

# Computational Intelligence

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Prof. Dr. Günter Rudolph

Lehrstuhl für Algorithm Engineering (LS 11)

Fakultät für Informatik

TU Dortmund

## Plan for Today

## Lecture 07

- Fuzzy relations
- Fuzzy logic
  - Linguistic variables and terms
  - Inference from fuzzy statements

## Fuzzy Relations

## Lecture 07

relations with conventional sets  $\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n$ :

$$R(\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n) \subseteq \mathcal{X}_1 \times \mathcal{X}_2 \times \dots \times \mathcal{X}_n$$

notice that cartesian product is a **set**!

⇒ all set operations remain valid!

crisp membership function (of  $x$  to relation  $R$ )

$$R(x_1, x_2, \dots, x_n) = \begin{cases} 1 & \text{if } (x_1, x_2, \dots, x_n) \in R \\ 0 & \text{otherwise} \end{cases}$$

## Fuzzy Relations

## Lecture 07

### Definition

**Fuzzy relation** = fuzzy set over crisp cartesian product  $\mathcal{X}_1 \times \mathcal{X}_2 \times \dots \times \mathcal{X}_n$  ■

→ each tuple  $(x_1, \dots, x_n)$  has a degree of membership to relation

→ degree of membership expresses  
*strength of relationship* between elements of tuple

appropriate representation: n-dimensional membership matrix

**example:** Let  $X = \{ \text{New York}, \text{Paris} \}$  and  $Y = \{ \text{Beijing}, \text{New York}, \text{Dortmund} \}$ .

relation  $R$  = “very far away”

membership matrix →

relation $R$	New York	Paris
Beijing	1.0	0.9
New York	0.0	0.7
Dortmund	0.6	0.3

**Definition**

Let  $R(X, Y)$  be a fuzzy relation with membership matrix  $R$ . The **inverse fuzzy relation** to  $R(X, Y)$ , denoted  $R^{-1}(Y, X)$ , is a relation on  $Y \times X$  with membership matrix  $R^{-1} = R^t$ . ■

**Remark:**  $R^t$  is the transpose of membership matrix  $R$ .

Evidently:  $(R^{-1})^{-1} = R$  since  $(R^t)^t = R$

**Definition**

Let  $P(X, Y)$  and  $Q(Y, Z)$  be fuzzy relations. The operation  $\circ$  on two relations, denoted  $P(X, Y) \circ Q(Y, Z)$ , is termed **max-min-composition** iff

$$R(x, z) = (P \circ Q)(x, z) = \max_{y \in Y} \min \{ P(x, y), Q(y, z) \}. \quad \blacksquare$$

further methods for realizing compositions of relations:

**max-prod composition**

$$(P \odot Q)(x, z) = \max_{y \in Y} \{ P(x, y) \cdot Q(y, z) \}$$

**generalization: sup-t composition**

$$(P \circ Q)(x, z) = \sup_{y \in Y} \{ t(P(x, y), Q(y, z)) \}, \text{ where } t(\cdot, \cdot) \text{ is a t-norm}$$

e.g.:  $t(a, b) = \min\{a, b\} \Rightarrow$  max-min-composition  
 $t(a, b) = a \cdot b \Rightarrow$  max-prod-composition

**Theorem**

- a) max-min composition is associative.
- b) max-min composition is not commutative.
- c)  $(P(X, Y) \circ Q(Y, Z))^{-1} = Q^{-1}(Z, Y) \circ P^{-1}(Y, X)$ .

membership matrix of max-min composition  
determinable via "fuzzy matrix multiplication":  $R = P \circ Q$

fuzzy matrix multiplication  $r_{ij} = \max_k \min\{p_{ik}, q_{kj}\}$

crisp matrix multiplication  $r_{ij} = \sum_k p_{ik} \cdot q_{kj}$

**Binary fuzzy relations on  $X \times X$  : properties**

- **reflexive**  $\Leftrightarrow \forall x \in X: R(x, x) = 1$
- **irreflexive**  $\Leftrightarrow \exists x \in X: R(x, x) < 1$
- **antireflexive**  $\Leftrightarrow \forall x \in X: R(x, x) < 1$
- **symmetric**  $\Leftrightarrow \forall (x, y) \in X \times X: R(x, y) = R(y, x)$
- **asymmetric**  $\Leftrightarrow \exists (x, y) \in X \times X: R(x, y) \neq R(y, x)$
- **antisymmetric**  $\Leftrightarrow \forall (x, y) \in X \times X: R(x, y) \neq R(y, x)$
- **transitive**  $\Leftrightarrow \forall (x, z) \in X \times X: R(x, z) \geq \max_{y \in Y} \{ R(x, y), R(y, z) \}$
- **intransitive**  $\Leftrightarrow \exists (x, z) \in X \times X: R(x, z) < \max_{y \in Y} \{ R(x, y), R(y, z) \}$
- **antittransitive**  $\Leftrightarrow \forall (x, z) \in X \times X: R(x, z) < \max_{y \in Y} \{ R(x, y), R(y, z) \}$

