



INSTITUTE FOR LOGIC,
LANGUAGE AND COMPUTATION



UNIVERSITY OF AMSTERDAM

**JUSTIFIED BELIEF BASED ON
UNCERTAIN AND CONTRADICTIONARY EVIDENCE**

LILAC SEMINAR
APR. 2, 2024

DAIRA PINTO PRIETO

joint work with

AYBÜKE ÖZGÜN


and


RONALD DE HAAN


“Computing (degree of) belief
based on evidence.”

THE PROBLEM

Suspects = { , , ,  }

 := { Dog, Cat, Rabbit }


 := { Dog }


 := { Rabbit, Turtle }


Images: Freepik.com, Flaticon.com



Suspects = { , , ,  }

 := { Dog, Cat, Rabbit }

 := { Dog }


 := { Rabbit, Turtle }


Do I believe in P ?


Solution: Topological Models of Evidence 🙌

evidence, argument, justification


Suspects = { , , ,  }

 := ({Dog, Cat, Rabbit}, 0.95)

 := ({Dog}, 0.70)

 := ({Rabbit, Turtle}, 0.45)

Which is my degree of belief in P ?

Solution: Demspter-Shafer Theory 

uncertainty, ignorance

INPUT

$$S = \{\text{Dog, Cat, Rabbit, Turtle}\}$$

$$E_1 = (\{\text{Dog, Cat, Rabbit}\}, 0.95)$$

$$E_2 = (\{\text{Dog}\}, 0.70)$$

$$E_3 = (\{\text{Rabbit, Turtle}\}, 0.45)$$

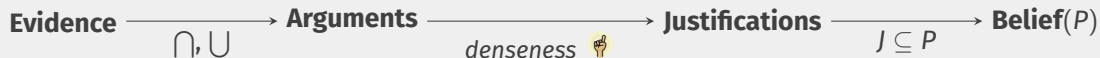
OUTPUT

Degree_of_Belief(P)

where P is a subset of S

How to combine *imperfect* evidence and compute degree of belief based on it?
(Capturing both Topological Models of Evidence and Dempster-Shafer Theory)

OUR SOLUTION: MULTI LAYER BELIEF MODEL



{Dog, Cat, Rabbit}

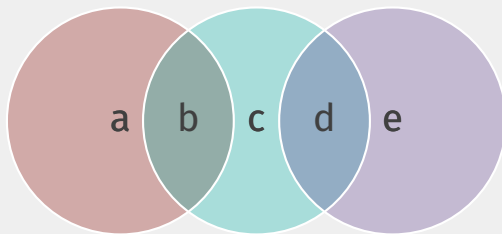
{Dog}

{Rabbit}

{Dog, Rabbit}

{Rabbit, Turtle}

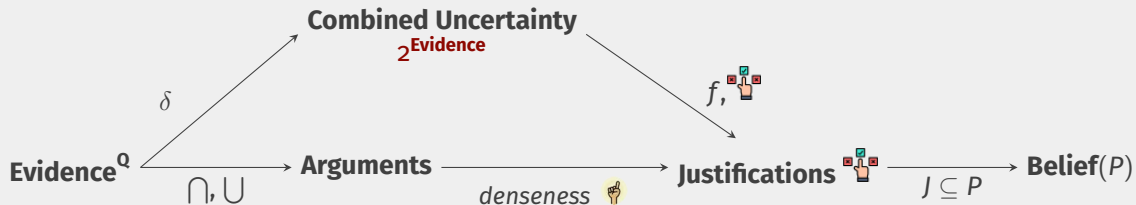
$$S = \{a, b, c, d, e\}$$



$$B(\{b, d, e\}) = 1$$

$$B(\{d, e\}) = 0$$

OUR SOLUTION: MULTI LAYER BELIEF MODEL



({ Dog, Cat, Rabbit }, 0.95)

({ Dog }, 0.70)


({ Rabbit, Turtle }, 0.45)

({ Rabbit }, ?)

({ Dog, Rabbit }, ?)

