Exponential Programming

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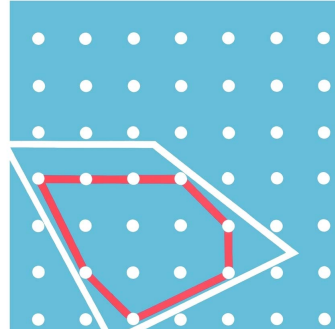
SC² Workshop 2025

Based on an **ICALP'24** paper with Dmitry Chistikov and Mikhail Starchak,
and a **preprint** with S Hitarth and Guru Shabadi

Integer Linear Programming (ILP)

$$\begin{aligned} &\text{maximize} && a_d \cdot x_d + \cdots + a_1 \cdot x_1 + a_0 \\ &\text{subject to} && \begin{bmatrix} * & \cdots & * \\ \vdots & \ddots & \vdots \\ * & \cdots & * \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ \vdots \\ x_d \end{bmatrix} \leq \begin{bmatrix} * \\ \vdots \\ * \end{bmatrix} \\ &&& x_1, \dots, x_d \in \mathbb{N} \end{aligned}$$

THEORY OF LINEAR AND INTEGER PROGRAMMING



ALEXANDER SCHRIJVER

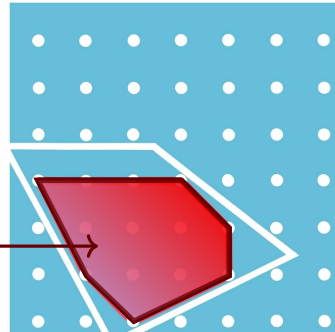
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$$\begin{aligned} &\text{maximize } x + y \text{ subject to} && \left. \begin{aligned} y &\leq 3 \\ x + y &\leq 6 \\ x &\leq 4 \\ x - 2 \cdot y &\leq 2 \\ -x - y &\leq -2 \\ -2 \cdot x - y &\leq -3 \end{aligned} \right\} \end{aligned}$$

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Integer Linear-Exponential Programming (ILEP)

Linear-exponential term: $a_0 + \sum_{i=1}^d \left(a_i \cdot x_i + b_i \cdot 2^{x_i} + \sum_{j=1}^d c_{i,j} \cdot (x_i \bmod 2^{x_j}) \right)$

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Integer linear-exponential programming:

maximize $\tau(\mathbf{x})$

subject to $\rho_i(\mathbf{x}) \leq 0$ for $i \in [1..n]$

$\mathbf{x} \in \mathbb{N}^d$

where $\tau, \rho_1, \dots, \rho_n$ are linear-exponential terms.

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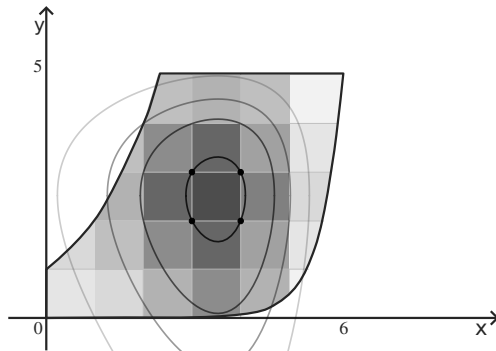
$x \in \mathbb{N}^d$

where $\tau, \rho_1, \dots, \rho_n$ are linear-exponential terms.

Question: Is there an algorithm to find an (optimal) solution?

An example

$$\begin{aligned} &\text{maximize } \tau(x, y) := 8x + 4y - (2^x + 2^y) \\ &\text{subject to } \varphi(x, y, z) := \begin{aligned} &y \leq 5 \\ &y \leq 2^x \\ &2^z \leq 2^{16}y \\ &z = 3 \cdot x \end{aligned} \end{aligned}$$



Setting (x, y) to any point in $\{3, 4\} \times \{2, 3\}$ yields an optimal solution.

