Fuzzy Sets and Fuzzy Logic



EQUENS

Sound solutions, solid results



News

- > PaySquare connected to Discover network
- > Interview with Paolo Zaccardi, member of EISB
- > Inside Sibos: Daily reports
- More news items

Publications

- > Article Future Banking
- > EQ Magazine 1st edition
- > EQ Magazine 2nd edition
- More publications

Support

Connect with us

Discover our SEPA products

(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

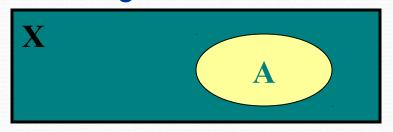
 Collection of definite, well-definable objects (elements).

Representation of sets:

list of all elements

$$A=\{x_1,\;\ldots,x_n\},\;x_j\in\;X$$

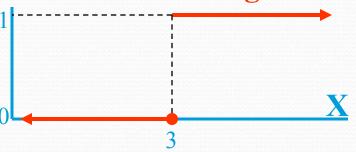
- elements with property P
 A={x|x satisfies P},x ∈ X
- Venn diagram



characteristic function

$$f_A: X \to \{0,1\},$$
 $f_A(x) = 1, \Leftrightarrow x \in A$
 $f_A(x) = 0, \Leftrightarrow x \notin A$

Real numbers larger than 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

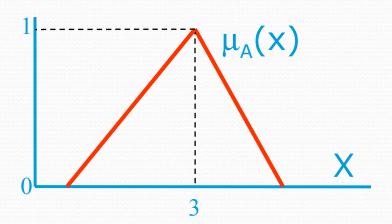
- Sets with fuzzy, gradual boundaries (Zadeh 1965)
- A fuzzy set A in X is characterized by its membership function $\mu_A: X \to [0,1]$

A fuzzy set A is completely determined by the set of ordered pairs

$$A=\{(x,\mu_A(x))|x\in X\}$$

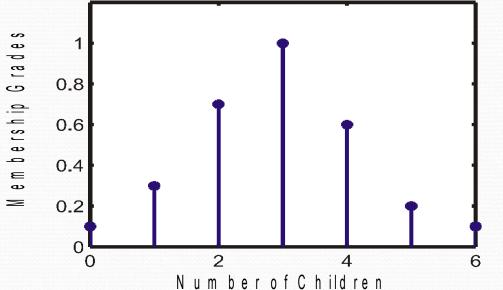
X is called the *domain* or *universe* of *discourse*

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, o.9), (Boston, o.8), (LA, o.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)}



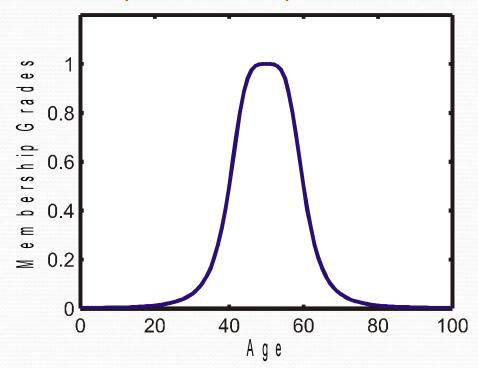
Fuzzy sets on continuous universes

Fuzzy set B = "about 50 years old"

X = Set of positive real numbers (continuous)

B =
$$\{(x, \mu_B(x)) | x \text{ in } X\}$$

$$\mu_B(x) = \frac{1}{1 + \left(\frac{x - 50}{10}\right)^2} = \frac{\frac{1}{2} \cdot 0.6}{\frac{2}{3} \cdot 0.4}$$



Membership Function formulation

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

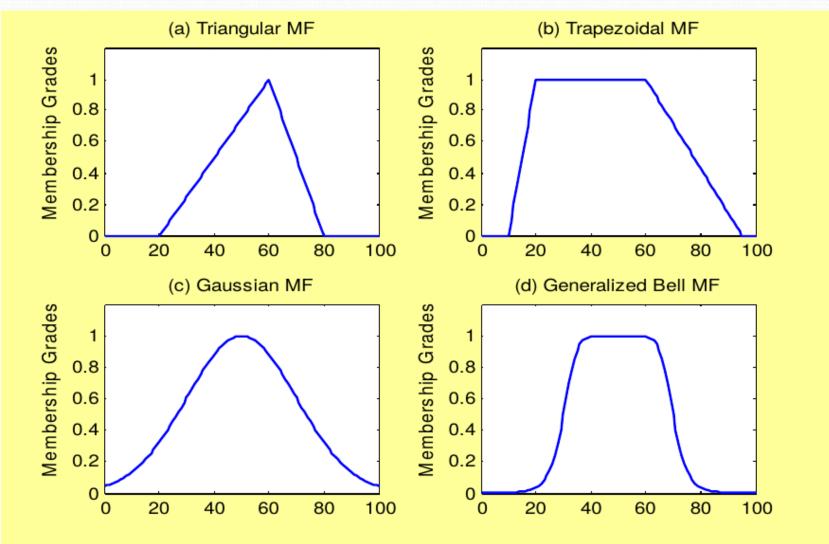
trapmf
$$(x; a, b, c, d) = \max \left(\min \left(\frac{x-a}{b-a}, 1, \frac{d-x}{d-c} \right), 0 \right)$$