Input: Universe S , Evidence + Certainty Degree $\mathcal{E} = \{(E_i, p_i)\}_{i=1, \dots, k}$, 2 Parameters

Output: Degree of belief for every $P \in 2^S$

Justification Frame  (Qualitative Layer)

$\mathcal{J}_1 :=$ Every element of the topology of evidence (except \emptyset).

e.g., $\mathcal{J}_1 =$

$\{\{\text{Dog, Cat, Rabbit}\}, \{\text{Dog}\}, \{\text{Rabbit, Turtle}\}, \{\text{Rabbit}\}, \{\text{Rabbit, Dog}\}, \{\text{Dog, Rabbit, Turtle}\}, \{\text{Dog, Cat, Rabbit, Turtle}\}\}$

$\mathcal{J}_2 :=$ Dense elements of the topology of evidence.

e.g.,

$\mathcal{J}_2 = \{\{\text{Dog, Cat, Rabbit}\}, \{\text{Dog, Rabbit}\}, \{\text{Dog, Rabbit, Turtle}\}, \{\text{Dog, Cat, Rabbit, Turtle}\}\}$

Input: Universe S , Evidence + Certainty Degree $\mathcal{E} = \{(E_i, p_i)\}_{i=1, \dots, k}$, 2 Parameters

Output: Degree of belief for every $P \in 2^S$

δ -function (Quantitative Layer)

$$\delta : 2^{\mathcal{E}} \rightarrow [0, 1]$$

$$\delta(\mathbf{E}) = \prod_{\substack{(E_i, p_i) \text{ s.t.} \\ E_i \in \mathbf{E}}} p_i \prod_{\substack{(E_i, p_i) \text{ s.t.} \\ E_i \notin \mathbf{E}}} (1 - p_i)$$


e.g., $\delta(\{\{\text{Dog, Cat, Rabbit}\}, \{\text{Rabbit, Turtle}\}\}) = 0.95 \cdot 0.45 \cdot (1 - 0.70) = 0.13$

✓ The Certainty values are combined and distributed over $2^{\mathcal{E}}$.


✗ Our justifications are in 2^S , not in $2^{\mathcal{E}}$.

Input: Universe S , Evidence + Certainty Degree $\mathcal{E} = \{(E_i, p_i)\}_{i=1, \dots, k}$, 2 Parameters

Output: Degree of belief for every $P \in 2^S$

Allocation Function  (Bridging Layer)

$f : 2^{\mathcal{E}} \rightarrow$ Topology of Evidence


Intersection, Union 

e.g., if $f =$ intersection

then $f(\{\{\text{Dog, Cat, Rabbit}\}, \{\text{Rabbit, Turtle}\}\}) = \{\text{Rabbit}\}$

and the δ -value of $\{\{\text{Dog, Cat, Rabbit}\}, \{\text{Rabbit, Turtle}\}\}$ goes for $\{\text{Rabbit}\}$.

✓ The certainty values are combined and distributed over 2^S .

✓ After some operations , the certainty values are combined and distributed over \mathcal{I} .

OUR SOLUTION: MULTI LAYER BELIEF MODEL

Input: Universe S , Evidence + Certainty Degree $\mathcal{E} = \{(E_i, p_i)\}_{i=1, \dots, k}$, 2 Parameters

Output: Degree of belief for every $P \in 2^S$

e.g., $\text{BELIEF}(\{\text{Dog, Rabbit}\}) = 0.81$; $\text{BELIEF}(\{\text{Dog}\}) = 0.40$ under certain parameters.

e.g., $\text{BELIEF}(\{\text{Dog, Rabbit}\}) = 0$; $\text{BELIEF}(\{\text{Dog}\}) = 0$ under certain parameters.

Layers

Justification Frame 

δ -function

Allocation Function 

Belief Function

$\text{BELIEF}(P) \in [0, 1]$

adding δ -values of justifications contained in P

- ✓ Partial evidence
- ✓ Uncertain evidence
- ✓ Inconsistent evidence
- ✓ It replicates Topological Models of Evidence and Demspter's rule of combination
- ✓ It does not increase the computational complexity of the problem (#P-COMPLETE)

- Computational Complexity analysis of Demspter's rule of combination, Topological Models of Evidence and Multi-Layer Belief Model.
- Knowledge Compilation approach for Demspter's rule of combination.
- Logic for reasoning about comparative strengths of evidence and belief under uncertainty.