actually, here: max-min-transitivity ( $\rightarrow$  in general: sup-t-transitivity)

binary fuzzy relation on  $X \times X$ : example

Let  $X$  be the set of all cities in Germany.

Fuzzy relation  $R$  is intended to represent the concept of „very close to“.

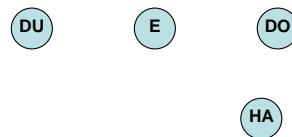
- $R(x,x) = 1$ , since every city is certainly very close to itself.

⇒ **reflexive**

- $R(x,y) = R(y,x)$ : if city  $x$  is very close to city  $y$ , then also vice versa.

⇒ **symmetric**

- $R(\text{Dortmund}, \text{Essen}) = 0.8$



$$R(\text{Essen}, \text{Duisburg}) = 0.7$$

$$R(\text{Dortmund}, \text{Duisburg}) = 0.5$$

$$R(\text{Dortmund}, \text{Hagen}) = 0.9$$

⇒ **intransitive**

**crisp:**

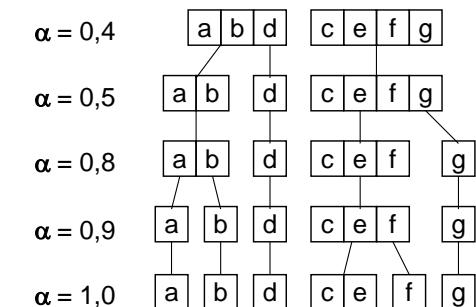
relation  $R$  is equivalence relation  $\Leftrightarrow R$  reflexive, symmetric, transitive

**fuzzy:**

relation  $R$  is similarity relation  $\Leftrightarrow R$  reflexive, symmetric, (max-min-) transitive

Bsp:

	a	b	c	d	e	f	g
a	1,0	0,8	0,0	0,4	0,0	0,0	0,0
b	0,8	1,0	0,0	0,4	0,0	0,0	0,0
c	0,0	0,0	1,0	0,0	1,0	0,9	0,5
d	0,4	0,4	0,0	1,0	0,0	0,0	0,0
e	0,0	0,0	1,0	0,0	1,0	0,9	0,5
f	0,0	0,0	0,9	0,0	0,9	1,0	0,5
g	0,0	0,0	0,5	0,0	0,5	0,5	1,0

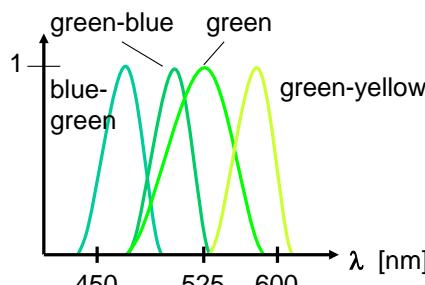
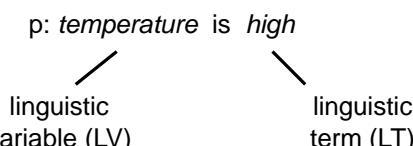
**linguistic variable:**

variable that can attain several values of linguistic / verbal nature

e.g.: **color** can attain values **red, green, blue, yellow, ...**

values (red, green, ...) of linguistic variable are called **linguistic terms**

linguistic terms are associated with fuzzy sets

fuzzy proposition

- LV may be associated with several LT : *high, medium, low, ...*
- *high, medium, low* temperature are fuzzy sets over numerical scale of crisp temperatures
- trueness of fuzzy proposition „temperature is high“ for a given **concrete crisp** temperature value  $v$  is interpreted as equal to the degree of membership  $high(v)$  of the fuzzy set *high*

fuzzy proposition

$p: V \text{ is } F$

linguistic variable (LV)      linguistic term (LT)

actually:

$p: V \text{ is } F(v)$

and

$T(p) = F(v)$  for a concrete crisp value  $v$

trueness( $p$ )

establishes connection between *degree of membership* of a fuzzy set and the *degree of trueness* of a fuzzy proposition

fuzzy proposition

$p: \text{IF } X \text{ is } A, \text{THEN } Y \text{ is } B$

LV    LT    LV    LT

How can we determine / express degree of trueness  $T(p)$  ?