Some stuff ILEP can express

x is of bit length y : $2^y \leq x < 2 \cdot 2^y$

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x is a tower of 2s of height n : $\exists x_1, \dots, x_n : x = 2^{x_1} \wedge \left(\bigwedge_{i=1}^{n-1} x_i = 2^{x_{i+1}} \right) \wedge x_n = 1$

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the y th digit of x is 1: $\exists z : z = y + 1 \wedge (x \bmod 2^z) - (x \bmod 2^y) \geq 1$

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	(msd)		y			(lsd)
x :	*	*	*	*	*	*

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	(msd)		y			(lsd)
x :	0	0	*	*	*	*

$$x \bmod 2^{y+1}$$

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	(msd)		y				(lsd)
x :	0	0	*	0	0	0	0

$$(x \bmod 2^{y+1}) - (x \bmod 2^y)$$

Some stuff ILEP can express

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In relation with regular languages

Can express all regular languages of polynomial growth (Haase and Różycki, '21):

$$v_0 w_1^* v_1 w_2^* v_2 \dots v_{k-1} w_k^* v_k \quad \text{where all } v_i, w_i \text{ are in } \{0, 1\}^*$$

Can express non-regular languages (see all prior examples)

Cannot express all regular languages (Starchak, '24): ILEP cannot express $\{01, 10\}^*$

A comparison between ILP and ILEP

When solutions are encoded in binary:

ILP	ILEP
<hr/>	
<i>When feasible, is there a solution of polynomial size?</i>	
✓	✗
<i>When an optimum exists, is there one of polynomial size?</i>	
✓	✗
<i>Are optimal solutions located “near” the boundary of the feasible region?</i>	
✓	✗
<i>Can solutions be checked in polynomial time?</i>	
✓	✓

A comparison between ILP and ILEP

When we give solutions in a **compressed form**:

ILP	ILEP
<hr/>	
<i>When feasible, is there a solution of polynomial size?</i>	
✓	✓ (ICALP'24)
<i>When an optimum exists, is there one of polynomial size?</i>	
✓	✓ (preprint)
<i>Are optimal solutions located “near” the boundary of the feasible region?</i>	
✓	✗ *unproblematic*
<i>Can solutions be checked in polynomial time?</i>	
✓	✓ *kind of*

Integer Linear-Exponential Straight-Line Programs (ILESPLs)

Linear-Exponential Straight-Line Program (LESPL): A sequence of assignments

$$x_i \leftarrow 0, \quad x_i \leftarrow x_j + x_k, \quad x_i \leftarrow 2^{x_j}, \quad x_i \leftarrow a \cdot x_j, \quad \text{where } a \in \mathbb{Q} \text{ and } j, k < i.$$

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ILESPL: An LESPL in which **all** variables evaluate to an integer.

Theorem (Hitarth, M., Shabadi. *preprint*)

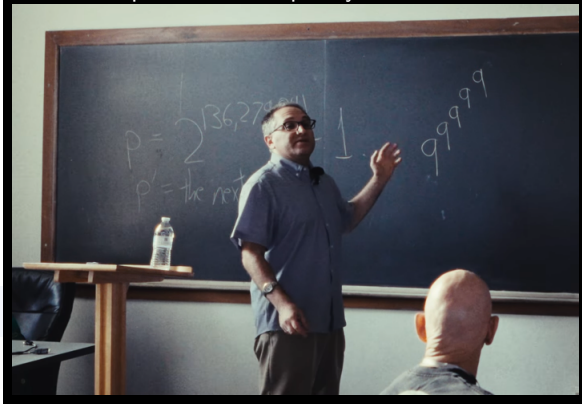
If an ILEP has optimal solutions, then one is encoded by a polynomial-size ILESPL.

(Proven by giving an algorithm for constructing optimal solutions.)

Integer Linear-Exponential Straight-Line Programs (ILESPLs)

Linear-Exponential Straight-Line Program (LESPL): A sequence of assignments

Prof. Scott Aaronson: Why Philosophers Should Care About Computational Complexity @ UT Austin



$a \cdot x_j$, where $a \in \mathbb{Q}$ and $j, k < i$.

$$x_3 \leftarrow \frac{1}{2}, \quad x_4 \leftarrow \sqrt{2}$$

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nt)