Starting point:

Harmanec, David; Hájek, Petr. 2011. **A qualitative belief logic.** *International Journal of Uncertainty, Fuzziness and Knowledge-Based Systems.*

$$\mathcal{M} = \langle S, \text{bel}, \mathcal{V} \rangle$$

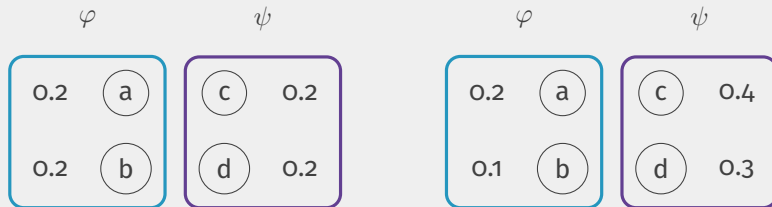
$$\text{Bel}(\varphi) = \text{bel}(\llbracket \varphi \rrbracket)$$

$$\mathcal{M}, s \models \varphi \triangleleft \psi \text{ if and only if } \text{Bel}(\varphi) \leq \text{Bel}(\psi)$$

$$\mathcal{M} = \langle S, \mathcal{E}^Q, \text{Bel}, \mathcal{V} \rangle$$

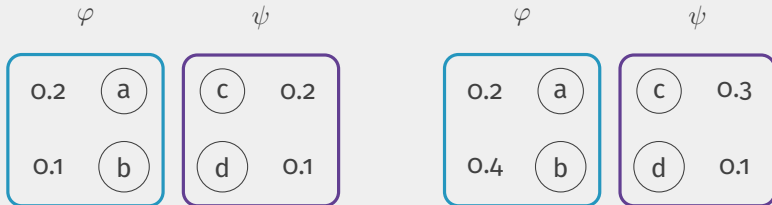
$$\mathcal{M}, s \models \varphi \triangleleft_{\text{Bel}} \psi \text{ if and only if } \text{Bel}(\varphi) \leq \text{Bel}(\psi)$$

$$\mathcal{M}, s \models \varphi \triangleleft_{\bullet} \psi \text{ if and only if ?}$$



$\varphi \triangleleft \bullet \psi$ if and only if for every $X, Y \in \mathcal{E}$ such that $X \subseteq \llbracket \varphi \rrbracket$ and $Y \subseteq \llbracket \psi \rrbracket$, $q_X \leq q_Y$.

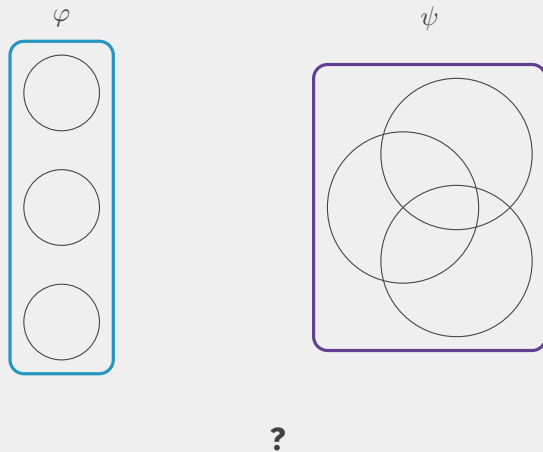
DEFINING A ELEMENT-WISE COMPARISON OPERATOR




$\varphi \triangleleft \bullet \psi$ if and only if



for every $X \in \mathcal{E}$ such that $X \subseteq \llbracket \varphi \rrbracket$ there exists $Y \in \mathcal{E}$ such that $Y \subseteq \llbracket \psi \rrbracket$ and $q_X \leq q_Y$,
and for every $Y \in \mathcal{E}$ such that $Y \subseteq \llbracket \psi \rrbracket$ there exists $X \in \mathcal{E}$ such that $X \subseteq \llbracket \varphi \rrbracket$ and $q_X \leq q_Y$.