- For crisp, given values  $x, y$  we know  $A(x)$  and  $B(y)$
- $A(x)$  and  $B(y)$  must be processed to single value via relation  $R$
- $R(x, y) = \text{function}(A(x), B(y))$  is fuzzy set over  $X \times Y$
- as before: interpret  $T(p)$  as degree of membership  $R(x, y)$

fuzzy proposition

$p: \text{IF heating is hot, THEN energy consumption is high}$

LV    LT    LV    LT

expresses relation between

- a) temperature of heating and
- b) quantity of energy consumption

$p: (\text{heating}, \text{energy consumption}) \in R$

fuzzy proposition

$p: \text{IF } X \text{ is } A, \text{THEN } Y \text{ is } B$

$A$  is fuzzy set over  $X$

$B$  is fuzzy set over  $Y$

$R$  is fuzzy set over  $X \times Y$

$\forall (x,y) \in X \times Y: R(x, y) = \text{Imp}(A(x), B(y))$

What is  $\text{Imp}(\cdot, \cdot)$  ?

$\Rightarrow$  „appropriate“ fuzzy implication  $[0,1] \times [0,1] \rightarrow [0,1]$

**assumption:** we know an „appropriate“  $\text{Imp}(a,b)$ .

How can we determine the degree of trueness  $T(p)$  ?

**example:**

let  $\text{Imp}(a, b) = \min\{ 1, 1 - a + b \}$  and consider fuzzy sets

A:	$x_1$	$x_2$	$x_3$
	0.1	0.8	1.0

B:	$y_1$	$y_2$
	0.5	1.0

$\Rightarrow$	R	$x_1$	$x_2$	$x_3$
	$y_1$	1.0	0.7	0.5
	$y_2$	1.0	1.0	1.0

z.B.

$$R(x_2, y_1) = \text{Imp}(A(x_2), B(y_1)) = \text{Imp}(0.8, 0.5) = \min\{1.0, 0.7\} = 0.7$$

and  $T(p)$  for  $(x_2, y_1)$  is  $R(x_2, y_1) = 0.7$  ■

**inference from fuzzy statements**

- let  $\forall x, y: y = f(x)$ .

IF  $X = x$  THEN  $Y = f(x)$

- IF  $X \in A$  THEN  $Y \in B = \{ y \in Y: y = f(x), x \in A \}$

**inference from fuzzy statements**

- Let relationship between  $x$  and  $y$  be a relation  $R$  on  $X \times Y$

IF  $X = x$  THEN  $Y \in B = \{ y \in Y: (x, y) \in R \}$

- IF  $X \in A$  THEN  $Y \in B = \{ y \in Y: (x, y) \in R, x \in A \}$

**inference from fuzzy statements**

IF  $X \in A$  THEN  $Y \in B = \{ y \in Y: (x, y) \in R, x \in A \}$

also expressible via characteristic functions of sets  $A, B, R$ :

$$\forall y \in Y: B(y) = \sup_{x \in X} \min \{ A(x), R(x, y) \}$$

**Now:**  $A'$ ,  $B'$  fuzzy set over  $X$  resp.  $Y$

Assume  $R$  and  $A'$  are given:

$$\forall y \in Y: B'(y) = \sup_{x \in X} \min \{ A'(x), R(x, y) \}$$

**composition rule of inference (in matrix form):  $B' = A' \circ R$**

**inference from fuzzy statements**

- conventional:  
modus ponens

$$\begin{array}{c} a \Rightarrow b \\ a \\ \hline b \end{array}$$

- fuzzy:  
generalized modus ponens (GMP)

$$\begin{array}{c} \text{IF } X \text{ is } A, \text{ THEN } Y \text{ is } B \\ X \text{ is } A' \\ \hline Y \text{ is } B' \end{array}$$

e.g.: IF heating is hot, THEN energy consumption is high  
heating is warm  
energy consumption is normal

**inference from fuzzy statements**

- conventional:  
modus tollens

$$\begin{array}{c} a \Rightarrow b \\ \overline{b} \\ \hline \overline{a} \end{array}$$

- fuzzy:  
generalized modus tollens (GMT)

$$\begin{array}{c} \text{IF } X \text{ is } A, \text{ THEN } Y \text{ is } B \\ Y \text{ is } B' \\ \hline X \text{ is } A' \end{array}$$

e.g.: IF heating is hot, THEN energy consumption is high  
energy consumption is normal  
heating is warm

