Fuzzy Sets and Fuzzy Logic

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(From a press release)

Equens to offer RiskShield Fraud Protection for Card Payments

Today Equens, one of the largest pan-European card and payment Processors, announced that it has selected RiskShield from INFORM GmbH as the basis for a new approach to fraud detection and behaviour monitoring. By utilising the flexibility offered by RiskShield, Equens will be able to offer tailor-made fraud management services to issuers and acquirers.

UTRECHT, The Netherlands, 30/10/2012

 From the brochure of "RiskShield"



RiskShield

is one of the world's leading software products for risk assessment and fraud prevention. RiskShield customers are typically banks, payment and processing service providers, telecommunication and insurance companies.



ADVANTAGES OF

RiskShield

Based on Fuzzy Logic

Evaluates historical data such as typical user or contractual partner behaviour

Maximum performance (decisions within milliseconds)

High transparency for decisions

Short reaction times to new risk and fraud patterns

Reduces losses due to fraud or a missing risk assessment that can amount to millions

Crisp sets

 Collection of definite, well-definable objects (elements).

Representation of sets:

- list of all elements
 - $A=\{x_1, \ldots, x_n\}, x_j \in X$
- elements with property P
 A={x|x satisfies P},x ∈ X
- Venn diagram



 characteristic function $f_{\Delta}: X \rightarrow \{0,1\},$ $f_{\Delta}(x) = 1, \Leftrightarrow x \in A$ $f_{\Delta}(x) = 0, \Leftrightarrow x \notin A$ **Real numbers larger than 3:** 3

Crisp (traditional) logic

 Crisp sets are used to define interpretations of first order logic

If *P* is a unary predicate, and we have no functions, a possible interpretation is $A = \{0,1,2\}$ $P^{I} = \{0,2\}$ within this interpretation, *P*(*o*) and *P*(*2*) are true,

and P(1) is false.

 Crisp logic can be "fragile": changing the interpretation a little can change the truth value of a formula dramatically.

Fuzzy sets

- Sets with fuzzy, gradual boundaries (Zadeh 1965)
- A fuzzy set A in X is characterized by its membership function $\mu_A: X \rightarrow [0,1]$
- A fuzzy set A is completely determined by the set of ordered pairs
 - $A=\{(x,\mu_A(x))|\ x\in\ X\}$
- X is called the *domain* or *universe of discourse*

Real numbers about 3:



Fuzzy sets on discrete universes

Fuzzy set C = "desirable city to live in" X = {SF, Boston, LA} (discrete and non-ordered) $C = \{(SF, 0.9), (Boston, 0.8), (LA, 0.6)\}$ Fuzzy set A = "sensible number of children" X = {0, 1, 2, 3, 4, 5, 6} (discrete universe) $A = \{(0, .1), (1, .3), (2, .7), (3, 1), (4, .6), (5, .2), (6, .1)\}$ S 1 U U 0.8 ershi 0.6 0.4 E

0.2

0

0

2

Number of Children

 \geq

Fuzzy sets on continuous universes

Fuzzy set B = "about 50 years old"

X = Set of positive real numbers (continuous)



Membership Function formulation

trimf
$$(x; a, b, c) = \max\left(\min\left(\frac{x-a}{b-a}, \frac{c-x}{c-b}\right), 0\right)$$

Trapezoidal MF:

Triangular MF:

trapmf (x; a, b, c, d) = max
$$\left(\min\left(\frac{x-a}{b-a}, 1, \frac{d-x}{d-c}\right), 0 \right)$$

Gaussian MF:

$$gaussmf(x;a,b) = e^{-\frac{1}{2}\left(\frac{x-a}{b}\right)^2}$$

Generalized bell MF: $gbellmf(x;a,b,c) = \frac{1}{1 + \left|\frac{x-c}{b}\right|^{2a}}$

MF formulation



Fuzzy sets & fuzzy logic

- Fuzzy sets can be used to define a level of truth of facts
- Fuzzy set C = "desirable city to live in"
 - X = {SF, Boston, LA} (discrete and non-ordered)
 - $C = \{(SF, 0.9), (Boston, 0.8), (LA, 0.6)\}$

corresponds to a fuzzy interpretation in which C(SF) is true with degree 0.9 C(Boston) is true with degree 0.8 C(LA) is true with degree 0.6 \rightarrow membership function $\mu_C(x)$ can be seen as a (fuzzy) predicate.

