# Optimizing Low Dimensional Functions over the Integers

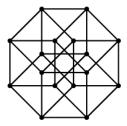
Daniel Dadush, <u>Arthur Léonard</u>, Lars Rohwedder, José Verschae

IPCO 2023

[Hunkenschröder, Pokutta, Weismantel, 2022]

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$$\begin{split} g &: \mathbb{R}^m \to \mathbb{R} \text{ is a nice convex function} \\ W &\in \mathbb{Z}^{n \times m} \text{ matrix (known or unknown),} \\ \|W\|_{\infty} &\leq \Delta, \ m \ll n. \end{split}$$



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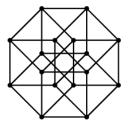
 $\rightarrow$  The algorithm runs in time  $O(n(m\Delta)^{O(m^2)})$  when W is known and g is separable convex.

#### The Problem

Our contribution : an algorithm to compute an optimal solution to :

min 
$$c^{\top}x + g(Wx)$$
  
 $l_i \le x_i \le u_i$  for all  $i \in \{1, \dots, n\}$   
 $x \in \mathbb{Z}^n$ 

 $W \in \mathbb{Z}^{m \times n}$ ,  $\|W\|_{\infty} \leq \Delta$ ,  $m \ll n$ .



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The complexity of this algorithm is :

$$O(n^{O(m)} \cdot (m\Delta)^{O(m^2)} \cdot Q)$$

#### Requirement on g

We suppose we can do oracle queries. Given :

- a partition of the variables  $I \dot{\cup} J = [n]$  with  $|I| \leq m$
- a fixing  $z \in \mathbb{Z}^J$  of the *J*-variables

it solves the following problem in time  ${\boldsymbol{Q}}$  :

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#### Remark

If g is convex and accessible by function and gradient evaluation, oracle queries can be implemented using the algorithm in [Kannan, 1983] in  $O(m^{O(m)}\langle \text{input} \rangle^{O(1)})$ 

min 
$$c^{\top}x + d^{\top}y$$
  
 $Wx + By = b$   
 $l_i \le x_i \le u_i \text{ for all } i \in \{1, \dots, n\}$   
 $x \in \mathbb{Z}^n, y \in P \subset \mathbb{R}^h$ 

The polytope P imposes constraints on the continuous variables.

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Encoding  
$$g(Wx) := \begin{cases} \min\{d^{\top}y : By = b - Wx, y \in P\} & \text{if it exists,} \\ \infty & \text{otherwise.} \end{cases}$$

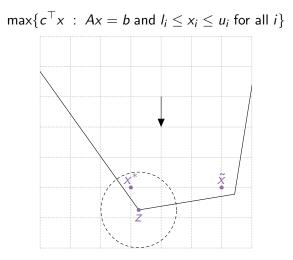
$$\max \quad \left\{ \sum_{i=1}^n p_i x_i - g\left( \sum_{i=1}^n w_i x_i \right) \ : \ x_i \in \{0, \dots, u_i\} \text{ for all } i \right\}$$

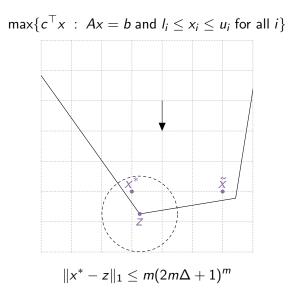
- $p_i$  : the value of item *i*
- w<sub>i</sub> : the space needed for item i
- x<sub>i</sub> : the number of item *i* that we take
- $u_i$ : the number of available item *i*

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If g is convex, our algorithm works in time  $(n + w_{max})^{O(1)}$ .





## A proximity bound

Proximity lemma [Eisenbrand, Weismantel, 2020]

Let z be an optimal **vertex** solution to

$$\max\{c^{ op}x \ : \ Ax = b \text{ and } I_i \leq x_i \leq u_i \text{ for all } i\}$$

where  $A \in \mathbb{Z}^{m \times n}$  has entries of size at most  $\Delta$ .

If there exists an optimal integer solution  $\tilde{x}$ , then there exists an optimal integer solution  $x^*$  with :

$$\|x^*-z\|_1 \leq m(2m\Delta+1)^m$$

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They give an algorithm solving the IP in time  $O(n(m\Delta)^{O(m^2)})$ 

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If  $u = \infty$ , the running time is  $O(n(m\Delta)^{O(m)})$ .

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#### Proximity lemma

 $\exists$  an optimal integer solution  $x^*$  with  $||x^* - z||_1 = O(m\Delta)^m$ 

The vector *z* has at least n - m zero components.

