

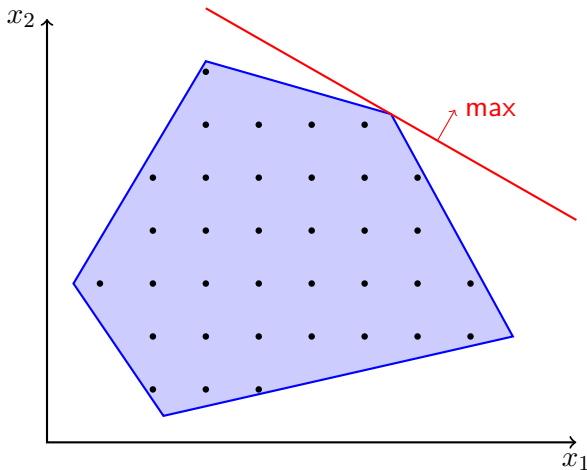
Optimization over Integers with Robustness in Cost and Few Constraints

Kai-Simon Goetzmann

(Joint work with Sebastian Stiller and Claudio Telha)

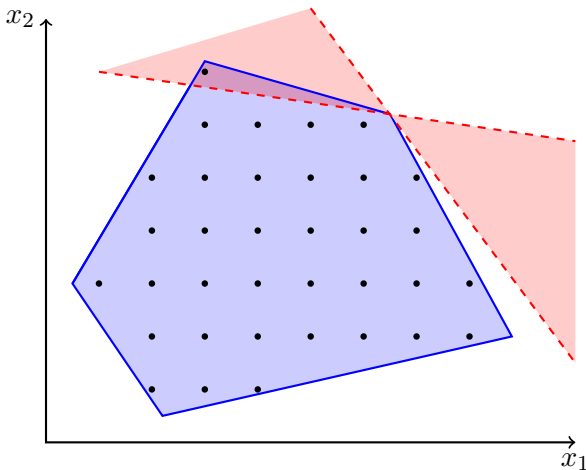
WAOA 2011 – Sept 09

Classical Optimization over Integers.



e.g. totally unimodular IPs, Unbounded Knapsack Problem, IPs with two variables per inequality, ...

Classical Optimization over Integers. But what if...?



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Cost Robust Counterpart

Given

- set of feasible solutions $X \subseteq \mathbb{Z}^n$
- cost vector $c \in \mathbb{Q}^n$

find $x \in X$ minimizing the cost

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$$c^T x + \max_{\substack{S \subseteq [n] \\ |S| \leq \Gamma}} \sum_{i \in S} |d_i x_i|$$

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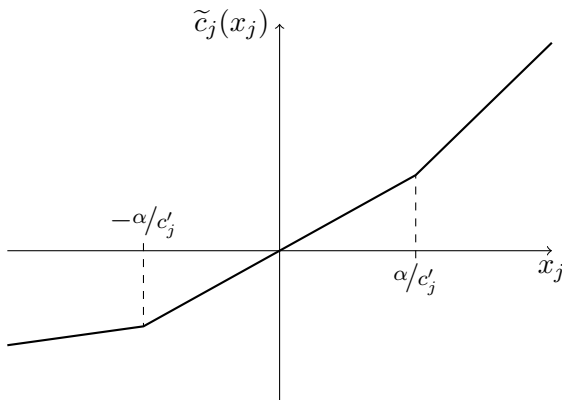
This defines the (d, Γ) -CRC of $P = (c, X)$.

- 1 Introduction
- 2 General Result for Cost Robust IPs
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To solve CRC of P : Solve the (α, c') -MMin of P .

$$\min_{x \in X} \sum_{j=1}^n \tilde{c}_j(x_j), \quad \tilde{c}_j(x_j) = c_j + \max\{c'_j x_j - \alpha, 0\} + \max\{-c'_j x_j - \alpha, 0\}$$



Theorem

Let $X \subseteq \mathbb{Z}^n, c \in \mathbb{Q}^n, d \in \mathbb{N}^n, \Gamma \in [n]$. If

- *there is a ρ -approximation algorithm for the (α, d) -MMin of P for all $\alpha \geq 0$*
- *bounds u_j on the absolute value of x_j can be computed in polynomial time,*

then there is a pseudopolynomial ρ -approximation algorithm for the (d, Γ) -CRC of $P = (c, X)$.