Notation

Many texts (especially older ones) do not use a consistent and clear notation

X is discrete

$$A = \sum_{x_i \in X} \mu_A(x_i) / x_i$$

$$A = \sum_{x_i \in X} \mu_A(x_i) x_i$$

X is continuous

$$A = \int_{X} \mu_A(x) / x$$
$$A = \int_{X} \mu_A(x) x$$

Note that Σ and integral signs stand for the union of membership grades; "/" stands for a marker and does not imply division.

Fuzzy partition

Fuzzy partition formed by the linguistic values "young", "middle aged", and "old":



Fuzzy logic formulas

- Membership functions: $\mu_C(x), \mu_B(x)$
 - B="City is beautiful"
 - C="City is clean"

Formulas:
$$\mu_C(x) \land \neg \mu_B(x), \mu_C(x) \lor \mu_B(x), ...$$

- What is the truth value of such formulas for given x?
- We need to define a meaning for the connectives

Set theoretic operations /Fuzzy logic connectives

(Specific case)

• Subset:

$$A \subseteq B \Leftrightarrow \mu_A \leq \mu_B$$

Complement:

 $A = X - A \Leftrightarrow \mu_{\overline{A}}(x) = 1 - \mu_{A}(x)$ • Union:

 $C = A \cup B \Leftrightarrow \mu_c(x) = \max(\mu_A(x), \mu_B(x)) = \mu_A(x) \lor \mu_B(x)$

Intersection:

 $C = A \cap B \Leftrightarrow \mu_c(x) = \min(\mu_A(x), \mu_B(x)) = \mu_A(x) \land \mu_B(x)$

Set theoretic operations





Generalized negation

- General requirements:
 - Boundary: N(0)=1 and N(1)=0
 - Monotonicity: N(a) > N(b) if a < b
 - Involution: N(N(a)) = a
- Two types of fuzzy complements:
 - Sugeno's complement:

$$N_s(a) = \frac{1-a}{1+sa}$$

- Yager's complement:

$$N_w(a) = (1 - a^w)^{1/w}$$

Sugeno's complement: Yager's complement:

$$N_s(a) = \frac{1-a}{1+sa}$$



$$N_w(a) = (1 - a^w)^{1/w}$$



Generalized intersection (Triangular/T-norm, logical and)

- Basic requirements:
 - Boundary: T(0, a) = T(a, 0) = 0, T(a, 1) = T(1, a) = a
 - Monotonicity: T(a, b) <= T(c, d) if a <= c and b <= d</p>
 - Commutativity: T(a, b) = T(b, a)
 - Associativity: T(a, T(b, c)) = T(T(a, b), c)

Generalized intersection (Triangular/T-norm)

- Examples:
 - Minimum:
 - Algebraic product:
 - Bounded product:
 - Drastic product:

$$T(a,b) = min(a,b)$$

$$T(a,b) = a \cdot b$$

$$T(a,b) = max(0,(a+b-1))$$

$$T(a,b) = \begin{cases} a & if \ b=1 \\ b & if \ a=1 \\ 0 & otherwise \end{cases}$$

T-norm operator

Minimum: T_m(a, b)







Algebraic

product:



Bounded product: T_b(a, b)

2











Generalized union (t-conorm)

- Basic requirements:
 - Boundary: S(1, a) = 1, S(a, 0) = S(0, a) = a
 - Monotonicity: S(a, b) < S(c, d) if a < c and b < d</p>
 - Commutativity: S(a, b) = S(b, a)
 - Associativity: S(a, S(b, c)) = S(S(a, b), c)
- Examples:
 - Maximum:

$$S(a,b) = max(a,b)$$

- Algebraic sum:
- Bounded sum:
- Drastic sum

$$S(a,b) = a + b - a \cdot b$$

$$S(a, b) = min(1, (a+b))$$

T-conorm operator

 $\begin{array}{ll} \text{Maximum:} & \leq \\ & S_{m}(a, b) \end{array}$



Bounded sum: S₀(a, b)

 \leq

Drastic sum: Sd(a, b)

<



 $X = X^{10}$

0.5

20

Y = y

0

0













Generalized De Morgan's Law

- T-norms and T-conorms are duals which support the generalization of DeMorgan's law:
 - T(a, b) = N(S(N(a), N(b)))
 - S(a, b) = N(T(N(a), N(b)))

