

A Generalization of Credal Networks

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hierarchical model:

- description with **two levels**:

probabilistic level: set \mathcal{P} of probability measures P on (Ω, \mathcal{A})

possibilistic level: likelihood function $lik : \mathcal{P} \rightarrow (0, \infty)$

- description as **set of measures**:

set \mathcal{M} of measures μ on (Ω, \mathcal{A}) with $\mu(\Omega) \in (0, \infty)$

connection between the two descriptions:

- two levels \rightsquigarrow set of measures:

$$\mathcal{M} = \{lik(P) P : P \in \mathcal{P}\}$$

- set of measures \rightsquigarrow two levels:

probabilistic level: $\mathcal{P} = \left\{ \frac{\mu}{\mu(\Omega)} : \mu \in \mathcal{M} \right\}$

possibilistic level: $lik : \mathcal{P} \rightarrow (0, \infty)$ with

$$lik(P) \propto \sup_{\mu \in \mathcal{M} : \frac{\mu}{\mu(\Omega)} = P} \mu(\Omega)$$

updating of hierarchical model (event $A \in \mathcal{A}$ observed):

- description with two levels:

probabilistic level: $\mathcal{P} \rightsquigarrow \mathcal{P}' = \{P(\cdot | A) : P \in \mathcal{P}, P(A) > 0\}$

possibilistic level: $lik \rightsquigarrow lik' : \mathcal{P}' \rightarrow (0, \infty)$ with

$$lik'(P') \propto \sup_{P \in \mathcal{P} : P(\cdot | A) = P'} lik(P) P(A)$$

- description as set of measures:

$\mathcal{M} \rightsquigarrow \mathcal{M}' = \{\mu(\cdot \cap A) : \mu \in \mathcal{M}, \mu(A) > 0\}$

in particular (**Theorem 2**): if $\mathcal{M} = \text{ch}(\{\mu_i : i \in \{1, \dots, m\}\})$,
then $\mathcal{M}' = \text{ch}(\{\mu_i(\cdot \cap A) : i \in \{1, \dots, m\}, \mu_i(A) > 0\})$

hierarchical network:

directed acyclic graph with nodes $X_1 \in \mathcal{X}_1, \dots, X_k \in \mathcal{X}_k$, such that to each node X_i are associated a set \mathcal{P}_i of stochastic kernels P_i from $\mathcal{P}\mathcal{A}_i$ (the set of all possible values of the parents of X_i) to \mathcal{X}_i , and a likelihood function $lik_i : \mathcal{P}_i \rightarrow (0, \infty)$

- description with two levels:

prob. level: $\mathcal{P} = \{P_{P_1, \dots, P_k} : P_1 \in \mathcal{P}_1, \dots, P_k \in \mathcal{P}_k\}$ with

$$P_{P_1, \dots, P_k}(x_1, \dots, x_k) = \prod_{i=1}^k P_i(x_i | pa_i(x_1, \dots, x_k))$$

poss. level: $lik : \mathcal{P} \rightarrow (0, \infty)$ with

$$lik(P) \propto \sup_{\substack{P_1 \in \mathcal{P}_1, \dots, P_k \in \mathcal{P}_k : \\ P_{P_1, \dots, P_k} = P}} \prod_{i=1}^k lik_i(P_i)$$

- description as set of measures:

$\mathcal{M} = \{\mu_{P_1, \dots, P_k} : P_1 \in \mathcal{P}_1, \dots, P_k \in \mathcal{P}_k\}$ with

$$\mu_{P_1, \dots, P_k}(x_1, \dots, x_k) = \prod_{i=1}^k [lik_i(P_i) P_i(x_i | pa_i(x_1, \dots, x_k))]$$

inference about the value $g(P)$ of $g : \mathcal{P} \rightarrow \mathcal{G}$ (**extension principle**):