Proof (Sketch):

$$\min_{x \in X} \left\{ c^T x + \max_{\substack{S \subseteq [n] \\ |S| \leq \Gamma}} \sum_{j \in S} |d_j x_j| \right\}$$

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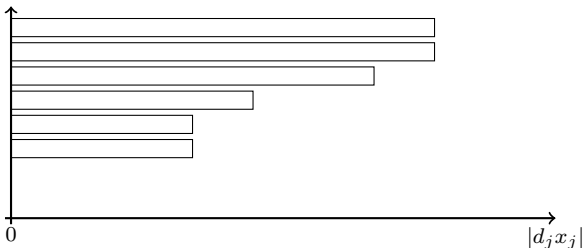
$$\min_{\substack{x \in X \\ \vartheta \geq 0}} \left\{ c^T x + \Gamma \vartheta + \sum_{j=1}^n (\max\{d_j x_j - \vartheta, 0\} + \max\{-d_j x_j - \vartheta, 0\}) \right\}$$

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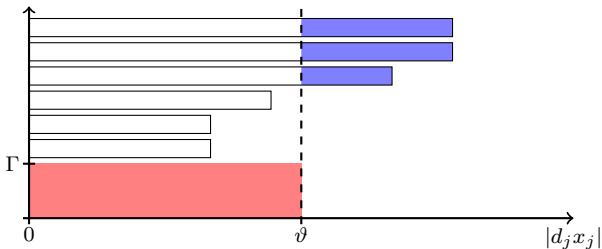


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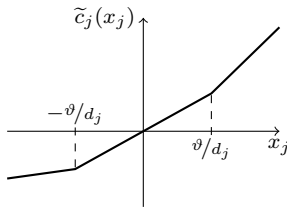
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$$= \min_{\vartheta \geq 0} \left\{ \Gamma \vartheta + \min_{x \in X} \left\{ \sum_{j=1}^n \tilde{c}_j(x_j) \right\} \right\}$$



Finding the optimal ϑ

$$\min_{\substack{x \in X \\ \vartheta \geq 0}} \left\{ c^\top x + \Gamma \vartheta + \sum_{j=1}^n (\max\{d_j x_j - \vartheta, 0\} + \max\{-d_j x_j - \vartheta, 0\}) \right\}$$



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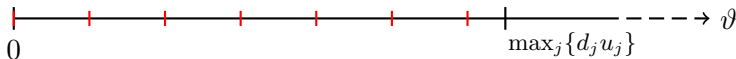
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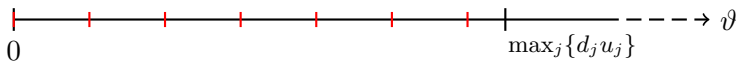
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\Rightarrow only $\max_j \{d_j u_j\} + 1$ possible values for ϑ^* ,
can enumerate in pseudopolynomial time. □

Theorem (Extension 1)

Let $X \subseteq \mathbb{Z}^n, c \in \mathbb{Q}^n, d \in \mathbb{N}^n, \Gamma \in [n]$. If

- there is a ρ -approximation algorithm for the (α, d) -MMin of P for all $\alpha \geq 0$
- bounds u_j on the absolute value of x_j can be computed in polynomial time,

then there is a pseudopolynomial ρ -approximation algorithm for the (d, Γ) -CRC of $P = (c, X)$.

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If $\rho = 1$, and if the optimal values of the (α, d) -MMin of P are convex in α , then there is an exact polynomial time algorithm for the (d, Γ) -CRC of P .

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Proof (Idea): Binary search for ϑ^* .

