

## Introduction

### 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?
- extend earlier work ([1, pp. 118–125, Sec. 5.2] and [2, Sec. 2])

### Example

- 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  $2I_b + 2I_c$ :

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

subject to

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

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

### Instability

- (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  $\bar{E}(2I_b + 2I_c) = 2$
- (C) +  $(S_\epsilon)$ , 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,  $\bar{E}(2I_b + 2I_c) = 1$

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

## Main Results

### Canonical Linear Program

$$\text{maximize } \sum_{\omega} x(\omega) f(\omega) \quad (1a)$$

$$\text{subject to } x(\omega) \geq 0, \sum_{\omega} x(\omega) = 1, \bar{P}(g) \geq \sum_{\omega} x(\omega) g(\omega) \quad (1b)$$

### Canonical Dual Linear Program

$$\text{mimimize } a + \sum_g \lambda_g \bar{P}(g) \quad (2a)$$

$$\text{subject to } \lambda_g \geq 0, a + \sum_g \lambda_g g \geq f \quad (2b)$$

### Stability Theorem

Let  $\bar{P}$  be any upper prevision. The following conditions are equivalent.

- The linear program (1) is stable.
- The linear program (2) is stable.
- The system of linear inequalities and equalities of (1) is regular.
- All sufficiently small perturbations of  $\bar{P}$  avoid sure loss.
- There is a linear prevision  $x$  such that  $\bar{P}(g) > x(g)$  for all  $g$  in  $\mathcal{K}$ .
- $\bar{P}$  dominates a non-linear linear-vacuous mixture.
- $\bar{P}$  avoids sure loss and  $\underline{E}(g) < \bar{E}(g)$  for all  $g$  in  $\mathcal{K}$ .

### Stable Representation Theorem

Let  $\bar{P}$  be any upper prevision. The following conditions are equivalent.

- The linear program (1) has a stable representation.
- The linear program (2) has a stable representation.
- The system of linear inequalities and equalities of (1) has a regular representation.
- There is a linear prevision  $x$  in the credal set of  $\bar{P}$  such that  $x(\omega) > 0$  for all  $\omega \in \Omega$ .
- $\bar{P}$  avoids sure loss and  $\bar{E}(I_\omega) > 0$  for all  $\omega \in \Omega$ .

### Practical Consequences

for any  $0 < \alpha \leq 1$ :

- $(1 - \alpha)\bar{P} + \alpha \sup_{\omega \in \Omega}$  is *always* stable
- $(1 - \alpha)\bar{P} + \alpha \frac{1}{|\Omega|} \sum_{\omega \in \Omega}$  *always* has a stable representation

### Open Problems

- extension to non-finite spaces
- conditional lower previsions

### References

- Robert Hable. *Data-Based Decisions under Complex Uncertainty*. PhD thesis, Ludwig-Maximilians-Universität München, 2008.
- Robert Hable. Finite approximations of data-based decision problems under imprecise probabilities. *International Journal of Approximate Reasoning*, 50(7):1115–1128, 2009.
- Stephen M. Robinson. A characterization of stability in linear programming. *Operations Research*, 25(3):435–447, 1977.