

Fuzzy Set Operators

Prof. Dr. Rudolf Kruse

In set theory, **operators** are defined by **propositional logics operator**

Let X be universal set (often called universe of discourse). Then we define

$$A \cap B = \{x \in X \mid x \in A \wedge x \in B\}$$

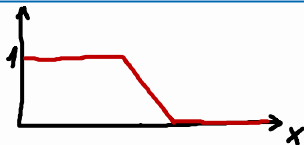
$$A \cup B = \{x \in X \mid x \in A \vee x \in B\}$$

$$A^c = \{x \in X \mid x \notin A\} = \{x \in X \mid \neg (x \in A)\}$$

$A \subseteq B$ if and only if $(x \in A) \rightarrow (x \in B)$ for all $x \in X$

Fuzzy Set Operators can be defined by using **multivalued logics operators**

Standard Fuzzy Set Operators



$$(\mu \wedge \mu')(x) := \min\{\mu(x), \mu'(x)\}$$

intersection ("AND"),

$$(\mu \vee \mu')(x) := \max\{\mu(x), \mu'(x)\}$$

union ("OR"),

$$\neg\mu(x) := 1 - \mu(x)$$

complement ("NOT").

μ is subset of μ' if and only if $\mu \leq \mu'$.

Theorem

$(F(X), \wedge, \vee, \neg)$ is a complete distributive lattice, but no Boolean algebra.

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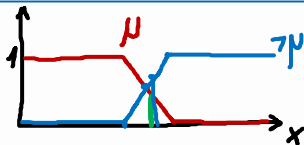
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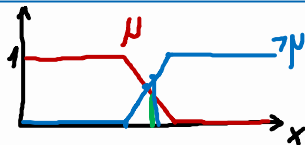
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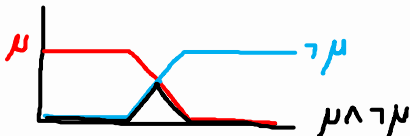
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Fuzzy Set Complement

$$\bar{\mu}(x) = 1 - \mu(x)$$

a	$\bar{a} = 1 - a$
0	1
0.3	0.7
1	0



$$N: [0, 1] \rightarrow [0, 1]$$

Fuzzy Complement/Fuzzy Negation

Definition

Let X be a given set and $\mu \in \mathcal{F}(X)$. Then the *complement* $\bar{\mu}$ can be defined pointwise by $\bar{\mu}(x) := \sim(\mu(x))$ where $\sim : [0, 1] \rightarrow [0, 1]$ satisfies the conditions

$$\sim(0) = 1, \quad \sim(1) = 0$$

and

for $x, y \in [0, 1]$, $x \leq y \implies \sim x \geq \sim y$ (\sim is non-increasing).

Abbreviation: $\sim x := \sim(x)$

Strict and Strong Negations

Additional properties may be required

- $x, y \in [0, 1], x < y \implies \sim x > \sim y$ (\sim is strictly decreasing)
- \sim is continuous
- $\sim \sim x = x$ for all $x \in [0, 1]$ (\sim is involutive)

According to conditions, two subclasses of negations are defined:

Definition

A negation is called *strict* if it is also strictly decreasing and continuous. A strict negation is said to be *strong* if it is involutive, too.

$\sim x = 1 - x^2$, for instance, is strict, not strong, thus not involutive

Families of Negations

standard negation:

$$\sim x = 1 - x$$

threshold negation:

$$\sim_{\theta}(x) = \begin{cases} 1 & \text{if } x \leq \theta \\ 0 & \text{otherwise} \end{cases}$$

Cosine negation:

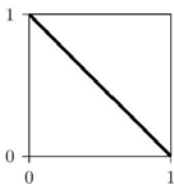
$$\sim x = \frac{1}{2} (1 + \cos(\pi x))$$

Sugeno negation:

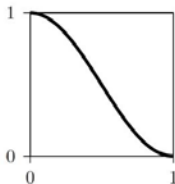
$$\sim_{\lambda}(x) = \frac{1-x}{1+\lambda x}, \quad \lambda > -1$$