**example: GMP**

consider

$x_1$	$x_2$	$x_3$
0.5	1.0	0.6

$y_1$	$y_2$
1.0	0.4

with the rule: IF  $X$  is  $A$  THEN  $Y$  is  $B$

given fact

$x_1$	$x_2$	$x_3$
0.6	0.9	0.7

$R$	$x_1$	$x_2$	$x_3$
$y_1$	1.0	1.0	1.0
$y_2$	0.9	0.4	0.8

with  $\text{Imp}(a,b) = \min\{1, 1-a+b\}$

thus:  $A' \circ R = B'$

$$(0.6 \quad 0.9 \quad 0.7) \circ \begin{pmatrix} 1.0 & 0.9 \\ 1.0 & 0.4 \\ 1.0 & 0.8 \end{pmatrix} = (0.9 \quad 0.7)$$

**inference from fuzzy statements**

- conventional:  
modus tollens

$$\begin{array}{c} a \Rightarrow b \\ \overline{b} \\ \hline \overline{a} \end{array}$$

- fuzzy:  
generalized modus tollens (GMT)

$$\begin{array}{c} \text{IF } X \text{ is } A, \text{ THEN } Y \text{ is } B \\ Y \text{ is } B' \\ \hline X \text{ is } A' \end{array}$$

e.g.: IF heating is hot, THEN energy consumption is high  
energy consumption is normal  
heating is warm

**example: GMT**

consider

$x_1$	$x_2$	$x_3$
0.5	1.0	0.6

$y_1$	$y_2$
1.0	0.4

with the rule: IF  $X$  is  $A$  THEN  $Y$  is  $B$

given fact

$y_1$	$y_2$
0.9	0.7

$R$	$x_1$	$x_2$	$x_3$
$y_1$	1.0	1.0	1.0
$y_2$	0.9	0.4	0.8

with  $\text{Imp}(a,b) = \min\{1, 1-a+b\}$

$$(0.9 \quad 0.7) \circ \begin{pmatrix} 1.0 & 1.0 & 1.0 \\ 0.9 & 0.4 & 0.8 \end{pmatrix} = (0.9 \quad 0.9 \quad 0.9)$$

**inference from fuzzy statements**

- conventional:  
hypothetic syllogism

$$\begin{array}{c} a \Rightarrow b \\ b \Rightarrow c \\ \hline a \Rightarrow c \end{array}$$

- fuzzy:  
generalized HS

$$\begin{array}{c} \text{IF } X \text{ is A, THEN } Y \text{ is B} \\ \text{IF } Y \text{ is B, THEN } Z \text{ is C} \\ \hline \text{IF } X \text{ is A, THEN } Z \text{ is C} \end{array}$$

e.g.: IF heating is hot, THEN energy consumption is high  
IF energy consumption is high, THEN living is expensive  
IF heating is hot, THEN living is expensive

**example: GHS**

let fuzzy sets  $A(x)$ ,  $B(y)$ ,  $C(z)$  be given

⇒ determine the three relations

$$\begin{array}{l} R_1(x,y) = \text{Imp}(A(x),B(y)) \\ R_2(y,z) = \text{Imp}(B(y),C(z)) \\ R_3(x,z) = \text{Imp}(A(x),C(z)) \end{array}$$

and express them as matrices  $R_1$ ,  $R_2$ ,  $R_3$

**We say:**

GHS is valid if  $R_1 \circ R_2 = R_3$

**So, ... what makes sense for  $\text{Imp}(\cdot, \cdot)$  ?**

$\text{Imp}(a,b)$  ought to express fuzzy version of implication ( $a \Rightarrow b$ )

conventional:  $a \Rightarrow b$  identical to  $\bar{a} \vee b$

But how can we calculate with fuzzy “boolean” expressions?

**request:** must be compatible to crisp version (and more) for  $a,b \in \{0, 1\}$

a	b	$a \wedge b$	$t(a,b)$
0	0	0	0
0	1	0	0
1	0	0	0
1	1	1	1

a	b	$a \vee b$	$s(a,b)$
0	0	0	0
0	1	1	1
1	0	1	1
1	1	1	1

a	$\bar{a}$	$c(a)$
0	1	1
1	0	0

**example: GHS**

let fuzzy sets  $A(x)$ ,  $B(y)$ ,  $C(z)$  be given

⇒ determine the three relations

$$\begin{array}{l} R_1(x,y) = \text{Imp}(A(x),B(y)) \\ R_2(y,z) = \text{Imp}(B(y),C(z)) \\ R_3(x,z) = \text{Imp}(A(x),C(z)) \end{array}$$