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

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, o.9), (Boston, o.8), (LA, o.6)}
```

corresponds to a fuzzy interpretation in which

C(*SF*) is true with degree 0.9

C(*Boston*) is true with degree o.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

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

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

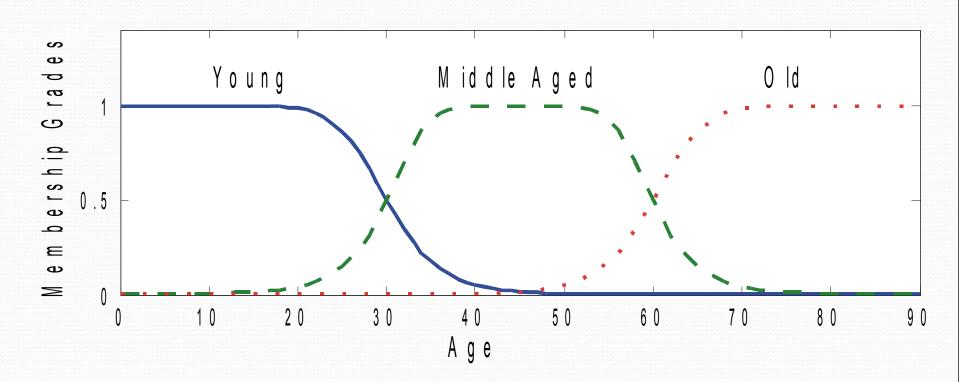
X is continuous

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

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:

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

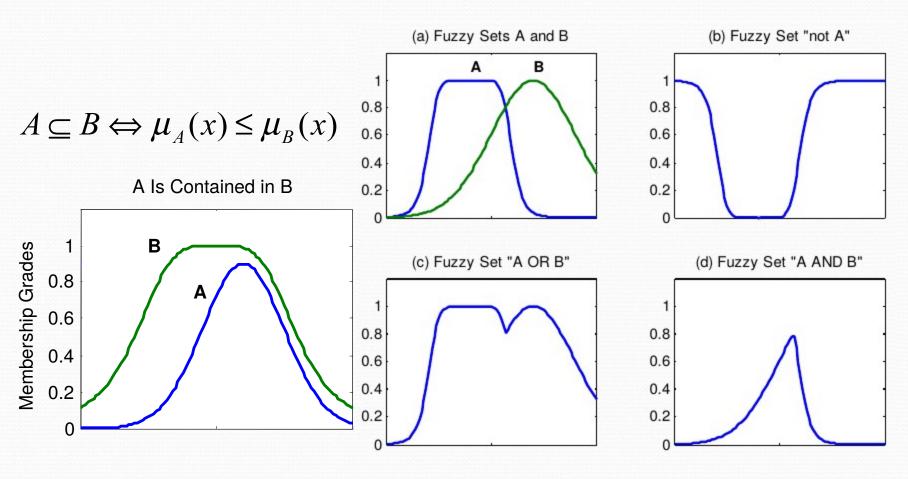
Union:

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

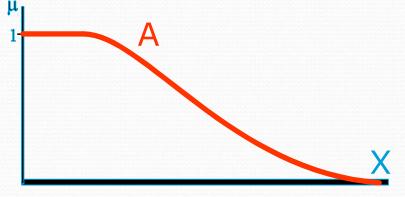
Intersection:

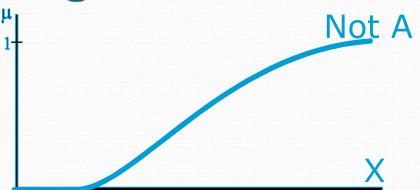
$$C = A \cap B \Leftrightarrow \mu_c(x) = \min(\mu_A(x), \mu_B(x)) = \mu_A(x) \land \mu_B(x)$$

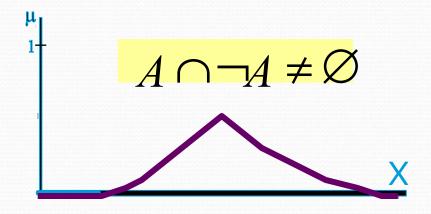
Set theoretic operations

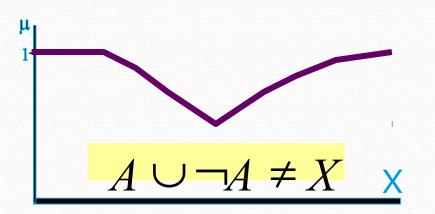


Combinations with negation









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:

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

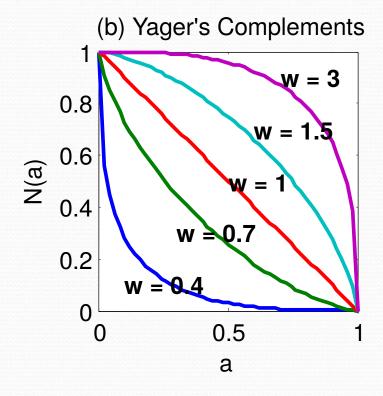