Yager negation:

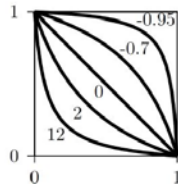
$$\sim_{\lambda}(x) = (1 - x^{\lambda})^{\frac{1}{\lambda}}$$



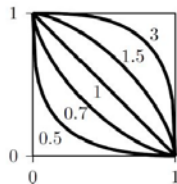
standard



cosine

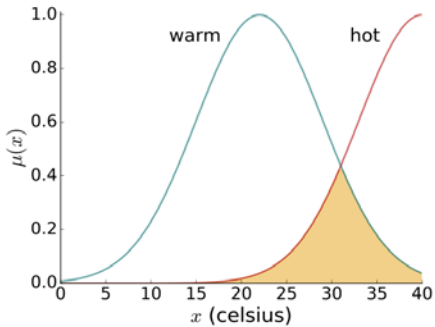


Sugeno

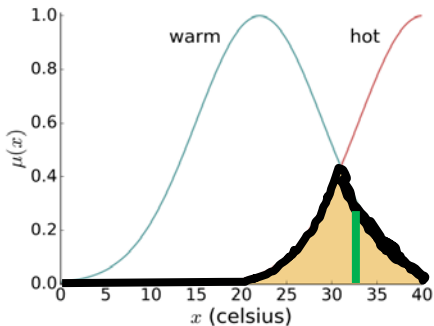


Yager

Fuzzy Set Intersection and Union



warm and hot ?



Zadeh' Intersection

a and b = min (a,b), for all membership degrees a,b

$(\mu_{\text{warm}} \cap \mu_{\text{hot}})(x) = \min(\mu_{\text{warm}}(x), \mu_{\text{hot}}(x))$, for all real numbers x

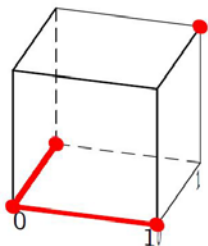
Classical Intersection and Union

Classical set intersection represents logical conjunction.

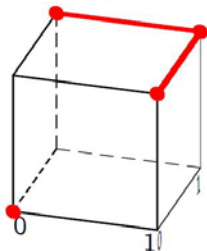
Classical set union represents logical disjunction.

Generalization from $\{0, 1\}$ to $[0, 1]$ as follows:

$x \wedge y$	0	1
0	0	0
1	0	1



$x \vee y$	0	1
0	0	1
1	1	1



Fuzzy Set Intersection and Union

Let A, B be fuzzy subsets of X , *i.e.* $A, B \in F(X)$.

Their **intersection** and **union** are often defined pointwise using:

$$(A \cap B)(x) = \top(A(x), B(x)) \quad \text{where} \quad \top : [0, 1]^2 \rightarrow [0, 1]$$

$$(A \cup B)(x) = \perp(A(x), B(x)) \quad \text{where} \quad \perp : [0, 1]^2 \rightarrow [0, 1].$$

Triangular Norms and Conorms

T is a *triangular norm (t-norm)* $\iff T$ satisfies conditions T1-T4

\perp is a *triangular conorm (t-conorm)* $\iff \perp$ satisfies C1-C4

Identity Law

T1: $T(x, 1) = x$

C1: $\perp(x, 0) = x$

Commutativity

T2: $T(x, y) = T(y, x)$

C2: $\perp(x, y) = \perp(y, x)$

Associativity

T3: $T(x, T(y, z)) = T(T(x, y), z)$

C3: $\perp(x, \perp(y, z)) = \perp(\perp(x, y), z)$

Monotonicity

T4: $y \leq z$ implies $T(x, y) \leq T(x, z)$ **C4:** $y \leq z$ implies $\perp(x, y) \leq \perp(x, z)$.