DEFINING A ELEMENT-WISE COMPARISON OPERATOR



- Shafer, G. 1976. *A Mathematical Theory of Evidence*. Princeton, NJ: Princeton University Press.
- Baltag, A.; Bezhanishvili, N.; Özgün, A.; and Smets, S. 2022. Justified belief, knowledge, and the topology of evidence. *Synthese* 200(6):1–51.
- Pinto Prieto, D.; de Haan, R.; and Özgün, A. 2023. A belief model for conflicting and uncertain evidence – connecting dempster-shafer theory and the topology of evidence. In *Proceedings of the 20th International Conference on Principles of Knowledge Representation and Reasoning (KR 2023)*, 552–561.
- Harmanec, David; Hájek, Petr. 2011. A qualitative belief logic. *International Journal of Uncertainty, Fuzziness and Knowledge-Based Systems*.

[multi-layer-belief-model](#) / [multilayermodel.ipynb](#) 



 **dpintoprieto** Add visual representations and explanation of the model 0f57d9f · last month  **History**

Preview

Code

Blame

2742 lines (2742 loc) · 319 KB

Raw  

Compute degrees of belief with Multi-Layer Belief Model in python

Welcome to this Jupyter Notebook, where we present an implementation of the Multi-Layer Belief model as defined in the paper titled "A Belief Model for Conflicting and Uncertain Evidence: Connecting Dempster-Shafer Theory and the Topology of Evidence" by Dajira Pinto Prieto, Ronald de Haan, and Aybuke Özgön. You can find it at [this link](#).

The primary goal of this notebook is to show data researchers an accessible way to explore the potential benefits of this theoretical model for their experiments, using a simple toy example. Please note that this research is a work in progress, and we have not yet developed an optimized implementation.

Consider using this model if:

1. Your data are independent pieces of evidence that can be model by sets.
2. Each piece of evidence has a degree of certainty between 0 and 1.
3. You want to combine these pieces of evidence to derive a degree of belief.

Input: Universe S , Evidence + Certainty Degree $\mathcal{E} = \{(E_i, p_i)\}_{i=1, \dots, k}$

Output: Degree of belief for every $P \in 2^S$

e.g., $\text{BEL}(\{\text{Dog}, \text{Rabbit}\}) = 0.75$; $\text{BEL}(\{\text{Dog}\}) = 0.56$

Dempster's Rule of Combination

$$E_1 \oplus E_2$$



Belief Functions

$$\text{BEL}(P) \in [0, 1]$$



✓ Partial evidence ✓ Uncertain evidence ~ Inconsistent evidence

Computational Complexity: #P-COMplete

Definition (Basic probability assignment)

$m : 2^S \rightarrow [0, 1]$ such that $m(\emptyset) = 0$ and $\sum_{A \subseteq S} m(A) = 1$.

Definition (Dempster's rule of combination)

Let m_1 and m_2 be b.p.a. Then $m_1 \oplus m_2$, is:

$m(\emptyset) = 0$ and

$m(C) = \frac{\sum_{A_i \cap B_j = C} m_1(A_i) m_2(B_j)}{K}$, where K is $1 - \sum_{A_i \cap B_j = \emptyset} m_1(A_i) m_2(B_j)$, for all nonempty sets $C \subseteq S$.

Definition (Belief function)

$BEL : 2^S \rightarrow [0, 1]$ such that

$BEL(\emptyset) = 0$, $BEL(S) = 1$ and $BEL(\cup_1^n A_i) \geq \sum_{\emptyset \neq I \subseteq \{1, \dots, n\}} (-1)^{|I|+1} BEL(\cap_{i \in I} A_i)$.

Input: Universe S , Evidence $\mathcal{E} = \{E_i\}_{i=1,\dots,k}$ (no uncertainty!)

Output: Boolean belief operator for every $P \in 2^S$

e.g., $B(\{\text{Dog}, \text{Rabbit}\}) = 1$; $B(\{\text{Dog}\}) = 0$

Topology of Evidence

Evidence set
closed under finite intersections
and arbitrary unions



Belief Operator

$B(P) = 1$ iff
there is a *dense* set in the topology
contained in P



✓ Partial evidence ✗ Uncertain evidence ✓ Inconsistent evidence

Computational Complexity: EXP (applying formal definition straightforward)

Definition (Topological Space)

A *topological space* is a pair (S, τ) , where S is a nonempty set and τ is a family of subsets of S such that $S, \emptyset \in \tau$, and τ is closed under finite intersections and arbitrary unions.

