Logical Approximations of Qualitative Probability

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Comparative structures provide a natural bridge between the logical and probabilistic representation of uncertainty, of relevance both for the foundations of probability and statistics [4], and AI[5]. Formally, a comparative structure is a pair (\mathcal{A}, \preceq) where \mathcal{A} is a boolean algebra and \preceq is interpreted as a qualitative probability (relation) on \mathcal{A} , i.e. we write $\theta \preceq \phi$ to say that θ is no-more-probable-than ϕ , for any $\theta, \phi \in \mathcal{A}$. The relations $\theta \approx \phi$ and $\theta \prec \phi$ are defined from \preceq as usual.

Definition 1 (Comparative structure). (A, \preceq) is a comparative structure if: 1) \preceq is a total preorder over A; 2) $\bot \prec \top$; 3) $\alpha \preceq \beta$ holds, whenever $\alpha \sqsubseteq \beta$; 4) If $\alpha \land \gamma = \bot$ and $\beta \land \gamma = \bot$, then $\alpha \preceq \beta$ if and only if $\alpha \lor \gamma \preceq \beta \lor \gamma$.

Here by \sqsubseteq we denote the lattice order of the Boolean algebra, to be distinguished by \preceq . The definition above is essentially due to De Finetti [3], who conjectured that such structures would be representable by usual probability functions. Let us recall that a comparative structure (\mathcal{A}, \preceq) is said to be: representable, if there exists a finitely additive probability P such that $\alpha \preceq \beta$ if and only if $P(\alpha) \leq P(\beta)$; almost representable in case the former claim only holds in the left-to right direction.

Contrary to De Finetti's conjecture, even almost representability fails to hold for general comparative structures, as shown in 1959 by [7]. Since then, various authors have proposed additional suitable axioms for establishing (almost) representability, e.g. [9, 8, 10, 6]. All the approaches to comparative structures still make, however, strong idealizing assumptions, e.g.:

- <u>≺</u> has to contain the classical relation <u>□</u>. But finding out whether <u>□</u> holds is non-feasible, according to standard assumptions in computational complexity.
- The axioms that need to be added to Definition 1, to obtain representability, either postulate [9] that there are arbitrarily fine-grained events to be compared, or impose conditions which are hard to interpret intuitively [8, 10]

In this work, we address these problems, by developing a sequence of comparative structures, which are meant to be *approximations* of almost representable comparative structures. Each structure in the sequence is not by itself representable, and may be rather seen as a qualitative counterpart of a (bounded) Belief function[1, 11]. We then attain the representability result only in the limit, provided that the sequence satisfies certain conditions.

Our framework is crucially based on Depth-Bounded Boolean Logics [2], instead of classical logic. These logics are centered around the idea of limiting the applications of the bivalence principle, which holds unboundedly for classical logic. In natural deduction-style the principle may be presented as follows:

$$\begin{array}{ccc}
[\varphi] & [\neg \varphi] \\
\vdots & \vdots \\
\frac{\psi}{\psi} & \psi
\end{array} (PB)$$

^{*}Speaker.

This means that to infer the formula ψ , it suffices to infer it both under the assumption that φ is the case and under the assumption that $\neg \varphi$ is the case. The square brackets around the formulas φ and $\neg \varphi$ signal that those are pieces of information assumed for the sake of deriving ψ , but not actually held true (they are *discharged*, in natural deduction terminology). We call this type of information *hypothetical*, in contrast to the *actual* information which an agent may hold as her premises.

[2] introduces a logic \vdash_0 , which does not allow any manipulation of hypothetical information, i.e. any application of PB, and may be defined proof-theoretically by a core set of introduction and elmination rules (Intelim Rules [2]) for each connective, both when occurring positively (as the main connective of a formula) and negatively (in the scope of a negation). The family of Depth-Bounded Boolean Logics $\{\vdash_k\}_{k\in\mathbb{N}}$ is then characterized, for k>0, by allowing, in addition to the rules of \vdash_0 , at most k nested applications of PB.

Results in [2] show that:

- $\vdash_0 \subset \vdash_1 \subset \cdots \subset \vdash_k \subset \cdots$, so the depth-bounded consequence relations form a hierarchy;
- $\lim_{k\to\infty} \vdash_k = \vdash$, i.e. at the limit, the hierarchy of depth-bounded boolean logics coincides with classical logic;
- for each k, \vdash_k has a polynomial decision procedure.

These properties make these logics a suitable starting point for addressing the problems of classical comparative structures discussed above. In this work, we shall: define a sequence of bounded comparative structures, based on the sequence of Depth-Bounded Boolean logics; identify the conditions under which our bounded comparative structures are asymptotically (almost) representable by a probability measure and, conversely, those conditions under which a representable qualitative probability structure can be approximated.

References

- [1] P. Baldi and H. Hosni. Depth-bounded belief functions. *International Journal of Approximate Reasoning*, 123:26–40, 2020.
- [2] M. D'Agostino, M. Finger, and D.M. Gabbay. Semantics and proof-theory of depth bounded Boolean logics. Theoretical Computer Science, 480:43-68, 2013.
- [3] B. de Finetti. Sul significato soggettivo della probabilità. Fundamenta Mathematicae, 17:289–329, 1931.
- [4] B. de Finetti. Recent suggestions for the reconciliation of theories of probability. *Proceedings of the Second Berkley Symposium on Mathematical Statistics and Probability*, 1:217–225, 1951.
- [5] J. P. Delgrande, B. Renne, and J. Sack. The logic of qualitative probability. Artificial Intelligence, 275:457–486, 2019
- [6] P.C. Fishburn. Finite Linear Qualitative Probability. Journal of Mathematical Psychology, 40(1):64-77, 1996.
- [7] C. Kraft, J. Pratt, and A. Seidenerg. Intuitive Probability On Finite Sets. The Annals of Mathematical Statistics, 30(2):408–419, 1959.
- [8] D. Kranz, R.D. Luce, P. Suppes, and A. Tversky. Foundations of measurement. Volume 1. Academic Press, New York, 1971.
- [9] L.J. Savage. The Foundations of Statistics. Dover, 2nd edition, 1972.
- [10] D. Scott. Measurement Structures and Linear Inequalities. (1956):233-247, 1964.
- [11] S.K.M. Wong, Y.Y. Yao, P. Bollmann, and H.C. Burger. Axiomatization of qualitative belief structure. *IEEE Transactions on Systems, Man, and Cybernetics*, 21(4):726–734, 1991.