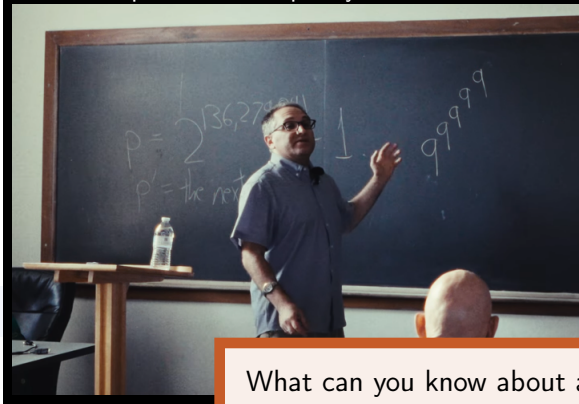
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What can you know about an ILESPLP? What can you do with it?

(Proven by giving an algorithm for computing \log_2 of a rational number.)

Recognizing that an LESLP is an ILESLP

An LESLP is an ILESLP whenever:

- In assignments $x \leftarrow 2^y$, the variable y evaluates to a non-negative integer
- In assignments $x \leftarrow \frac{m}{g} \cdot y$, the variable y evaluates to a multiple of $\frac{g}{\gcd(m,g)}$

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NAT_{ILESLP}

Input: an ILESLP σ .

Question: does the last expression in σ evaluate to a non-negative integer?

DIV_{ILESLP}

Input: an ILESLP σ and a positive integer h .

Question: does the last expression in σ evaluate to a multiple of h ?

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DIV_{ILESLP}

In polynomial time with an oracle for factoring

Input: an ILESLP σ and a positive integer h .

(not in BPP, unless some crypto assumption breaks)

Question: does the last expression in σ evaluate to a multiple of h ?

Recognizing that an LESLP is an ILESLP

An LESLP is an ILESLP whenever:

The set of all ILESLPs is not recognizable in polynomial time!

However, given an ILESLP σ , there is a small set $\mathbb{P}(\sigma)$ of small primes such that

$$\{(\sigma, \mathbb{P}(\sigma)) : \sigma \text{ is an ILESLP}\}$$

is recognizable in polynomial time.

We use $(\sigma, \mathbb{P}(\sigma))$ as certificates of solutions.

Input: an ILESLP σ and a positive integer h .

(not in BPP, unless some crypto assumption breaks)

Question: does the last expression in σ evaluate to a multiple of h ?

Checking solutions

Input: An integer linear-exponential program φ , and $(\sigma, \mathbb{P}(\sigma))$ with σ ILESLP.

Question: Is σ a solution to φ ?

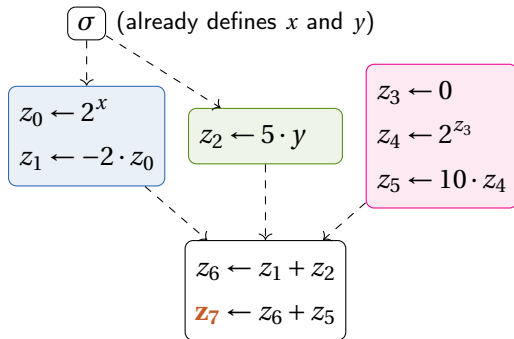
Input: An integer linear-exponential program φ , and $(\sigma, \mathbb{P}(\sigma))$ with σ ILESLP.

Question: Is σ a solution to φ ?

For inequalities without $(x \bmod 2^y)$:

$$-2 \cdot 2^x + 5 \cdot y + 10 \geq 0$$

1. Append $-2 \cdot 2^x + 5 \cdot y + 10$ to σ
2. Query the algorithm for $\text{NAT}_{\text{ILESLP}}$.



Input: An integer linear-exponential program φ , and $(\sigma, \mathbb{P}(\sigma))$ with σ ILESLP.

Question: Is σ a solution to φ ?

Expressions $(x \bmod 2^y)$ can be eliminated:

Mod computation

Input: an ILESLP σ , two variables x and y in σ , and the set $\mathbb{P}(\sigma)$.

Output: an ILESLP σ' whose last expression evaluates to $(\sigma(x) \bmod 2^{\sigma(y)})$.

Algorithm runs in **polynomial time**.

If $\mathbb{P}(\sigma)$ is not provided in input, the algorithm requires an **integer factoring oracle**.

