

Min-based fusion of possibilistic networks

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Motivations

- **Important problem** : databases, expert opinions pooling, preference agregations, etc.
- **Why merging?**
 - exploit complementarities between the sources,
 - get a global and coherent point of view,
 - reduce imprecision,
 - deals with redunduncies, etc.

Motivations

- Different formats for representing uncertain information:
 - Possibilistic logic knowledge bases
 - Possibilistic networks
- General classes of merging operators:
 - Conjunctive merging
 - Disjunctive merging
 - Adaptative merging
- Syntactic fusion defined on possibilistic knowledge bases

Aim

- **Input :**
 - n consistent possibilistic networks (provided by n experts)
 - Conjunctive operator (minimum)
- **Goal** Compute a new possibilistic network representing the result of merging

Outline

- 1 **Definitions of Min-based possibilistic networks**
- 2 **Fusion of same-structure networks**
- 3 **Fusion of networks with different structures**
 - Case 1 : Union of graphs is acyclic
 - Case 2 : Union of graphs contains cycles

Outline

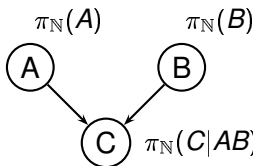
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Possibility Distributions

- $\pi : \Omega \rightarrow [0, 1], \max_{\omega \in \Omega} \pi(\omega) = 1.$

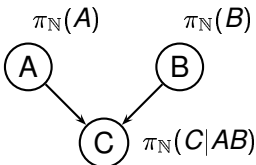
- Two dual measures :
 - Possibility measure of ϕ :
 $\Pi(\phi) = \max\{\pi(\omega) : \omega \models \phi\}$
 \Rightarrow The compatibility degree of ω with available knowledge.
 - Necessity measure of ϕ :
 $N(\phi) = 1 - \Pi(\neg\phi)$
 \Rightarrow The certainty degree associated with ϕ from available pieces of information encoded by π .

Possibilistic Networks $\mathbb{N} = (\pi_{\mathbb{N}}, G_{\mathbb{N}})$



- $G_{\mathbb{N}}$: Directed Acyclic Graph (DAG)
 - Node \leftarrow variable
 - Arc \leftarrow causal relationship

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- $G_{\mathbb{N}}$: Directed Acyclic Graph (DAG)
 - Node \leftarrow variable
 - Arc \leftarrow causal relationship
- $\pi_{\mathbb{N}}$: Conditional possibility distributions in the context of each parent

$$\pi_{\mathbb{N}}(ABC) = \min(\pi_{\mathbb{N}}(A), \pi_{\mathbb{N}}(B), \pi_{\mathbb{N}}(C|AB))$$

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Fusion of same-structure networks

Input : Two possibilistic networks having the same graph:

$\mathbb{N}1 = (\pi_{\mathbb{N}1}, G_{\mathbb{N}1})$ and $\mathbb{N}2 = (\pi_{\mathbb{N}2}, G_{\mathbb{N}2})$ s.t.

$$G_{\mathbb{N}1} = G_{\mathbb{N}2}$$

- **Result of merging is immediate**

$\mathbb{N}\oplus = (\pi_{\mathbb{N}\oplus}, G_{\mathbb{N}\oplus})$ such that:

- $G_{\mathbb{N}\oplus} = G_{\mathbb{N}1} = G_{\mathbb{N}2}$

- $\forall A$:

$$\pi_{\mathbb{N}\oplus}(A | U_A) = \min(\pi_{\mathbb{N}1}(A | U_A), \pi_{\mathbb{N}2}(A | U_A)).$$

Fusion of same-structure networks

Input : Two possibilistic networks having the same graph:

$N1 = (\pi_{N1}, G_{N1})$ and $N2 = (\pi_{N2}, G_{N2})$ s.t.

$$G_{N1} = G_{N2}$$

- **Result of merging is immediate**

$N\oplus = (\pi_{N\oplus}, G_{N\oplus})$ such that:

- $G_{N\oplus} = G_{N1} = G_{N2}$

- $\forall A$:

$$\pi_{N\oplus}(A | U_A) = \min(\pi_{N1}(A | U_A), \pi_{N2}(A | U_A)).$$

- We have :

$$\forall \omega \in \Omega, \pi_{N\oplus}(\omega) = \min(\pi_{N1}(\omega), \pi_{N2}(\omega))$$

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Case 1 : Union of graphs is acyclic

Input $\mathbb{N}1 = (\pi_{\mathbb{N}1}, G_{\mathbb{N}1})$ and $\mathbb{N}2 = (\pi_{\mathbb{N}2}, G_{\mathbb{N}2})$ s.t. :

$$G_{\mathbb{N}1} \neq G_{\mathbb{N}2}$$

Result of merging in three steps:

- **step 1** : Define $G_{\mathbb{N}\oplus}$ the union of $G_{\mathbb{N}1}$ and $G_{\mathbb{N}2}$
- **step 2** : Expand equivalently $\mathbb{N}1$ and $\mathbb{N}2$ into :
 $\mathbb{N}1' = (\pi_{\mathbb{N}1'}, G_{\mathbb{N}\oplus})$ and $\mathbb{N}2' = (\pi_{\mathbb{N}2'}, G_{\mathbb{N}\oplus})$
- **step 3** : Apply Fusion of same structure networks to $\mathbb{N}1'$ and $\mathbb{N}2'$.

