26:198:722 Expert Systems

- Mid-Term Examination
- Possibility Theory
- Spohn's Epistemic Calculus
- Belief Functions and Assignment 5

Mid-Term Examination



- Possibility Theory is a computational implementation (largely inspired by Dubois and Prade) of Zadeh's Fuzzy Logic
- Shenoy showed that it may be re-formulated as a variant of Valuation Networks, so that possibilities may be propagated in Join Trees using the Shenoy-Shafer algorithm

A possibility function is a function

$$\pi:2^{\Omega_p}\to[0,1]$$

such that:

$$\exists x \in \Omega_p : \pi(x) = 1$$

and

$$\forall a \in 2^{\Omega_p} : \pi(a) = \max\{\pi(x) | x \in a\}$$

- By virtue of this second condition, possibility functions are completely determined by their values for singleton elements
- Intuitively, the degree of possibility for a subset a is $1-\pi(\sim a)$ and the degree of impossibility is $1-\pi(a)$

Marginalization

$$\forall y \in \Omega_{a-\{X\}} : \pi^{\downarrow a-\{X\}}(y) = \max\{\pi(y,x) | x \in \Omega_X\}$$

Combination

$$\pi_1 \otimes \pi_2(x) = \begin{cases} K^{-1}\pi_1(x^{\downarrow a})\pi_2(x^{\downarrow b}) & K \neq 0 \\ 0 & K = 0 \end{cases}$$

where
$$K = \max \left\{ \pi_1(x^{\downarrow a}) \cdot \pi_2(x^{\downarrow b}) \middle| x \in \Omega_{a \cup b} \right\}$$

Example: Marginalization

Objective 1	Objective 2	Possibility
true	true	1
true	false	.5
false	true	.3
false	false	.1

Example: Marginalization

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Objective 1 Possibility
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true 1

false .3

Objective 2 Possibility

true 1

false .5

- So for Objective 1, the degree of possibility for "true" is 0.7, and the degree of impossibility is 0
- Similarly, for Objective 2, the degree of possibility for "true" is 0.5, and the degree of impossibility is 0

Example: Combination

Objective 1 Possibility

true

false .4

Objective 1 Possibility

true 1

false .6

Example: Combination

Objective 1 Possibility

true 1

false .24

Example: Combination

Objective 1 Possibility

true 1

false .4

Objective 1 Possibility

true .6

Example: Combination

Objective 1 Possibility

true .6

false .4

Objective 1 Possibility

true 1

false .67

We can represent complete ignorance using possibilities as follows:

Objective 1	Objective 2	Possibility
true	true	1
true	false	1
false	true	1
false	false	1

■ We can define logical relationships such as "AND" nodes as follows: A= B&C:

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Α	В	C	Possibility
a	b	С	1
a	b	~C	0
a	~b	С	0
a	~b	~C	0
~a	b	С	0
~a	b	~C	1
~a	~b	С	1
~a	~b	~C	1

- Discounted "AND" nodes as discussed for probabilities and belief functions CANNOT be defined
- Dubois and Prade argue that statistical sampling is contrary to the non-probabilistic nature of possibility theory: however, it seems that it could be incorporated using normalized maximum likelihood functions

Joint possibilities generally cannot be uniquely determined from marginals, as for probabilities and belief functions:

X	Y	P1	P2	P3
X	у	1	1	1
X	~y	.4	.4	.4
~X	У	.2	.2	0
~X	~y	.2	.1	.2

have the same marginals for X and Y.

However, when combined with an "AND" node, all three produce the same results!!!

- When multiple nodes are combined in an "AND" node, the effect is that the degree of possibility for the conjunction is equal to the degree of possibility for the least possible conjunct
- This is significantly different from probabilities and belief functions, and is very relevant, for example, to auditing

- Spohn introduced his theory of epistemic states in order to represent plain human beliefs in a nonprobabilistic way easily amenable to revision
- Initially they were based on functions mapping into the ordinals; later he changed this to the natural numbers

For technical reasons, we will define disbeliefs as functions mapping to the natural numbers extended by adding a representation of infinity, ∞

A disbelief function is a function

$$\delta: 2^{\Omega_d} \to N^+$$

such that:

$$\exists x \in \Omega_p : \pi(x) = 0$$

and

$$\forall a \in 2^{\Omega_d} : \delta(a) = \min\{\pi(x) | x \in a\}$$

- By virtue of this second condition, disbelief functions are completely determined by their values for singleton elements
- Intuitively, the degree of disbelief for a subset a is $\delta(a)$ and the degree of belief is $\delta(\sim a)$

Marginalization

$$\forall y \in \Omega_{a-\{X\}} : \delta^{\downarrow a-\{X\}}(y) = \min\{\delta(y,x) | x \in \Omega_X\}$$

Combination

$$\delta_1 \otimes \delta_2(x) = \delta_1(x^{\downarrow a}) + \delta_2(x^{\downarrow b}) - K$$

where
$$K = \min \left\{ \delta_1(x^{\downarrow a}) + \delta_2(x^{\downarrow b}) \middle| x \in \Omega_{a \cup b} \right\}$$

Example: Marginalization

Objective 1	Objective 2	Disbelief	
true	true	0	
true	false	5	
false	true	7	
false	false	9	

Example: Marginalization

Objective 1 Disbelief

true 0

false 7

Objective 2 Disbelief

true 0

Example: Combination

Objective 1 Disbelief

true 0

false 6

Objective 1 Disbelief

true 0

Example: Combination

Objective 1 Disbelief

true 0

Example: Combination

Objective 1 Disbelief

true 0

false 6

Objective 1 Disbelief

true 4

Example: Combination

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Objective 1 Disbelief
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true 4

false 6

Objective 1 Disbelief

true 0

We can represent complete ignorance using disbeliefs as follows:

Objective 1	Objective 2	Disbelief
true	true	0
true	false	0
false	true	0
false	false	0

■ We can define logical relationships such as "AND" nodes as follows: A= B&C:

Α	В	С	Disbelie
а	b	С	0
а	b	~C	∞
a	~b	С	∞
а	~b	~C	∞
~a	b	С	∞
~a	b	~C	0
~a	~b	С	0
~a	~b	~C	0

- Discounted "AND" nodes as discussed for probabilities and belief functions CANNOT be defined
- Statistical sampling is contrary to the intended ordinal nature of epistemic (dis)beliefs

Joint disbeliefs generally cannot be uniquely determined from marginals, as for probabilities and belief functions:

X	Υ	P1	P2	P3
X	У	0	0	0
X	~y	4	4	4
~X	У	7	7	11
~X	~y	7	74	7

have the same marginals for X and Y.

However, when combined with an "AND" node, all three produce the same results!!!

- When multiple nodes are combined in an "AND" node, the effect is that the degree of belief for the conjunction is equal to the degree of belief for the least believed conjunct
- This is significantly different from probabilities and belief functions, and is very relevant, for example, to auditing

In addition, Spohn's system offers the prospect of elicitation of ordinal rankings rather than highly sensitive real values

Belief Functions