If T is the set of the zero components :

$$\|x_T^*\|_1 \le \|x_T^* - z_T\|_1 \le \|x^* - z\|_1 \le O(m\Delta)^m$$

Since the entries of W are bounded by  $\Delta$  :

$$\|W_T x_T^*\|_1 \le m\Delta \|x_T\|_1 \le O(m\Delta)^{m+1}$$

We guess T and  $b^{(T)} := W_T x_T^*$  among the  $n^m \times O(m\Delta)^{m(m+1)}$  choices.

The problem is now divided into :

$$\mathsf{min} \quad \left\{ c_{\mathcal{T}}^\top x_{\mathcal{T}} \ : \ \mathcal{W}_{\mathcal{T}} x_{\mathcal{T}} = b^{(\mathcal{T})} \text{ and } x_{\mathcal{T}} \in \mathbb{Z}_{\geq 0}^{|\mathcal{T}|} \right\}$$

and

$$\min \quad \left\{ c_L^\top x_L + g(W_L x^L + b^{(\tau)}) : x_L \in \mathbb{Z}_{\geq 0}^{|L|} \right\}$$

The first can be solved in  $O(n(m\Delta)^m)$  using the algorithm of Eisenbrand and Weismantel.

The second corresponds to an **oracle query**. Final running time :  $O(n^{O(m)} \cdot (m\Delta)^{O(m^2)} \cdot Q)$ 

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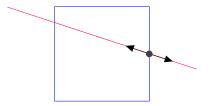
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The vector z is tight on n - m components : How to guess if  $z_i = l_i$  or  $u_i$  for these ?



We deduced the value of z on n - m components. We use the same algorithm as in the un-upper-bounded case. Final running time :

 $O(n^{O(m)} \cdot (m\Delta)^{O(m^2)} \cdot Q)$ 

When g is nice and separable convex, and W is unknown, [Hunkenschröder, Pokutta, Weismantel, 2022] give a  $O(n(m\Delta)^{O(m^3)})$ -time algorithm if we are given a value and gradient evaluation oracle for  $x \mapsto g(Wx)$ . When g is nice and separable convex, and W is unknown, [Hunkenschröder, Pokutta, Weismantel, 2022] give a  $O(n(m\Delta)^{O(m^3)})$ -time algorithm if we are given a value and gradient evaluation oracle for  $x \mapsto g(Wx)$ .

We improved it for any separable convex function in

 $O(n(m\Delta)^{O(m^2)}).$ 

 $\rightarrow$  Details in the paper !

Our contribution : an  $O(n^{O(m)} \cdot (m\Delta)^{O(m^2)} \cdot Q)$ -time algorithm to compute an optimal solution to :

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where  $W \in \mathbb{Z}^{m \times n}$  has entries bounded by  $\Delta$  in absolute value. Open problems : Can we get rid of the O(m) exponent ?  $\rightarrow$  Can be removed when c = 0 or  $u = \infty$ . Our contribution : an  $O(n^{O(m)} \cdot (m\Delta)^{O(m^2)} \cdot Q)$ -time algorithm to compute an optimal solution to :

$$\begin{array}{ll} \min & c^\top x + g(Wx) \\ & l_i \leq x_i \leq u_i \text{ for all } i \in \{1, \dots, n\} \\ & x \in \mathbb{Z}^n \end{array}$$

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Thank you for your attention !

### Questions ?

Thank you for your attention ! Make the kitten happy : ask a question !



(credit : Dall-E)

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min {
$$\|b - Wx\|_2$$
 :  $x \in \{0, 1, \dots, u_i\}^n, \|x\|_1 \le \sigma$ }

Let z be an optimal vertex solution of

$$\min\{c^{\top}x : Wx = b^*, l \le x \le u\}$$

whose dual problem is :

$$\begin{array}{ll} \max & b^{*^{\top}}y + l^{\top}s' - u^{\top}s^{u} \\ & c - W^{\top}y = s' - s^{u} \\ & s', s^{u} \in \mathbb{R}^{n}_{\geq 0} \\ & y \in \mathbb{R}^{m} \end{array}$$

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For an optimal vertex solution of the dual,  $s_i^l = s_i^u = 0$  for at least *m* components such that  $(W^{\top})_i$  are linearly independent rows.

$$\min\{c^\top x : Wx = b^*, l \le x \le u\}$$

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By guessing these rows amongst  $O(n^m)$  choices, we recover y exactly, thus  $c - W^\top y$ .

- If  $(c W^{\top}y)_i < 0$  then  $s_i^u > 0$  and  $z_i = u_i$
- If  $(c W^{\top}y)_i > 0$  then  $s'_i > 0$  and  $z_i = l_i$