Case 1 : Union of graphs is acyclic

Input $\mathbb{N}1 = (\pi_{\mathbb{N}1}, G_{\mathbb{N}1})$ and $\mathbb{N}2 = (\pi_{\mathbb{N}2}, G_{\mathbb{N}2})$ s.t. :

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How to equivalently expand possibilistic networks?

Expanding networks : Adding new variables

● Input

- A possibilistic network $\mathbb{N} = (\pi_{\mathbb{N}}, G_{\mathbb{N}})$ built on a set of variables V
- A new variable A

● Output : $\mathbb{N}_1 = (\pi_{\mathbb{N}_1}, G_{\mathbb{N}_1})$ such that :

- $G_{\mathbb{N}_1}$ is equal to $G_{\mathbb{N}}$ plus a root node A , and
- $\pi_{\mathbb{N}_1} = \pi_{\mathbb{N}}$ for variables in V , and is uniform on A (i.e., $\forall a \in D_A, \pi_{\mathbb{N}_1}(a)=1$).

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Then, we have : $\forall \omega \in \times_{A_i \in V} D_{A_i}$,

$$\pi_{\mathbb{N}}(\omega) = \prod_{\mathbb{N}_1}(\omega)$$

Expanding networks : Adding new links

● Input

- A possibilistic network $\mathbb{N} = (\pi_{\mathbb{N}}, G_{\mathbb{N}})$
- B a variable which is not a parent of A

● Output : $\mathbb{N}_1 = (\pi_{\mathbb{N}_1}, G_{\mathbb{N}_1})$ such that :

- $G_{\mathbb{N}_1}$ is obtained from $G_{\mathbb{N}}$ by adding a link from B to A ,
- The new conditionnal possibility distribution of A is:

$$\forall a \in D_a, b \in D_B, u \in D_{Par(A)},$$

$$\pi_{\mathbb{N}_1}(a \mid ub) = \pi_{\mathbb{N}}(a \mid u).$$

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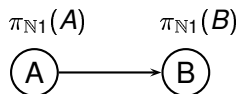
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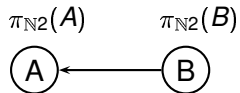
$$\forall \omega, \pi_{\mathbb{N}}(\omega) = \pi_{\mathbb{N}_1}(\omega)$$

Case 2 : Union of graphs contains cycles

- G_1

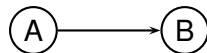


- G_2

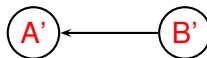


Step 1 rename variables of G_2

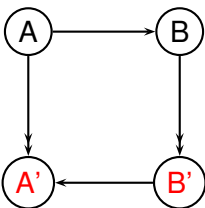
- G_1



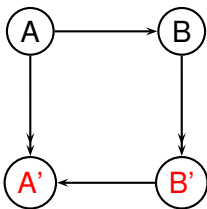
- G_2



Step : Relating old and new variables



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This defines $G_{N\oplus}$.

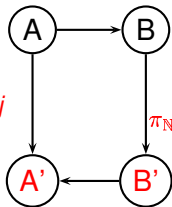
Step 3 : Defining conditional possibility distributions

$$\pi_{\mathbb{N}\oplus}(A) = \pi_{\mathbb{N}1}(A)$$

$$\pi_{\mathbb{N}\oplus}(B | A) = \pi_{\mathbb{N}1}(B | A)$$

$$\pi_{\mathbb{N}\oplus}(a'_i | a_j b'_k) = \begin{cases} \pi_{\mathbb{N}2}(a_i | b'_k) & \text{if } i = j \\ 0 & \text{other.} \end{cases}$$

$$\pi_{\mathbb{N}\oplus}(b'_i | b_j) = \begin{cases} \pi_{\mathbb{N}2}(b_i) & \text{if } i = j \\ 0 & \text{other.} \end{cases}$$



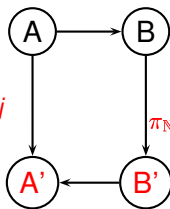
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We have :

$$\min(\pi_{\mathbb{N}1}(\omega), \pi_{\mathbb{N}2}(\omega)) = \Pi_{\mathbb{N}\oplus}(\omega).$$

Conclusion

- Fusion of possibilistic networks
- Efficient fusion procedures for networks :
 - having same graphs
 - the union of their graphs is free of cycles
- Investigate other combination operators