Sugeno's and Yager's complements

Sugeno's complement:

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

Yager's complement:

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



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:

- Drastic product:

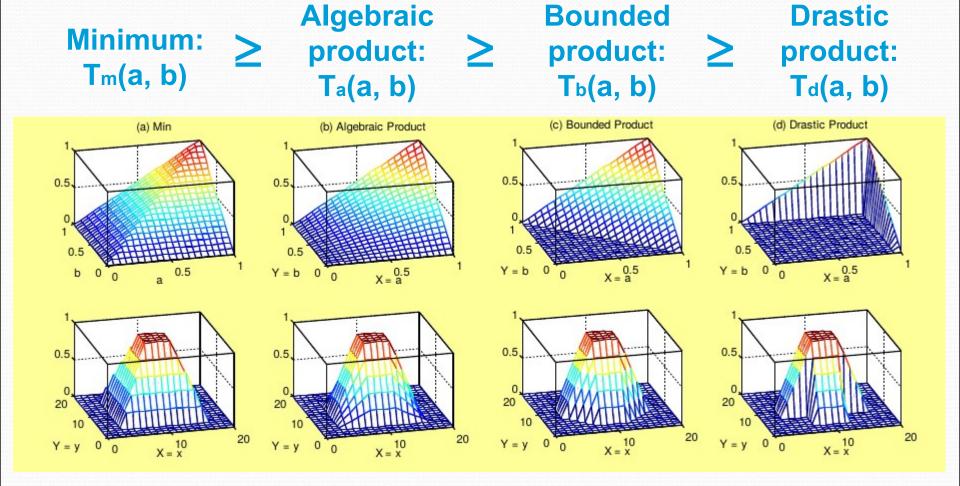
$$T(a,b) = min(a,b)$$

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

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

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

T-norm operator



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
 - 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: $S(a,b) = a + b a \cdot b$
 - Bounded sum: S(a,b)=min(1,(a+b))
 - Drastic sum

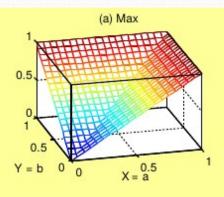
T-conorm operator

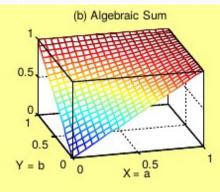
Maximum: ≤ S_m(a, b)

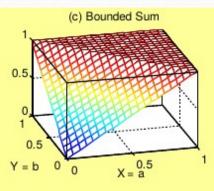
Algebraic sum: Sa(a, b)

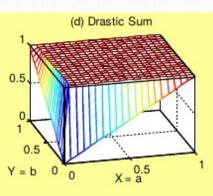
Bounded sum:
Sb(a, b)

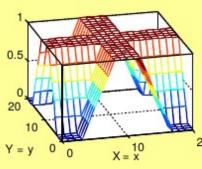
Drastic
 sum:
 Sd(a, b)

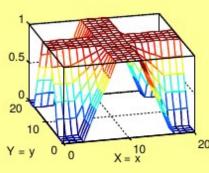


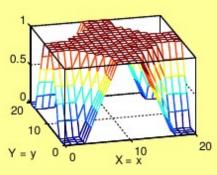


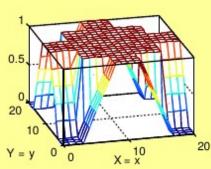












Generalized De Morgan's Law

- 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)))

