

Interval Uncertainties

Robust Optimization

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Often we use different notation:

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The MIN-MAX robust optimization problem for $\mathcal{X}$ with interval uncertainties $\mathcal{U}_I$ can be solved using just one execution of the classic linear optimization problem for $\mathcal{X}$.

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Proof.

Solve an instance of the classic linear optimization problem for $\mathcal{X}$ with linear costs g given by the upper bounds $g(e) = c(e) + d(e)$. □

Interval Uncertainties and Regret

MIN-MAX REGRET

$$\min_{x \in \mathcal{X}} \max_{c \in \mathcal{U}} c(x) - \text{opt}(c)$$

where

$$\text{opt}(c) = \min_{x \in \mathcal{X}} c(x)$$

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Lemma

For the MIN-MAX REGRET problem the worst case scenario $c^{-(x)} \in \mathcal{U}_I$ is

$$c^{-(x)}(e) = \begin{cases} c^+(e) & \text{if } x_e = 1 \\ c^-(e) & \text{if } x_e = 0. \end{cases}$$

Lemma

For the MIN-MAX REGRET problem the worst case scenario $c^{-(x)} \in \mathcal{U}_I$ for any solution x is

$$c^{-(x)}(e) = \begin{cases} c^+(e) & \text{if } x_e = 1 \\ c^-(e) & \text{if } x_e = 0. \end{cases}$$

Proof.

Let $I(x) = \{e: x_e = 1\}$.

$$\begin{aligned} r(x, c) &= c(I(x)) - c(I(x^*(c))) = c(I(x) \setminus I(x^*(c)) - c(I(x^*(c)) \setminus I(x))) \\ &\leq c^+(I(x) \setminus I(x^*(c)) - c^-(I(x^*(c)) \setminus I(x))) \\ &= c^{-(x)}(x) - c^{-(x)}(x^*(c)) \\ &\leq c^{-(x)}(x) - c^{-(x)}(x^*(c^{-(x)}(x))) \end{aligned}$$



Properties of the optimal solution

Theorem

For the MIN-MAX REGRET problem for $\mathcal{X}$ the optimal solution x^ corresponds to an optimal solution for its most favorable scenario $c^{+(x^*)}$ defined as*

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We call an interval $[a, a] = \{a\}$ degenerate.

Corollary

If the number of nondegenerate intervals in $\mathcal{U}_I$ is d , the MIN-MAX REGRET robust optimization problem for $\mathcal{X}$ can be solved using 2^d calls to the classic linear optimization problem for $\mathcal{X}$.

The proof is an exercise!

Theorem

Given an instance of MIN-MAX REGRET for $\mathcal{X}$ and let x^ be its optimal solution. Let $c'(e) = \frac{1}{2}(c^-(e) + c^+(e))$ and x' the optimal solution to the linear optimization problem for $\mathcal{X}$ with respect to c' . Then for the regret of x' it holds that $r(x') \leq 2r(x^*)$.*

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Lemma

Let $X, Y \in \mathcal{F}$. Then

$$r(Y) \geq c^+(Y \setminus X) - c^-(X \setminus Y) \quad (1)$$

$$r(Y) \leq r(X) + c^+(Y \setminus X) - c^-(X \setminus Y) \quad (2)$$

Computational Complexity: Interval Uncertainty

Problems	Min-max regret
SHORTEST PATH	Strongly <i>NP</i> -hard [18]
SPANNING TREE	Strongly <i>NP</i> -hard [18]
ASSIGNMENT	Strongly <i>NP</i> -hard [1]
KNAPSACK	<i>NP</i> -hard
MIN CUT	Polynomial [2]
MIN s — t CUT	Strongly <i>NP</i> -hard [2]

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EXACT COVER BY 3-SETS

Given: $q \in \mathbb{N}$, finite set B , $|B| = 3q$, family T of 3-element subsets of B .

Question: Does T contain an exact cover for B , i.e. $T' \subseteq T$ such that every element of B occurs in exactly one member of T' ?

NP-Hardness: Spanning Tree

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AUXILIARY PROBLEM

Given: Connected graph G' .

Task: Find the maximum number of connected components that can be obtained from G' by removing the edges of some spanning tree in G' .

Thank you!