Definition (Topology of Evidence)

Given a set of pieces of evidence \mathcal{E} , the topology $\tau_{\mathcal{E}}$ generated by \mathcal{E} is called the *topology of evidence*.

Definition (Mutually inconsistent/consistent)

We say that two pieces of evidence $E, E' \in \tau_{\mathcal{E}}$ are *mutually inconsistent* if $E \cap E' = \emptyset$, and mutually consistent otherwise.

Definition (Belief Operator)

A proposition P is *believed* if and only if there is a $T \in \tau_{\mathcal{E}} \setminus \{\emptyset\}$ and $T \subseteq P$, such that T is consistent with any piece of evidence in $\tau_{\mathcal{E}} \setminus \{\emptyset\}$, i.e., T is dense in S w.r.t. $\tau_{\mathcal{E}}$.

Definition (Set of evidence allocation functions)

Let (S, \mathcal{E}) be a qualitative evidence frame. A set of evidence allocation functions \mathfrak{F} on (S, \mathcal{E}) is a set of functions from $2^{\mathcal{E}}$ to $\tau_{\mathcal{E}}$ (the topology generated by \mathcal{E}) such that for all $f, g \in \mathfrak{F}$:

1. $f(\emptyset) = S$,
2. for all $\mathbf{E} \subseteq \mathcal{E}$, $f(\mathbf{E}) \in \tau_{\mathbf{E}}$ (the topology generated by \mathbf{E}) and it is dense in $\cup \mathbf{E}$ w.r.t. $\tau_{\mathbf{E}}$; or $f(\mathbf{E}) = \emptyset$.^a
3. for all $\mathbf{E} \subseteq \mathcal{E}$ and every f, g in \mathfrak{F} , $f(\mathbf{E}) \subseteq g(\mathbf{E})$ or $g(\mathbf{E}) \subseteq f(\mathbf{E})$.

^aNote that since $\mathbf{E} \subseteq \mathcal{E}$, we have $\tau_{\mathbf{E}} \subseteq \tau_{\mathcal{E}}$ (follows immediately by the definitions of generated topologies).

Definition (Minimal dense set allocation function)

$d : 2^{\mathcal{E}} \rightarrow \tau_{\mathcal{E}}$ such that

$$d(\mathbf{E}) = \begin{cases} \min((\text{dense}(\mathbf{E}), \subseteq)) & \text{if } \mathbf{E} \neq \emptyset, \\ S & \text{otherwise.} \end{cases}$$

MULTI-LAYER BELIEF MODEL: FORMULAS

Input:

$$S \quad \mathcal{E} \times [0, 1]$$

**Justification
frame:**

$$\mathcal{J}$$

**Allocation
Function:**

$$f : 2^{\mathcal{E}} \rightarrow \tau$$

**Mass
Function:**

$$\delta : 2^{\mathcal{E}} \rightarrow [0, 1]$$

$$\delta_{\tau}(T) = \sum_{\mathbf{E}: f(\mathbf{E})=T} \delta(\mathbf{E})$$

$$\delta_{\mathcal{J}}(A) = \frac{\delta_{\tau}(A)}{\sum_{T \in \mathcal{J}} \delta_{\tau}(T)}$$

$$Bel^{\mathcal{J}}(P) = \sum_{J \subseteq P} \delta_{\mathcal{J}}(J)$$

MULTI-LAYER BELIEF MODEL: FULL EXAMPLE

$\{\text{Dog, Rabbit}\}$	\mathcal{J}_1	\mathcal{J}_2
Intersection	0.75	0
Union	0.02	0
Minimal dense set	0.81	0.62

$\{\text{Dog}\}$	\mathcal{J}_1	\mathcal{J}_2
Intersection	0.56	0
Union	0.02	0
Minimal dense set	0.40	0

$$S = \{A, B, C\}$$

$$m_1(A) = 0.99, m_1(B) = 0.01,$$

$$m_2(C) = 0.99, m_2(B) = 0.01$$

$$\mathbf{m_1 \oplus m_2(B) = 1}$$

$$\mathbf{m_1 \oplus m_2(\{A, C\}) = 0}$$