Triangular Norms and Conorms II

Both identity law and monotonicity respectively imply

$$\forall x \in [0, 1] : T(0, x) = 0,$$

$$\forall x \in [0, 1] : \perp(1, x) = 1,$$

For any t -norm $T : T(x, y) \leq \min(x, y)$, for any t -conorm $\perp : \perp(x, y) \geq \max(x, y)$.

$$x = 1 \Rightarrow T(0, 1) = 0 \text{ and}$$

$$x \leq 1 \Rightarrow T(x, 0) \leq T(1, 0) = T(0, 1) = 0$$

De Morgan Triplet I

For every T and strong negation \sim , one can define t -conorm \perp by

$$\perp(x, y) = \sim T(\sim x, \sim y), \quad x, y \in [0, 1].$$

Additionally, in this case $T(x, y) = \sim \perp(\sim x, \sim y)$, $x, y \in [0, 1]$.

De Morgan Triplet II

Definition

The triplet (T, \perp, \sim) is called *De Morgan triplet* if and only if T is t -norm, \perp is t -conorm, \sim is strong negation, T, \perp and \sim satisfy $\perp(x, y) = \sim T(\sim x, \sim y)$.

In the following, some important De Morgan triplets will be shown, only the most frequently used and important ones.

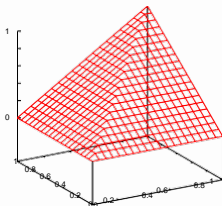
In all cases, the standard negation $\sim x = 1 - x$ is considered.

The Minimum and Maximum I

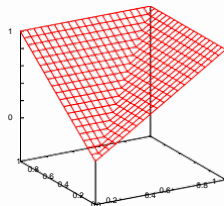
$$T_{\min}(x, y) = \min(x, y), \quad \perp_{\max}(x, y) = \max(x, y)$$

Minimum is the greatest t -norm and max is the weakest t -conorm.

$T(x, y) \leq \min(x, y)$ and $\perp(x, y) \geq \max(x, y)$ for any T and \perp



T_{\min}



\perp_{\max}

The Special Role of Minimum and Maximum I

T_{\min} and \perp_{\max} play key role for intersection and union, resp. In a practical sense, they are very simple.

Apart from the identity law, commutativity, associativity and monotonicity, they also satisfy the following properties for all $x, y, z \in [0, 1]$:

Distributivity

$$\perp_{\max}(x, T_{\min}(y, z)) = T_{\min}(\perp_{\max}(x, y), \perp_{\max}(x, z)),$$

$$T_{\min}(x, \perp_{\max}(y, z)) = \perp_{\max}(T_{\min}(x, y), T_{\min}(x, z))$$

Continuity

T_{\min} and \perp_{\max} are continuous.

The Special Role of Minimum and Maximum II

Strict monotonicity on the diagonal

$x < y$ implies $\top_{\min}(x, x) < \top_{\min}(y, y)$ and $\perp_{\max}(x, x) < \perp_{\max}(y, y)$.

Idempotency

$$\top_{\min}(x, x) = x, \quad \perp_{\max}(x, x) = x$$

Absorption

$$\top_{\min}(x, \perp_{\max}(x, y)) = x, \quad \perp_{\max}(x, \top_{\min}(x, y)) = x$$

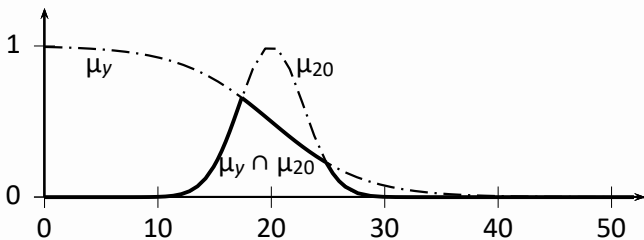
Non-compensation

$x < y < z$ imply $\top_{\min}(x, z) \neq \top_{\min}(y, y)$ and
 $\perp_{\max}(x, z) \neq \perp_{\max}(y, y)$.

The Minimum and Maximum II

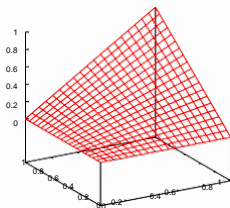
T_{\min} and \perp_{\max} can be easily processed numerically and visually,
e.g. linguistic values *young* and *approx. 20* described by μ_y , μ_{20} .