Comparing values of the objective function

So far we have seen:

- ✓: polynomial-size certificates encoding of solutions and optimal solutions.
- ✓: certificates checkable in polynomial time.

What about the objective function?

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Is $\varphi \wedge \tau \geq 50$
satisfiable?



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$[0..49]$: range in which optimum lie

No.



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φ : integer linear program

$[0..49]$: range in which optimum lie

Is $\varphi \wedge \tau \geq 24$
satisfiable?



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φ : integer linear program

[24..49] : range in which optimum lie

Yes.



Comparing values of the objective function

So far we have seen:

- ✓: polynomial-size certificates encoding of solutions and optimal solutions.
- ✓: certificates checkable in polynomial time.

What about the objective function?

In **ILEP** we do not know how to perform binary search on ILESLPs.

We can still **compare** values of the objective function τ in polynomial time:

For $(\sigma, \mathbb{P}(\sigma))$ and $(\eta, \mathbb{P}(\eta))$ solutions:

- build an ILESLP for $\tau(\sigma) - \tau(\eta)$
- call the algorithm for $\text{NAT}_{\text{ILESLP}}$.

This leads to a FNP^{NP} upper bound for ILEP.

Recap and an open problem

Key properties of ILEP:

1. *If there are (optimal) solutions, one can be represented with a short ILESLP.*
 2. *Checking if an ILESLP is a solution*
 3. *Comparing values of the objective on two ILESLPs*
- } in $P^{\text{FACTORING}}$...

...or in P if we add a small set of primes to the certificates

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Open problem: binary search on ILESLPs.

Let S be the set of all ILESLPs of size at most k . Is there an algorithm with runtime polynomial in k that, given as input $L, U \in S$, computes $M \in S$ such that the size of both sets $S_1 := \{\sigma \in S : L \leq \sigma \leq M\}$ and $S_2 := \{\sigma \in S : M \leq \sigma \leq U\}$ is in $\Omega(\#S_1 + \#S_2)$?

Above, “ $\sigma_1 \leq \sigma_2$ ” compares the value of the last expressions of σ_1 and σ_2 .

An algorithm for finding a solution

$\theta(x, y)$: ordering $2^x \geq 2^y \geq \dots \geq 2^{x_0} = 1$

$\varphi(x, y, r)$: linear-exponential program

Input: $\varphi(x)$ linear-exponential program

As a preliminary step:

- guess an ordering θ :

$$2^x \geq 2^y \geq \dots \geq 2^{x_0} = 1$$

- initialize a vector $r = \emptyset$ of remainder variables

In the invariant of the loop that follows, $r < 2^x$
and variables r do not occur in exponentials

- let $y = (y, \dots, x_0)$

An algorithm for finding a solution

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$\varphi(x, y, r)$: linear-exponential program

Step 1

$\gamma(q_x, q, u)$: linear program

$\psi(y, r_x, r')$: linear-exponential program

Elimination of x : Step I

Divisions by 2^y

An algorithm for finding a solution

Side conditions: $u = 2^{x-y}$ and $x = q_x \cdot 2^y + r_x$

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Elimination of x : Step I

Divisions by 2^y

Change of variables:

$$x = q_x \cdot 2^y + r_x \quad \text{and} \quad r = q \cdot 2^y + r'$$

+ constraints $r_x < 2^y$ and $r' < 2^y$ (loop invariant)

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Non-deterministic rewriting: guess a small $\ell \in \mathbb{Z}$

$$f(q_x, q, u) \cdot 2^y + g(y, r_x, r') \leq 0$$

↓

$$f(q_x, q, u) + \ell \leq 0 \wedge (\ell - 1) \cdot 2^y \leq g(y, r_x, r') < \ell \cdot 2^y$$

