

Robustness of Natural Extension

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- ▶ Gent → Pittsburgh → Durham
- ▶ decision making under severe uncertainty
- ▶ computational tools
- ▶ algorithms
- ▶ engineering applications

▶ Robert Hable

- ▶ Bayreuth → Munich → Bayreuth ↔ Leuven
- ▶ decision making under severe uncertainty
- ▶ nonparametric statistics
- ▶ robust statistics
- ▶ interval probability
- ▶ earlier work on robustness of natural extension ([1, pp. 118–125, Sec. 5.2] and [2, Sec. 2])

Aim

how sensitive is the natural extension of an upper prevision against small perturbations in the assessments?

- ▶ **various forms and representations** of natural extension:
does this affect stability?
- ▶ necessary and sufficient **conditions**?
- ▶ **transform instable problems** into stable ones?

Example 1: Instability of Avoiding Sure Loss

- ▶ $\Omega = \{\omega_1, \omega_2\}$, $\mathcal{L} = \mathbb{R}^\omega$, $X = \mathcal{L}^*$
- ▶ $\bar{P}(I_{\omega_2}) = 2/3$, $\bar{P}(I_{\omega_1}) = 1/3$
- ▶ **credal set** = subset of X

$$x_1 \geq 0, x_2 \geq 0, \sum_{i=1}^2 x_i = 1 \quad (\text{C})$$

$$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} 2/3 \\ 1/3 \end{bmatrix} \quad (\text{S})$$

- ▶ **natural extension** = maximize linear map x over credal set

$$\bar{E}(I_{\omega_1} + 2I_{\omega_2}) \rightarrow \text{maximize } \begin{bmatrix} 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

Example 1: Instability of Avoiding Sure Loss

- ▶ **small perturbation in assessments yields instability**
 - ▶ (C) + (S) has a non-empty feasible set
 - ▶ but (C) + $\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} 2/3 - \epsilon \\ 1/3 \end{bmatrix}$ has an empty feasible set, for any $\epsilon > 0$.
- ▶ linear programming algorithms will often fail to solve even this simple problem due to simple rounding errors!

Example 1: Stable representation

- ▶ the modified system $\begin{bmatrix} 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = [2/3]$
has the same feasible region as original problem
- ▶ perturbations $\begin{bmatrix} 0 \pm \epsilon & 1 \pm \delta \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = [2/3 \pm \eta]$
have a solution for any ϵ, δ, η sufficiently small
- ▶ solution of perturbed problem close to original solution

$$x_2 = 2/3 \sim \hat{x}_2 = \frac{2/3 + \eta - \epsilon}{1 + \delta - \epsilon}$$

- ▶ take-home message: **recognize implicit linearities**

Example 2: Instability of Natural Extension

▶ adapted from Robinson [3, p. 443]

▶ $\Omega = \{a, b, c, d\}$

$$\bar{P}(I_a + 2I_b/3 + 2I_d) = 1/2 \quad \bar{P}(I_b + 3I_c) = 3/2$$

▶ \bar{E} of $2I_b + 2I_c$:

$$\text{maximize } [0 \quad 2 \quad 2 \quad 0] \begin{bmatrix} x_a \\ x_b \\ x_c \\ x_d \end{bmatrix}$$

subject to

$$\begin{aligned} x_a \geq 0, x_b \geq 0, x_c \geq 0, x_d \geq 0 \\ x_a + x_b + x_c + x_d = 1 \end{aligned} \tag{C}$$

$$\begin{bmatrix} 1 & 2/3 & 0 & 2 \\ 0 & 1 & 3 & 0 \end{bmatrix} \begin{bmatrix} x_a \\ x_b \\ x_c \\ x_d \end{bmatrix} \leq \begin{bmatrix} 1/2 \\ 3/2 \end{bmatrix} \tag{S}$$

Example 2: Instability of Natural Extension

- ▶ (C) + (S) have a non-empty feasible set

$$\alpha \begin{bmatrix} 1/2 \\ 0 \\ 1/2 \\ 0 \end{bmatrix} + (1 - \alpha) \begin{bmatrix} 0 \\ 3/4 \\ 1/4 \\ 0 \end{bmatrix}$$

- ▶ so $\bar{E}(2I_b + 2I_c) = 2$

Example 2: Instability of Natural Extension

- ▶ (C) + (S_ϵ) , with

$$\begin{bmatrix} 1 & 2/3 - \epsilon & 0 & 2 \\ 0 & 1 & 3 & 0 \end{bmatrix} \begin{bmatrix} x_a \\ x_b \\ x_c \\ x_d \end{bmatrix} \leq \begin{bmatrix} 1/2 \\ 3/2 \end{bmatrix} \quad (S_\epsilon)$$

has only one feasible solution

$$\begin{bmatrix} 1/2 \\ 0 \\ 1/2 \\ 0 \end{bmatrix}$$

- ▶ so, now, $\bar{E}(2I_b + 2I_c) = 1$
- ▶ take-home message:

arbitrary small perturbation in assessments can have
disproportionally large effect on natural extension

Example 2: Bewildering Facts

- ▶ instability in example remains even after recognizing implicit linearities
- ▶ stable representation?
- ▶ example has perturbations that incur sure loss
- ▶ dual problem has an unbounded optimal solution
- ▶ example has arbitrarily close stable approximations:

$$(1 - \alpha)\bar{P} + \alpha \sup$$

is stable for *any* $0 < \alpha \leq 1$

all of the above are expressions of
the same mathematical result

Do you want to know why?
Come to the poster!

Thank you!!

References



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