$T_{\min}(\mu_y, \mu_{20})$ is shown below.

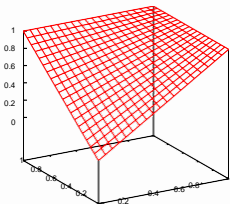


The Product and Probabilistic Sum

$$T_{\text{prod}}(x, y) = x \cdot y, \quad \perp_{\text{sum}}(x, y) = x + y - x \cdot y$$



T_{prod}



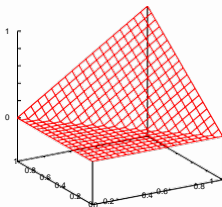
\perp_{sum}

The Łukasiewicz t -norm and t -conorm

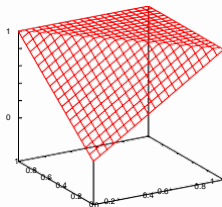
$$T_{\text{Łuka}}(x, y) = \max\{0, x + y - 1\},$$

$$\perp_{\text{Łuka}}(x, y) = \min\{1, x + y\}$$

$T_{\text{Łuka}}$, $\perp_{\text{Łuka}}$ are also called *bold intersection* and *bounded sum*.



$T_{\text{Łuka}}$



$\perp_{\text{Łuka}}$

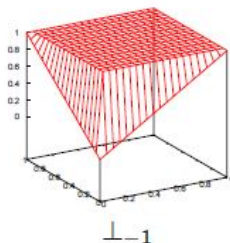
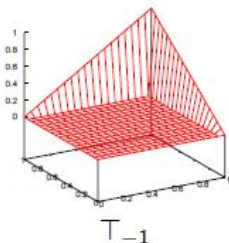
The Drastic Product and Sum

$$T_{-1}(x, y) = \begin{cases} \min(x, y) & \text{if } \max(x, y) = 1 \\ 0 & \text{otherwise} \end{cases}$$

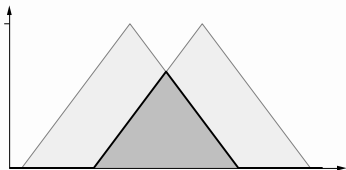
$$\perp_{-1}(x, y) = \begin{cases} \max(x, y) & \text{if } \min(x, y) = 0 \\ 1 & \text{otherwise} \end{cases}$$

T_{-1} is the weakest t -norm, \perp_{-1} is the strongest t -conorm.

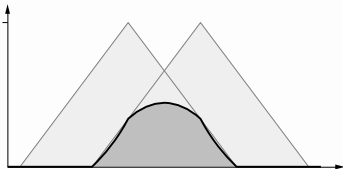
$T_{-1} \leq T \leq T_{\min}$, $\perp_{\max} \leq \perp \leq \perp_{-1}$ for any T and \perp



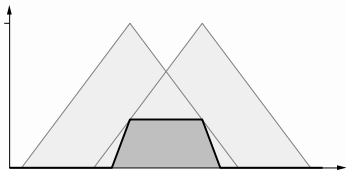
Examples of Fuzzy Intersections



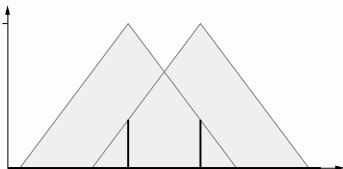
t -norm T_{\min}



t -norm T_{prod}



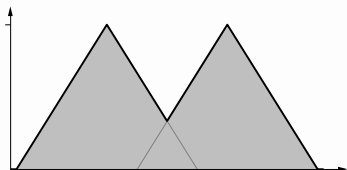
t -norm $T_{\text{Łuka}}$



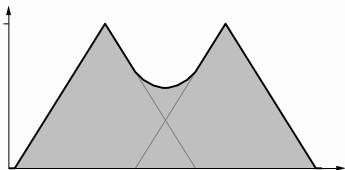
t -norm T_{-1}

Note that all fuzzy intersections are contained within upper left graph and lower right one.