An algorithm for finding a solution

$\theta(x, y)$: ordering $2^x \geq 2^y \geq \dots \geq 2^{x_0} = 1$

$\varphi(x, y, r)$: linear-exponential program

Step I

$\gamma(q_x, q, u)$: linear program

$\psi(y, r_x, r')$: linear-exponential program

Step II ⚠

$\gamma'(q_x, u)$: linear program

Side conditions: $u = 2^{x-y}$ and $x = q_x \cdot 2^y + r_x$

Elimination of x : Step II

Elimination of q

γ is linear, so we can use quantifier elimination

$$\exists q \gamma(q_x, q, u) \iff \bigvee_{\beta} \gamma'_{\beta}(q_x, u)$$

An algorithm for finding a solution

Side conditions: $u = 2^{x-y}$ and $x = q_x \cdot 2^y + r_x$

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Step I

$\gamma(q_x, q, u)$: linear program

$\psi(y, r_x, r')$: linear-exponential program

Step II ⚠

$\gamma'(q_x, u)$: linear program

Elimination of x : Step II

Elimination of q

γ is linear, so we can use quantifier elimination

$$\exists q \gamma(q_x, q, u) \iff \bigvee_{\beta} \gamma'_{\beta}(q_x, u)$$

⚠: For $\exists\text{PrA}$, only known to be in NEXPTIME.

⚠: It may lose all optimal solutions.

An algorithm for finding a solution

Side conditions: $u = 2^{x-y}$ and $x = q_x \cdot 2^y + r_x$

$\theta(x, y)$: ordering $2^x \geq 2^y \geq \dots \geq 2^{x_0} = 1$

$\varphi(x, y, r)$: linear-exponential program

Step I

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$\psi(y, r_x, r')$: linear-exponential program

Step II ⚠

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Elimination of x : Step II

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$$\exists q \gamma(q_x, q, u) \iff \bigvee_{\beta} \gamma'_{\beta}(q_x, u)$$

⚠: For $\exists\text{PrA}$, only known to be in NEXPTIME.

✓: Can be improved to NP (ICALP'24)

⚠: It may lose all optimal solutions.

✓: No, if we decompose the search space (preprint)

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Step II ⚠

$\gamma'(q_x, u)$: linear program

Step III

$\gamma''(q_x)$: linear program

$\psi''(y, r_x)$: linear-exponential program

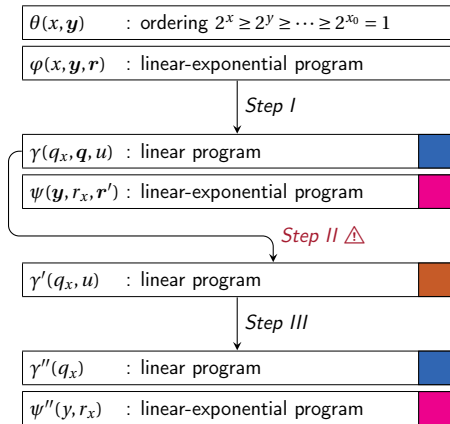
Elimination of x : Step III

Elimination of u and x

Apply side conditions to $\gamma'(q_x, u)$

$$a \cdot 2^{(q_x \cdot 2^y + r_x - y)} + b \cdot q_x + c \leq 0$$

An algorithm for finding a solution



Elimination of x : Step III

Elimination of u and x

Apply side conditions to $\gamma'(q_x, u)$

$$a \cdot 2^{(q_x \cdot 2^y + r_x - y)} + b \cdot q_x + c \leq 0$$

For an extremely small $L \in \mathbb{N}$, **non-det. rewrite** as

$$\blacksquare \quad q_x \cdot 2^y + r_x - y > L \wedge a \leq 0 \quad \text{or}$$

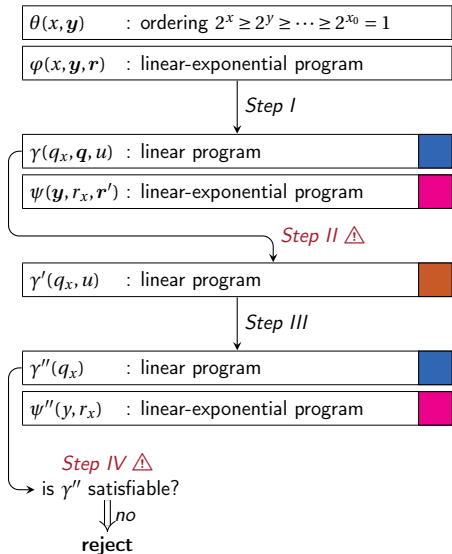
$$\blacksquare \quad q_x \cdot 2^y + r_x - y = \ell \wedge a \cdot 2^\ell + q_x + c \leq 0$$

for some guessed $\ell \in [0..L]$.

