

The role of generalised p-boxes in imprecise probability theories

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ISIPTA

Introducing Didier

Position



CNRS Research advisor, IRIT, Toulouse

Some (scientific) interests

- Numerical possibility theory / imprecise probability
- Qualitative possibilistic logic
- Information fusion
- Fuzzy sets, rough sets, fuzzy interval analysis, fuzzy rules, ...
- ...

Introducing me

Position

2005 - 2008 - Phd student at the Institute of radiological protection and nuclear safety: uncertainty representation and treatment

2009 permanent researcher at French agricultural research centre working for international development

Main interests

- uncertainty modeling
- uncertainty fusion
- uncertainty propagation and independence modelling
- (soon) learning methods and graphical models

Current situation



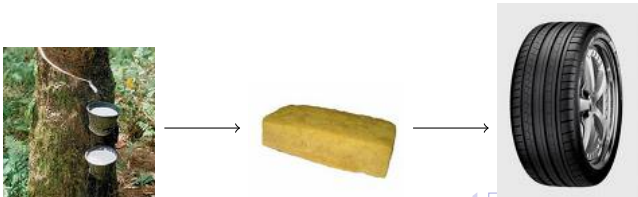
Current situation: applicative fields

→ agronomical chains, from crop to final product.

- wheat durum



- natural rubber (from Hevea tree)



Current situation: long terms goals

Build a multicriteria decision model of the whole chain, basing the modelling on:

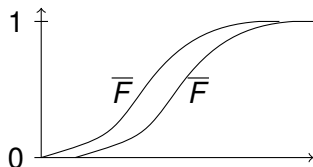
- 1 Expert knowledge (possibly modelled by IP)
- 2 Experimental data (through learning)
- 3 Physical models when available

Motivation: two useful models

P-boxes

Lower and upper prob. bounds over $(-\infty, x]$.

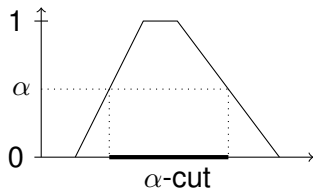
Useful for expert elicitation, risk analysis, reliability analysis, ...



Possibilities

Lower prob bounds given to nested sets

Useful for expert elicitation, linguistic assessments, ...



Generalised p-boxes: basic ideas

Take both attributes of possibility distributions and p-boxes to extend the to a single model, expressible by lower/upper probabilistic bounds over nested sets and represented by two distributions $\underline{F}, \overline{F}$

Set of constraints

$$\alpha_i \leq P(A_i) \leq \beta_i$$

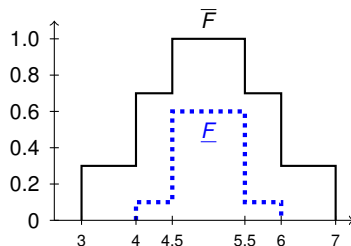
$$A_1 \subseteq A_2 \subseteq \dots \subseteq A_N \subseteq X$$

Example :

$$0.3 \leq P(\text{pH} \in [4.5, 5.5]) \leq 0.6;$$

$$0.7 \leq P(\text{pH} \in [4, 6]) \leq 0.9;$$

$$1 \leq P(\text{pH} \in [3, 7]) \leq 1.$$



Links with other models

- **Credal sets:** a gen. p-box induce a particular credal set $\mathcal{P} = \{P | \alpha_j \leq P(A_j) \leq \beta_j\}$
- **Random sets:** a gen. p-box induce an ∞ -monotone lower probability, and can therefore be seen as a random set
- **Possibility distributions:** the credal set of any gen. p-box can be described by the intersection of credal sets induced by two possibility distributions

→ to know to which extent this model is useful, need to consider its behavior and computational aspects for different treatments (conditioning, fusion, ...)

Treatments we've been looking at

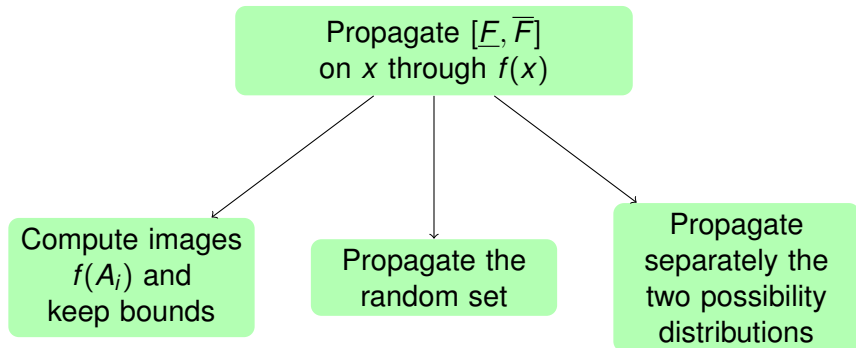
- Computing lower probabilities
- Propagation through (det.) function f
- Conditioning
- Merging

Computation of probabilistic bounds

For a given event A , how to compute $\underline{P}(A)$

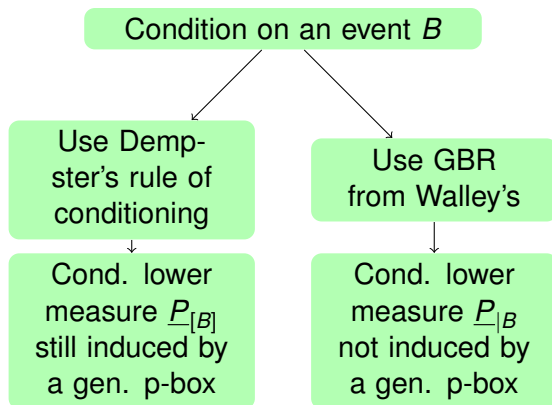
Easy due to the additivity of lower probability \underline{P} on specific collections of sets

Propagation

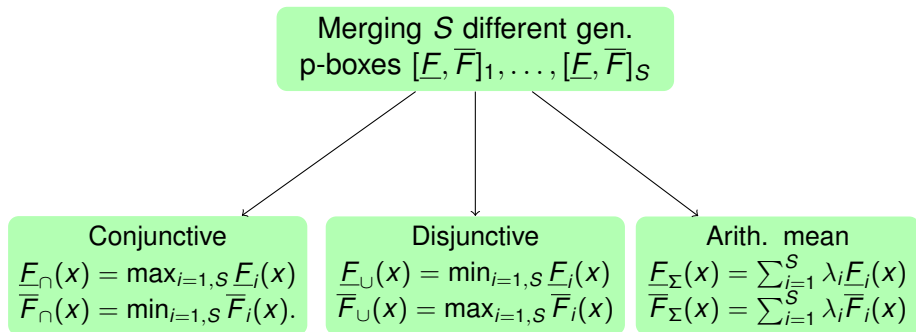


→ result usually not a gen. p-box

Conditioning



Merging (idempotent rules)



- Coherent with merging of usual p-boxes and possibility distributions
- Except under specific conditions, the result is not a gen. p-box, but a cloud (Neumaier, 04)

Conclusions

Main interests of gen. p-boxes lie in

- the elicitation and representation of information
- the use of it as (relatively gross) approximations
- the use of convex confidence regions to simplify problems (Fuchs, Neumaier, 09)

More details are on the poster!