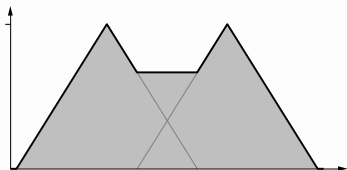
Examples of Fuzzy Unions



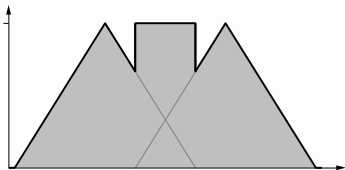
t -conorm \perp_{\max}



t -conorm \perp_{sum}



t -conorm \perp_{Luka}



t -conorm \perp_{-1}

Note that all fuzzy unions are contained within upper left graph and lower right one.

Continuous Archimedean t -norms and t -conorms

Often it is possible to representation functions with several inputs by a function with only one input , *e.g.*

$$K(x, y) = f^{(-1)}(f(x) + f(y))$$

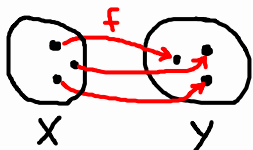
For a subclass of t -norms this is possible. The trick makes calculations simpler.

A t -norm T is called

- (a) *continuous* if T is continuous
- (b) *Archimedean* if T is continuous and $T(x, x) < x$ for all $x \in]0, 1[$.

A t -conorm \perp is called

- (a) *continuous* if \perp is continuous,
- (b) *Archimedean* if \perp is continuous and $\perp(x, x) > x$ for all $x \in]0, 1[$.

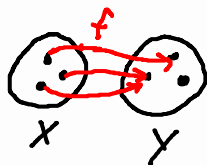
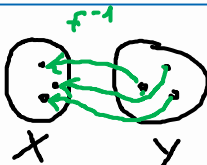
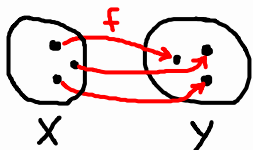


The concept of a pseudoinverse

Definition

Let $f : [a, b] \rightarrow [c, d]$ be a monotone function between two closed subintervals of extended real line. The pseudoinverse function to f is the function $f^{(-1)} : [c, d] \rightarrow [a, b]$ defined as

$$f^{(-1)}(y) = \begin{cases} \sup\{x \in [a, b] \mid f(x) < y\} & \text{for } f \text{ non-decreasing,} \\ \sup\{x \in [a, b] \mid f(x) > y\} & \text{for } f \text{ non-increasing.} \end{cases}$$



f^{-1} does not exist

The concept of a pseudoinverse

Definition

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$$f^{(-1)}(y) = \begin{cases} \sup\{x \in [a, b] \mid f(x) < y\} & \text{for } f \text{ non-decreasing,} \\ \sup\{x \in [a, b] \mid f(x) > y\} & \text{for } f \text{ non-increasing.} \end{cases}$$

Archimedean t -norms

Theorem

A t -norm \mathbb{T} is Archimedean if and only if there exists a strictly decreasing and continuous function $f : [0, 1] \rightarrow [0, \infty)$ with $f(1) = 0$ such that

$$\mathbb{T}(x, y) = f^{(-1)}(f(x) + f(y)) \quad (1)$$

where

$$f^{(-1)}(x) = \begin{cases} f^{-1}(x) & \text{if } x \leq f(0) \\ 0 & \text{otherwise} \end{cases}$$

is the pseudoinverse of f . Moreover, this representation is unique up to a positive multiplicative constant.

\mathbb{T} is generated by f if \mathbb{T} has representation (1).

f is called additive generator of \mathbb{T} .