Follow-up with the split performed in Step I:

$$q_x \cdot 2^y + r_x - y - L > 0$$

An algorithm for finding a solution



Elimination of x : Step IV

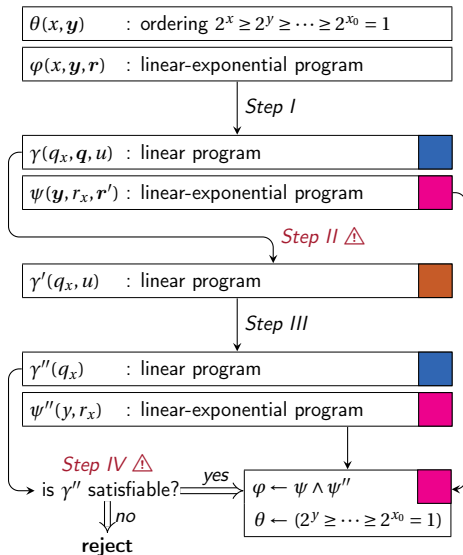
Only q_x is left

$\gamma''(q_x)$ is now completely independent from $\psi''(y, r_x)$

✓: For feasibility, simply check if γ'' is satisfiable

⚠: For optimization, decomposition not needed, but one has to choose a value for q_x carefully

An algorithm for finding a solution



Elimination of x : Step IV

Only q_x is left

$\gamma''(q_x)$ is now completely independent from $\psi''(y, r_x)$

✓: For feasibility, simply check if γ'' is satisfiable

⚠: For optimization, decomposition not needed, but one has to choose a value for q_x carefully

Continue with the next iteration!

All equations (e.g., $x = q_x \cdot 2^y + r$) are building an ILES LP!

Step II: Integer linear programming in NP through symbolic methods

Gaussian elimination

Input: an integer linear program $\varphi(x, z)$ with only inequalities

Ensure: all the variables in x are eliminated from the inequalities of φ

$\ell \leftarrow 1$

for x in x occurring in an inequality of φ **do**

$(a \cdot x + \tau = 0) \leftarrow$ an arbitrary inequality in φ , with a non-zero

$s \leftarrow$ guess an integer in $[0..|a| \cdot \text{mod}(\varphi) - 1]$

$\tau \leftarrow \tau + s$

$\varphi \leftarrow \varphi[\frac{-\tau}{a} / x]$

divide each equality in φ by ℓ

$\varphi \leftarrow \varphi \wedge (a \mid \tau)$

$\ell \leftarrow a$

return φ

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$$\varphi[\frac{-\tau}{a} / x]: \quad b \cdot x + \quad \rho = 0$$

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$$\varphi[\frac{-\tau}{a} / x]: \quad -b \cdot \tau + a \cdot \rho = 0$$

Note: $(-b \cdot \tau + a \cdot \rho) = \det \begin{bmatrix} a & \tau \\ b & \rho \end{bmatrix}$

Step II: Integer linear programming in NP through symbolic methods

Bareiss's algorithm

Input: an integer linear program $\varphi(x, z)$ with only inequalities

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return φ

Desnanot–Jacobi identity:

$$\begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} \begin{vmatrix} a_{11} & a_{13} \\ a_{21} & a_{23} \end{vmatrix} \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11} \cdot \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}$$

Step II: Integer linear programming in NP through symbolic methods

Presburger's quantifier elimination

Input: an integer linear program $\varphi(x, z)$ with only inequalities

Ensure: all the variables in x are eliminated from the inequalities of φ

$\ell \leftarrow 1$

for x in x occurring in an inequality of φ **do**

$(a \cdot x + \tau \leq 0) \leftarrow$ **guess an inequality** in φ , with a non-zero

$s \leftarrow$ **guess an integer** in $[0..|a| \cdot \text{mod}(\varphi) - 1]$

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$\varphi \leftarrow \varphi[\frac{-\tau}{a} / x]$

divide each inequality in φ by ℓ

$\varphi \leftarrow \varphi \wedge (a \mid \tau)$

$\ell \leftarrow a$

return φ

Consider $\tau \leq a \cdot x \leq \tau'$ with $a > 0$.