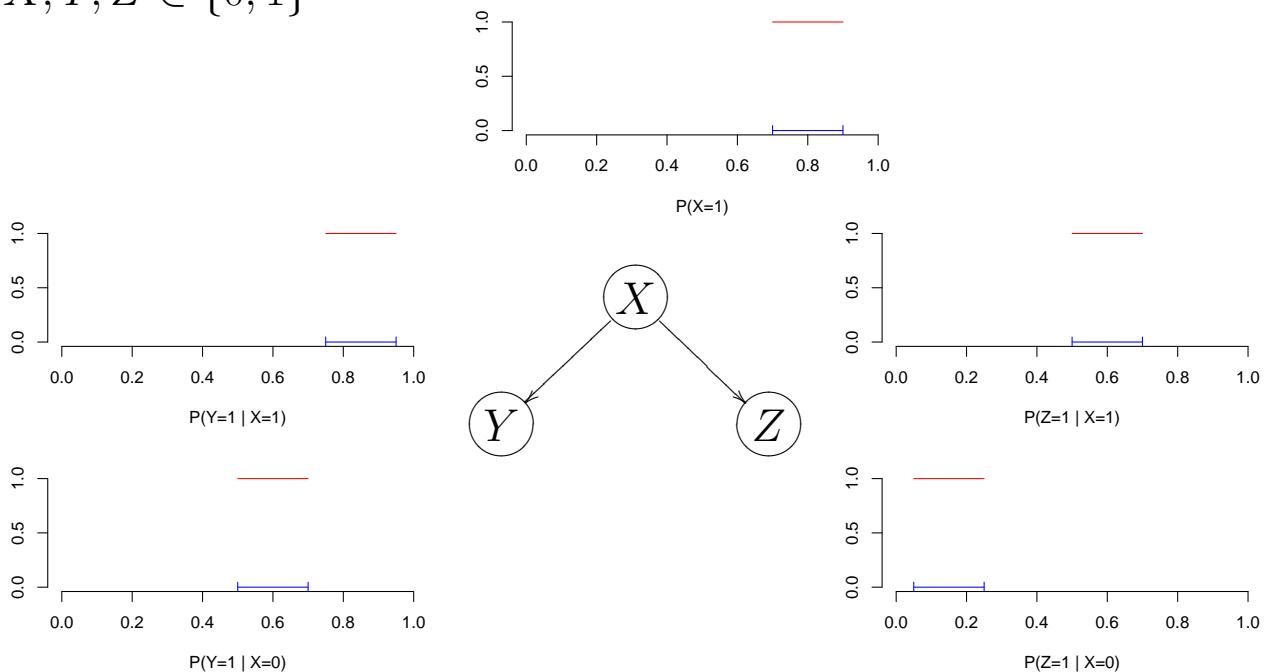
profile likelihood function $lik_g : \mathcal{G} \rightarrow (0, \infty)$ with

$$lik_g(\gamma) \propto \sup_{P \in \mathcal{P} : g(P)=\gamma} lik(P) \propto \sup_{\mu \in \mathcal{M} : g\left(\frac{\mu}{\mu(\Omega)}\right)=\gamma} \mu(\Omega)$$

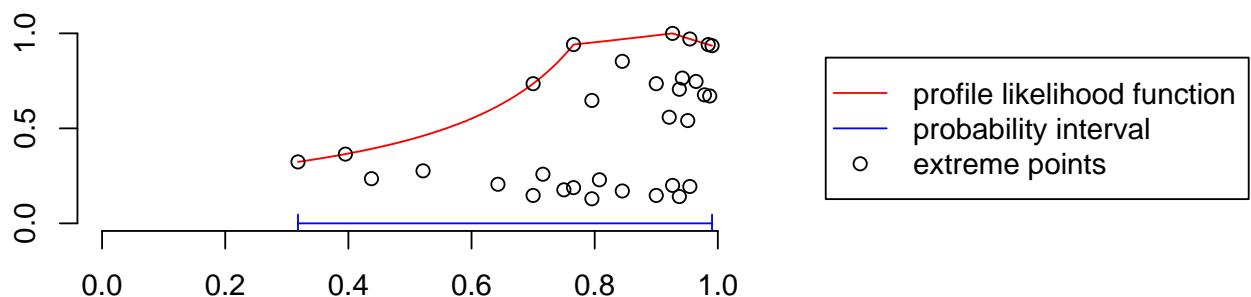
in particular (**Theorem 3**): if $\mathcal{M} = \text{ch}(\{lik(P_i) | P_i : i \in \{1, \dots, m\}\})$, then the profile likelihood function of the expected value $E_P(X)$ of a random variable X is **piecewise hyperbolic** and determined by the pairs $(E_{P_i}(X), lik(P_i))$ with $i \in \{1, \dots, m\}$

example of piecewise hyperbolic profile likelihood function:

$$X, Y, Z \in \{0, 1\}$$

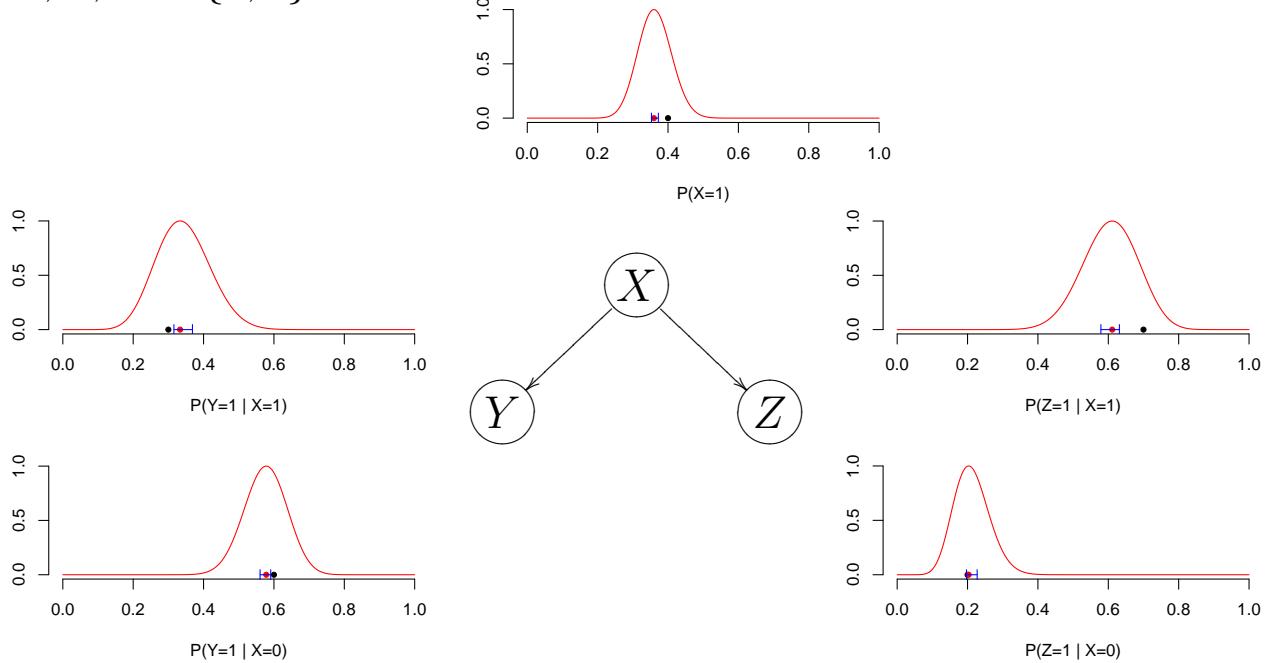


$$\Rightarrow P(X = 1 | Y = 0, Z = 1) :$$

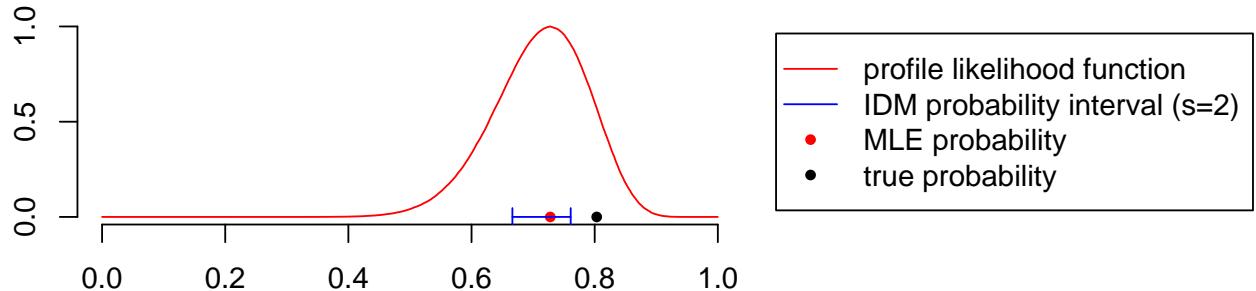


example with training data:

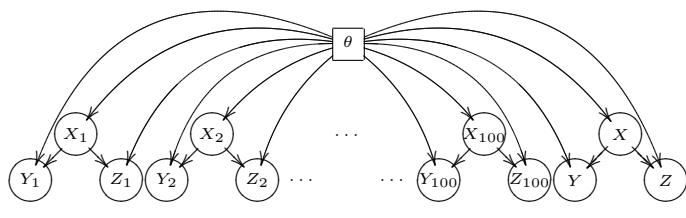
$$X, Y, Z \in \{0, 1\}$$



$$\Rightarrow P(X = 1 | Y = 0, Z = 1) :$$



global models (without and with IDMs):



$$\theta, t \in [0, 1]^5$$

prior ignorance about the root, and precise (conditional) probabilities about the other nodes