Additive Generators of t -norms – Examples

Find an additive generator f of $T_{\text{Luka}}(x, y) = \max\{x + y - 1, 0\}$.

for instance $f_{\text{Luka}}(x) = 1 - x$

then, $f_{\text{Luka}}^{(-1)}(x) = \max\{1 - x, 0\}$

thus $T_{\text{Luka}}(x, y) = f_{\text{Luka}}^{(-1)}(f_{\text{Luka}}(x) + f_{\text{Luka}}(y))$

Find an additive generator f of $T_{\text{prod}}(x, y) = x \cdot y$.

to be discussed in the exercise

hint: use of logarithmic and exponential function

Archimedean t -conorms

Theorem

A t -conorm \perp is Archimedean if and only if there exists a strictly increasing and continuous function $g : [0, 1] \rightarrow [0, \infty]$ with $g(0) = 0$ such that

$$\perp(x, y) = g^{(-1)}(g(x) + g(y)) \quad (2)$$

where

$$g^{(-1)}(x) = \begin{cases} g^{-1}(x) & \text{if } x \leq g(1) \\ 1 & \text{otherwise} \end{cases}$$

is the pseudoinverse of g . Moreover, this representation is unique up to a positive multiplicative constant.

\perp is generated by g if \perp has representation (2).

g is called additive generator of \perp .

Additive Generators of t -conorms – Two Examples

Find an additive generator g of $\perp_{\text{Luka}}(x, y) = \min\{x + y, 1\}$.

for instance $g_{\text{Luka}}(x) = x$

then, $g_{\text{Luka}}^{(-1)}(x) = \min\{x, 1\}$

thus $\perp_{\text{Luka}}(x, y) = g_{\text{Luka}}^{(-1)}(g_{\text{Luka}}(x) + g_{\text{Luka}}(y))$

Find an additive generator g of $\perp_{\text{sum}}(x, y) = x + y - x \cdot y$.

to be discussed in the exercise

hint: use of logarithmic and exponential function

Now, let us examine some typical families of operations.

Sugeno-Weber Family I

For $\lambda > -1$ and $x, y \in [0, 1]$, define

$$T_{\lambda}(x, y) = \max \left\{ \frac{x + y - 1 + \lambda xy}{1 + \lambda}, 0 \right\},$$

$$\perp_{\lambda}(x, y) = \min \{x + y + \lambda xy, 1\}.$$

$\lambda = 0$ leads to $T_{\text{Łuka}}$ and $\perp_{\text{Łuka}}$, resp.

$\lambda \rightarrow \infty$ results in T_{prod} and \perp_{sum} , resp.

$\lambda \rightarrow -1$ creates T_{-1} and \perp_{-1} , resp.

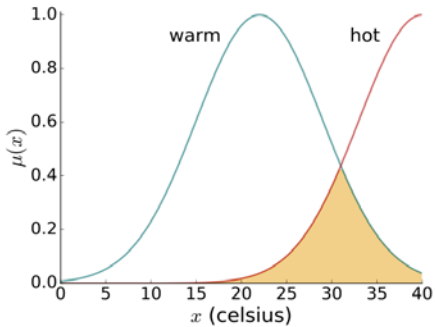
Sugeno-Weber Family II

Additive generators f_λ of T_λ are

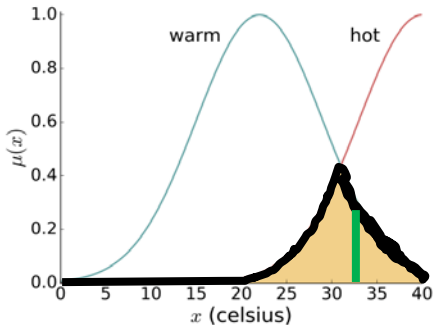
$$f_\lambda(x) = \begin{cases} 1 - x & \text{if } \lambda = 0 \\ 1 - \frac{\log(1+\lambda x)}{\log(1+\lambda)} & \text{otherwise.} \end{cases}$$

$\{T_\lambda\}_{\lambda > -1}$ are increasing functions of parameter λ .

Additive generators of \perp_λ are $g_\lambda(x) = 1 - f_\lambda(x)$.



warm and hot ?



