

Aggregation for Belief Models

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The very general theory of “Belief Models”¹ provides a neat generalisation of (part of) AGM belief revision theory.

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This gives us a new insight into how to aggregate the opinions of probabilistic agents.

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Of particular interest are those agents whose belief set K is consistent, and closed under entailment.

AGM Expansion



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6. For all K and A , K_A^+ is the smallest belief set satisfying the above conditions

Representation



If K_A^+ satisfies the above conditions, then $K_A^+ = Cn(K \cup \{A\})$.

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So: can we define “expansion” for different representations of belief states?

Belief structures



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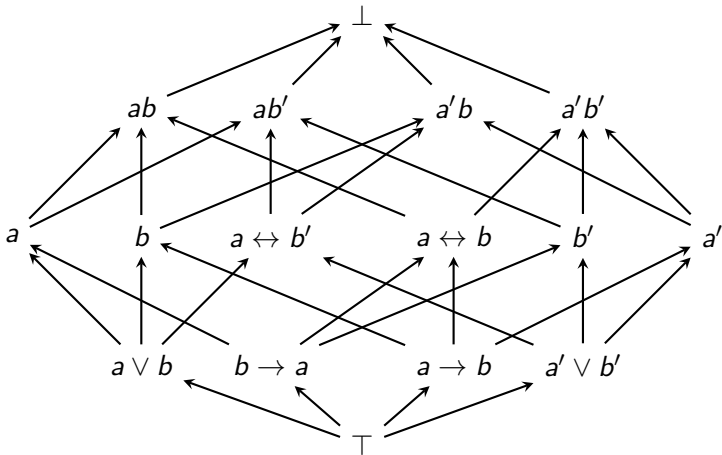
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$\langle \mathbf{S}, \mathbf{C}, \preceq \rangle$ is called a *belief structure*.

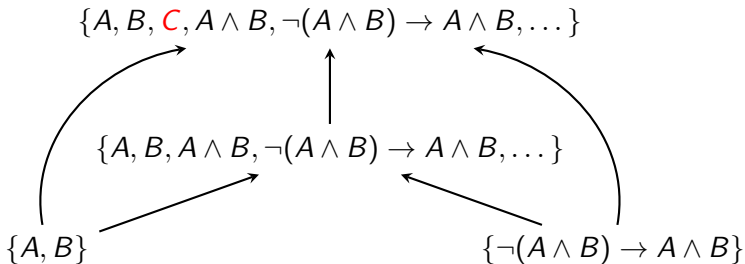
Lattice structure



Let $\bar{\mathbf{C}} = \mathbf{C} \cup \{1_{\mathbf{S}}\}$, and define:

$$Cl_{\mathbf{S}}(b) = \inf\{c \in \bar{\mathbf{C}}, b \preceq c\}$$

Closure for sets of sentences



Examples of belief structures



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- ▶ Sets of desirable gambles, choice functions. . .
- ▶ Preference relations, comparative confidence relations?

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Belief models

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5. If $b \preceq d$ then $E(b, c) \preceq E(d, c)$
6. $E(b, -)$ is the least informative of all the operators satisfying the above

AGM

If K_A^+ satisfies the above conditions, then

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Belief models

If E satisfies the above, then

$$E(b, c) = Cl_S(\sup\{b, c\}).$$

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Strong belief structures

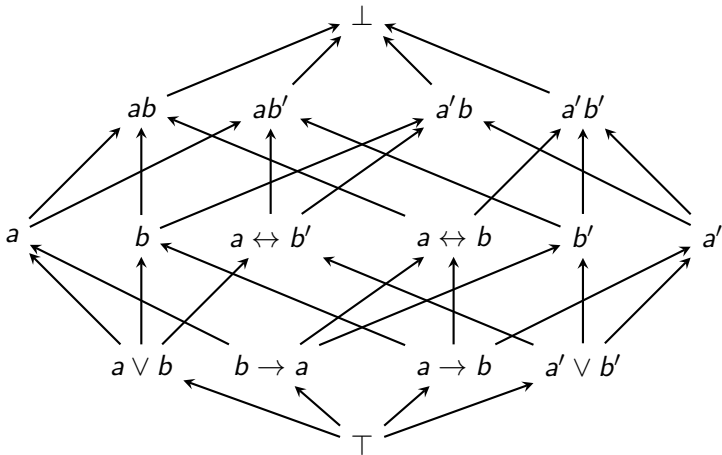


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Call a belief structure a *strong* belief structure, when, for all $c \in \mathbf{C}$, $c = \inf\{m \in \mathbf{M}, c \preceq m\}$.

