

# Decision making under interval uncertainty

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To make a decision, we must:

- find out the user's preference, and
- help the user select an alternative which is the best – according to these preferences.

A general way to describe user preferences is via the notion of *utility* (see, e.g., [7]): we select a very bad alternative  $A_0$  and a very good alternative  $A_1$ ; utility  $u(A)$  of an alternative  $A$  is then defined as the probability  $p$  for which  $A$  is equivalent to the lottery in which we get  $A_1$  with probability  $p$ , and  $A_0$  otherwise. One can prove that utility is determined uniquely modulo linear re-scaling (corresponding to different choices of  $A_0$  and  $A_1$ ), and that the utility of a decision with probabilistic consequences is equal to the expected utility of these consequences.

Once the utility function  $u(d)$  is elicited, we select the decision  $d_{\text{opt}}$  with the largest utility  $u(d)$ . Interval techniques can help in finding the optimizing decision; see, e.g., [4].

Often, we do not know the exact probability distribution, so we need to extract, from the sample, the characteristics of a distribution which are most appropriate for decision making. We show that, under reasonable assumptions, we should select moments and cumulative distribution function (cdf). Based on a finite sample, we can only find bounds on these characteristics, so we need to deal with bounds (intervals) on moments [6] and bounds on cdf [1] (a.k.a. p-boxes).

Once we know intervals  $[\underline{u}(d), \bar{u}(d)]$  of possible values of utility, which decision shall we select? We can simply select a decision  $d_0$  which *may* be optimal,

i.e., for which  $\bar{u}(d_0) \geq \max_d \underline{u}(d)$ , but there are usually many such decisions; which of them should be select? It is reasonable to assume that this selection should not depend on linear re-scaling of utility; under this assumption, we get Hurwicz optimism-pessimism criterion  $\alpha \cdot \bar{u}(d) + (a - \alpha) \cdot \underline{u}(d) \rightarrow \max$  [7]. The next question is how to select  $\alpha$ : interestingly, e.g., too optimistic values ( $\alpha > 0.5$ ) do not lead to good decisions.

In some situations, it is difficult to elicit even interval-valued utilities. In many such situations, there are reasonable symmetries which can be used to make a decision; see, e.g., [5]. We show how this method works on the example of selecting a location for a meteorological tower [3].

Finally, while optimization problems are ubiquitous, sometimes, we need to go beyond optimization: e.g., we need to make sure that the system is *controllable* for all disturbances within a given range. In such problems, modal intervals [2] naturally appear. In more complex situations, we need to go beyond modal intervals, to more general Shary's classes.

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