Zadeh' Intersection

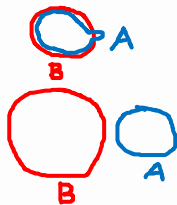
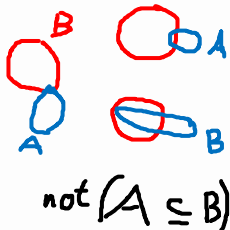
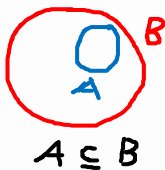
a and $b = \min(a,b)$, for all membership degrees a,b

$(\mu_{\text{warm}} \cap \mu_{\text{hot}})(x) = \min(\mu_{\text{warm}}(x), \mu_{\text{hot}}(x))$, for all real numbers x

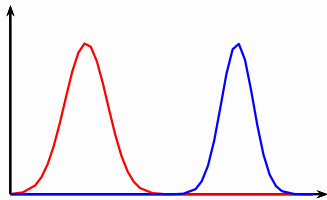
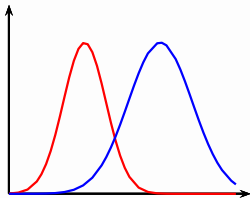
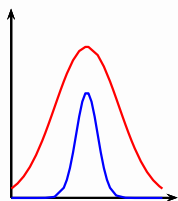
Fuzzy Sets Inclusion

Subset Property

For Classical Sets $x \in A \Rightarrow x \in B$,



For Fuzzy Sets : $x \in \mu \Rightarrow x \in \mu'$



Fuzzy Set Implication

How to model

if speed is **fast** then distance is **high**

A straightforward solution with a multivalued logic

- Define fuzzy sets for **fast** and **high**
- Determine for all speed values x and all distance values y the membership degrees (i.e. its truth value)
- Calculate for each pair x and y the truth value of the implication

$$\mu_{\text{fast}}(x) \Rightarrow \mu_{\text{high}}(y)$$

Definition of a Multivalued Implication

1. One way of defining I is to use the property that in classical logic the propositions $a \Rightarrow b$ and $\neg a \vee b$ have the same truth values for all truth assignments to a and b .

If we model the disjunction and negation as t -conorm and fuzzy complement, resp., then for all $a, b \in [0, 1]$ the following definition of a fuzzy implication seems reasonable:

$$I(a, b) = \perp(\sim a, b).$$

2. Another way is to use the concept of a residuum in classical logic: $a \Rightarrow b$ and $\max\{x \in \{0, 1\} \mid a \wedge x \leq b\}$ have the same truth values for all truth assignments for a , and b . If in a generalized logic the conjunction is modelled by a t -norm, then a reasonable generalization could be:

$$I(a, b) = \sup\{x \in [0, 1] \mid T(a, x) \leq b\}.$$

Definition of a Multivalued Implication

3. Another proposal is to use the fact that, in classical logic, the propositions $a \Rightarrow b$ and $\neg a \vee (a \wedge b)$ have the same truth for all truth assignments.

A possible extension to many valued logics is therefore

$$I(a, b) = \perp(\sim a, T(a, b)),$$

where (T, \perp, \sim) should be a *De Morgan triplet*.

So again, the classical definition of an implication is unique, whereas there is a „zoo“ of fuzzy implications.

Typical question for applications: **What to use when and why?**

S-Implications

Implications based on $I(a, b) = \perp(\sim a, b)$ are called **S-implications**.

Symbol S is often used to denote t -conorms.

Four well-known S -implications are based on $\sim a = 1 - a$:

Name	$I(a, b)$	$\perp(a, b)$
Kleene-Dienes	$I_{\max}(a, b) = \max(1 - a, b)$	$\max(a, b)$
Reichenbach	$I_{\text{sum}}(a, b) = 1 - a + ab$	$a + b - ab$
Łukasiewicz	$I_{\perp}(a, b) = \min(1, 1 - a + b)$	$\min(1, a + b)$
largest	$I_{-1}(a, b) = \begin{cases} b, & \text{if } a = 1 \\ 1 - a, & \text{if } b = 0 \\ 1, & \text{otherwise} \end{cases}$	$\begin{cases} b, & \text{if } a = 0 \\ a, & \text{if } b = 0 \\ 1, & \text{otherwise} \end{cases}$

R -Implications

$I(a, b) = \sup \{x \in [0, 1] \mid T(a, x) \leq b\}$ leads to R -implications.

Symbol R represents close connection to residuated semigroup.