Lattice structure



More on belief structures



For strong belief structures, we can do for revision what we just did for expansion!

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Interestingly, contraction seems more recalcitrant: de Cooman does not provide a “belief structure” version of contraction.

AGM generalised

Belief merge for propositions

The basic idea



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Consider a multiset Ψ of belief sets.

We want a function that maps Ψ to some belief set, subject to some constraints:

- ▶ It must satisfy some independent constraints (including consistency)
- ▶ It must be “as close” to the opinions of the members of Ψ as possible
- ▶ It must treat the different members of Ψ “fairly”

Merging operators



Call $\Delta(\Psi, \mu)$ – or $\Delta_\mu(\Psi)$ – a *merging operator* if Ψ is a multiset of belief sets, and μ is a belief set representing the constraints the aggregate belief must satisfy, and Δ satisfies:

1. $\Delta_\mu(\Psi) \vdash \mu$

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4. ...

Aside: a relation to AGM



If Δ is a merging operator, then define $K_{\mu}^* = \Delta_{\mu}(K)$. This is AGM revision.

Limits on aggregation



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- ▶ Preferences: Arrow's theorem
- ▶ Belief sets: Judgement Aggregation theorems
- ▶ Probabilities: McConway's theorem

Initial remarks on limitative theorems



- ▶ Note that typically what is required from aggregation is a maximally informative output (total order, maximal belief set, precise probability), and relaxing this condition is a natural way to avoid the limitative theorems.
- ▶ Many theorems also mention some sort of “complexity of agenda” requirement, which is hard to express in the belief models framework.

- ▶ The Belief Structures approach suggests a natural (and fairly mechanical) way of producing new results about probabilities by generalising results involving propositional logic
- ▶ The cost is that it is more natural to do things in terms of lower probabilities, rather than precise probabilities.
- ▶ (I don't actually think this is a "cost" at all, but that's another talk)

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2. If μ is consistent then $\Delta_\mu(\Psi)$ is consistent
3. If $\bigwedge \Psi$ is consistent with μ then $\Delta_\mu(\Psi) = \bigwedge \Psi \wedge \mu$
4. If $\phi \vdash \mu$ and $\phi' \vdash \mu$ then $\Delta_\mu(\phi \sqcup \phi') \wedge \phi$ is consistent iff $\Delta_\mu(\phi \sqcup \phi') \wedge \phi'$ is
5. $\Delta_\mu(\Psi_1) \wedge \Delta_\mu(\Psi_2) \vdash \Delta_\mu(\Psi_1 \sqcup \Psi_2)$
6. If $\Delta_\mu(\Psi) \wedge \Delta_\mu(\Psi_2)$ is consistent then, $\Delta_\mu(\Psi_1 \sqcup \Psi_2) \vdash \Delta_\mu(\Psi_1) \wedge \Delta_\mu(\Psi_2)$
7. $\Delta_{\mu_1}(\Psi) \wedge \mu_2 \vdash \Delta_{\mu_1 \wedge \mu_2}(\psi)$
8. If $\Delta_{\mu_1}(\Psi) \wedge \mu_2$ is consistent then $\Delta_{\mu_1 \wedge \mu_2}(\psi) \vdash \Delta_{\mu_1}(\Psi) \wedge \mu_2$

Syncretic assignments

A *syncretic assignment* is an assignment of a total preorder \trianglelefteq_{Ψ} to each multiset Ψ , such that:

1. If $a \vDash \Psi$ and $b \vDash \Psi$ then $a \trianglelefteq_{\Psi} b$
2. If $a \vDash \Psi$ but $b \not\vDash \Psi$ then $a \triangleleft_{\Psi} b$
3. If $\Psi_1 \leftrightarrow \Psi_2$, then $\trianglelefteq_{\Psi_1} = \trianglelefteq_{\Psi_2}$
4. For all $a \vDash \phi$ there is some $b \vDash \phi'$ such that $b \trianglelefteq_{\phi \sqcup \phi'} a$
5. If $a \trianglelefteq_{\Psi_1} b$ and $a \trianglelefteq_{\Psi_2} b$ then $a \trianglelefteq_{\Psi_1 \sqcup \Psi_2} b$
6. If $a \triangleleft_{\Psi_1} b$ and $a \trianglelefteq_{\Psi_2} b$ then $a \triangleleft_{\Psi_1 \sqcup \Psi_2} b$