"between τ and τ' there is a multiple of a "

One such multiple has the form $a \cdot x = \tau + s$.

Step II: Integer linear programming in NP through symbolic methods

Presburger's quantifier elimination meets Bareiss's algorithm

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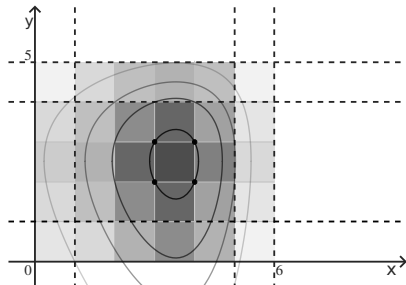
$\ell \leftarrow a$

return φ

$$\ell \cdot \rho + c \leq 0 \rightarrow \rho + \lceil \frac{c}{\ell} \rceil \leq 0$$

Step II: Monotone decompositions

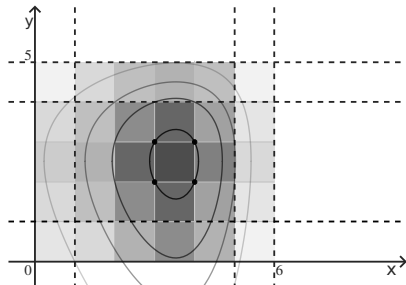
$$\begin{aligned} &\text{maximize } f(x, y) := 8x + 4y - (2^x + 2^y) \\ &\text{subject to } \varphi(x, y, z) := 0 \leq x \leq 6 \\ &\quad \quad \quad 0 \leq y \leq 5 \end{aligned}$$



Dashed lines: subspaces explored by the quantifier elimination procedure.

Step II: Monotone decompositions

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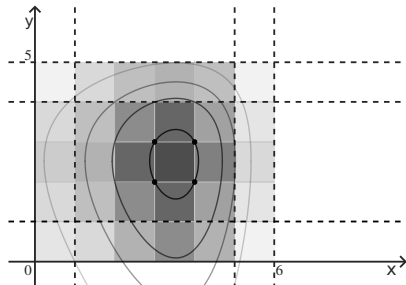
Dashed lines: subspaces explored by the quantifier elimination procedure.

Idea: when eliminating x , the objective f is monotone within $[0..3]$ and within $[4..6]$.

$(a \cdot x + \tau \leq 0) \leftarrow$ **guess** an inequality in φ , with $a \neq 0$, or an inequality in $\{x \leq 3, 4 \leq x\}$.

Step II: Monotone decompositions

$$\begin{aligned} &\text{maximize } f(x, y) := 8x + 4y - (2^x + 2^y) \\ &\text{subject to } \varphi(x, y, z) := 0 \leq x \leq 6 \\ &\quad \quad \quad 0 \leq y \leq 5 \end{aligned}$$



Dashed lines: subspaces explored by the quantifier elimination procedure.

Theorem (Hitarth, M., Shabadi. *preprint*)

One can always partition the search space into regions where the objective function is “periodically monotone”. Inequalities defining the boundaries of these regions can be added to those guessed by quantifier elimination, without affecting the NP runtime.

Recap and an open problem

Key properties of ILEP:

1. *If there are (optimal) solutions, one can be represented with a short ILESLP.*
 2. *Checking if an ILESLP is a solution*
 3. *Comparing values of the objective on two ILESLPs*
- } in $P^{\text{FACTORING}}$...

...or in P if we add a small set of primes to the certificates

Open problem: binary search on ILESLPs.

Let S be the set of all ILESLPs of size at most k . Is there an algorithm with runtime polynomial in k that, given as input $L, U \in S$, computes $M \in S$ such that the size of both sets $S_1 := \{\sigma \in S : L \leq \sigma \leq M\}$ and $S_2 := \{\sigma \in S : M \leq \sigma \leq U\}$ is in $\Omega(\#S_1 + \#S_2)$?

Above, “ $\sigma_1 \leq \sigma_2$ ” compares the value of the last expressions of σ_1 and σ_2 .