Three well-known R -implications are based on $\sim a = 1 - a$:

- Standard fuzzy intersection leads to **Gödel implication**

$$I_{\min}(a, b) = \sup \{x \mid \min(a, x) \leq b\} = \begin{cases} 1, & \text{if } a \leq b \\ b, & \text{if } a > b. \end{cases}$$

- Product leads to **Goguen implication**

$$I_{\text{prod}}(a, b) = \sup \{x \mid ax \leq b\} = \begin{cases} 1, & \text{if } a \leq b \\ b/a, & \text{if } a > b. \end{cases}$$

- Łukasiewicz t -norm leads to **Łukasiewicz implication**

$$I_{\text{Ł}}(a, b) = \sup \{x \mid \max(0, a + x - 1) \leq b\} = \min(1, 1 - a + b).$$

QL-Implications

Implications based on $I(a, b) = \perp(\sim a, \top(a, b))$ are called **QL-implications** (QL from quantum logic).

Four well-known QL-implications are based on $\sim a = 1 - a$:

- Standard min and max lead to **Zadeh implication**

$$I_Z(a, b) = \max[1 - a, \min(a, b)].$$

- The algebraic product and sum lead to

$$I_p(a, b) = 1 - a + a^2 b.$$

- Using \top_{\perp} and \perp_{\perp} leads to **Kleene-Dienes implication** again.
- Using \top_{-1} and \perp_{-1} leads to

$$I_q(a, b) = \begin{cases} b, & \text{if } a = 1 \\ 1 - a, & \text{if } a \neq 1, b \neq 1 \\ 1, & \text{if } a \neq 1, b = 1. \end{cases}$$

All I come from generalizations of the classical implication.

They collapse to the classical implication when truth values are 0 or 1.

Generalizing classical properties leads to following **propositions** :

- 1) $a \leq b$ implies $I(a, x) \geq I(b, x)$ (*monotonicity in 1st argument*)
- 2) $a \leq b$ implies $I(x, a) \leq I(x, b)$ (*monotonicity in 2nd argument*)
- 3) $I(0, a) = 1$ (*dominance of falsity*)
- 4) $I(1, b) = b$ (*neutrality of truth*)
- 5) $I(a, a) = 1$ (*identity*)
- 6) $I(a, I(b, c)) = I(b, I(a, c))$ (*exchange property*)
- 7) $I(a, b) = 1$ if and only if $a \leq b$ (*boundary condition*)
- 8) $I(a, b) = I(\sim b, \sim a)$ for fuzzy complement \sim (*contraposition*)
- 9) I is a continuous function (*continuity*)

Generator Function

I that satisfy all listed axioms are characterized by this theorem:

Theorem

A function $I : [0, 1]^2 \rightarrow [0, 1]$ satisfies Axioms 1–9 of fuzzy implications for a particular fuzzy complement \sim if and only if there exists a strict increasing continuous function $f : [0, 1] \rightarrow [0, \infty)$ such that $f(0) = 0$,

$$I(a, b) = f^{(-1)}(f(1) - f(a) + f(b))$$

for all $a, b \in [0, 1]$, and

$$\sim a = f^{-1}(f(1) - f(a))$$

for all $a \in [0, 1]$.

Example

Consider $f_\lambda(a) = \ln(1 + \lambda a)$ with $a \in [0, 1]$ and $\lambda > 0$.

Its pseudo-inverse is

$$f_\lambda^{(-1)}(a) = \begin{cases} \frac{e^a - 1}{\lambda}, & \text{if } 0 \leq a \leq \ln(1 + \lambda) \\ 1, & \text{otherwise.} \end{cases}$$

The fuzzy **negation** generated by f_λ for all $a \in [0, 1]$ is

$$n_\lambda(a) = \frac{1 - a}{1 + \lambda a}.$$

The resulting fuzzy implication for all $a, b \in [0, 1]$ is thus

$$I_\lambda(a, b) = \min \left(1, \frac{1 - a + b + \lambda b}{1 + \lambda a} \right).$$

If $\lambda \in (-1, 0)$, then I_λ is called **pseudo-Lukasiewicz implication**.