Theorem (Extensions 2 and 3)

Let $X \subseteq \mathbb{Z}^n, c \in \mathbb{Q}^n, d \in \mathbb{N}^n, \Gamma \in [n]$. If

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If $X \subseteq \mathbb{N}^n$ and $c \geq 0$, then for all $\varepsilon > 0$

there is a polynomial time $\rho(1 + \varepsilon)$ -approximation algorithm for the (d, Γ) -CRC of P **for minimization** problems.

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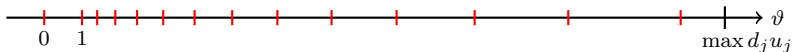
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If $X \subseteq \mathbb{N}^n$ and $c \geq 0$, and $\frac{d_j}{c_j} \leq \beta < 1$,

then there is a polynomial time 2ρ -approximation algorithm for the (d, Γ) -CRC of P for maximization problems.

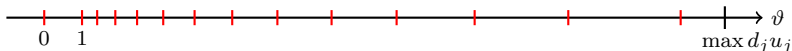
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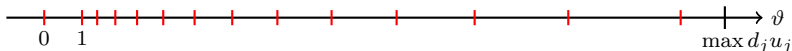
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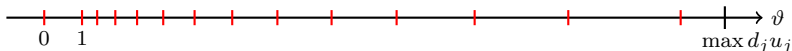


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Minimization \rightsquigarrow missing optimum by at most $\rho(1 + \varepsilon)$.

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Maximization with $\frac{d_j}{c_j} \leq \beta$, setting $\varepsilon := \frac{1-\beta}{2\beta}$

\rightsquigarrow missing optimum by at most 2ρ . □

Theorem (Extension 2)

Let $X \subseteq \mathbb{Z}^n$, $c \in \mathbb{Q}^n$, $d \in \mathbb{N}^n$, $\Gamma \in [n]$. If

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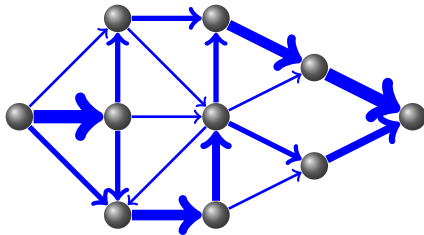
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Totally Unimodular Problems

If the minimization problem $P = (c, X)$ is given by a *bounded totally unimodular description*, i.e.

$$\text{conv}(X) = \{x \in \mathbb{R}^n : Ax \leq b, x \leq u\},$$

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then

- the (α, c') -MMin of P has a bounded TUM description,
- the optimal values of the (α, c') -MMin of P are convex in α .

\Rightarrow The (d, Γ) -CRC of P is efficiently solvable.

IPs with two variables per inequality

Bounded IP2:

$$\min_{x \in \mathbb{Z}^n} \{c^\top x : a_i^\top x \leq b_i \text{ for } i = 1, \dots, m, \quad \ell \leq x \leq u\}$$

with $b \in \mathbb{Z}^m, \ell, u \in \mathbb{Z}^n, c \in \mathbb{Q}^n,$
 $a_i \in \mathbb{Q}^n$ with only two non-zero entries.

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- Pseudopolynomial exact algorithm
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⇒ Pseudopolynomial exact algorithm for the CRC.

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- Pseudopolynomial exact algorithm for MMin of bounded *monotone* IP2s
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- Pseudopolynomial 2-approximation for MMin of bounded IP2s
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⇒ For any $\varepsilon > 0$, there is a $(2 + \varepsilon)$ -approximation algorithm for the (d, Γ) -CRC of UKP, if $\frac{d_j}{c_j} \leq \beta < 1$.

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Unbounded Knapsack Problem:

cost robust: polynomial $(2 + \varepsilon)$ -approximation,
exact in $\mathcal{O}(\max_j(d_j)n^2W^2)$
weight robust: exact in $\mathcal{O}(\max_j(\Delta w_j)n^2W^2)$
cost & weight robust: exact in $\mathcal{O}(\max_j(d_j) \max(\Delta w_j)n^2W^3)$

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Thank you for your attention.

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