and express them as matrices  $R_1$ ,  $R_2$ ,  $R_3$

**We say:**

GHS is valid if  $R_1 \circ R_2 = R_3$

**Fuzzy Logic****Lecture 07**

**So, ... what makes sense for  $\text{Imp}(\cdot, \cdot)$  ?**

**1st approach: S implications**

conventional:  $a \Rightarrow b$  identical to  $\bar{a} \vee b$

fuzzy:  $\text{Imp}(a, b) = s(c(a), b)$

**2nd approach: R implications**

conventional:  $a \Rightarrow b$  identical to  $\max\{x \in \{0,1\} : a \wedge x \leq b\}$

fuzzy:  $\text{Imp}(a, b) = \max\{x \in [0,1] : t(a, x) \leq b\}$

**3rd approach: QL implications**

conventional:  $a \Rightarrow b$  identical to  $\bar{a} \vee b \equiv \bar{a} \vee (a \wedge b)$  law of absorption

fuzzy:  $\text{Imp}(a, b) = s(c(a), t(a, b))$  (dual triple ?)

**example: S implication**

$$\text{Imp}(a, b) = s(c_s(a), b) \quad (c_s : \text{std. complement})$$

## 1. Kleene-Dienes implication

$$s(a, b) = \max\{ a, b \} \quad (\text{standard}) \quad \text{Imp}(a, b) = \max\{ 1-a, b \}$$

## 2. Reichenbach implication

$$s(a, b) = a + b - ab \quad (\text{algebraic sum}) \quad \text{Imp}(a, b) = 1 - a + ab$$

## 3. Łukasiewicz implication

$$s(a, b) = \min\{ 1, a + b \} \quad (\text{bounded sum}) \quad \text{Imp}(a, b) = \min\{ 1, 1 - a + b \}$$

**example: R implicationen**

$$\text{Imp}(a, b) = \max\{ x \in [0,1] : t(a, x) \leq b \}$$

## 1. Gödel implication

$$t(a, b) = \min\{ a, b \} \quad (\text{std.}) \quad \text{Imp}(a, b) = \begin{cases} 1 & , \text{ if } a \leq b \\ b & , \text{ else} \end{cases}$$

## 2. Goguen implication

$$t(a, b) = ab \quad (\text{algeb. product}) \quad \text{Imp}(a, b) = \begin{cases} 1 & , \text{ if } a \leq b \\ \frac{b}{a} & , \text{ else} \end{cases}$$

## 3. Łukasiewicz implication

$$t(a, b) = \max\{ 0, a + b - 1 \} \quad (\text{bounded diff.}) \quad \text{Imp}(a, b) = \min\{ 1, 1 - a + b \}$$

**example: QL implication**

$$\text{Imp}(a, b) = s(c(a), t(a, b))$$

## 1. Zadeh implication

$$t(a, b) = \min\{ a, b \} \quad (\text{std.}) \quad \text{Imp}(a, b) = \max\{ 1 - a, \min\{ a, b \} \}$$

2. „NN“ implication  $\odot$  (Klir/Yuan 1994)

$$t(a, b) = ab \quad (\text{algebr. prd.}) \quad \text{Imp}(a, b) = 1 - a + a^2b$$

## 3. Kleene-Dienes implication

$$t(a, b) = \max\{ 0, a + b - 1 \} \quad (\text{bounded diff.}) \quad \text{Imp}(a, b) = \max\{ 1 - a, b \}$$

**axioms for fuzzy implications**

1.  $a \leq b$  implies  $\text{Imp}(a, x) \geq \text{Imp}(b, x)$  monotone in 1st argument
2.  $a \leq b$  implies  $\text{Imp}(x, a) \leq \text{Imp}(x, b)$  monotone in 2nd argument
3.  $\text{Imp}(0, a) = 1$  dominance of falseness
4.  $\text{Imp}(1, b) = b$  neutrality of trueness
5.  $\text{Imp}(a, a) = 1$  identity
6.  $\text{Imp}(a, \text{Imp}(b, x)) = \text{Imp}(b, \text{Imp}(a, x))$  exchange property
7.  $\text{Imp}(a, b) = 1$  iff  $a \leq b$  boundary condition
8.  $\text{Imp}(a, b) = \text{Imp}(c(b), c(a))$  contraposition
9.  $\text{Imp}(\cdot, \cdot)$  is continuous continuity

## characterization of fuzzy implication

**Theorem:**

$\text{Imp}: [0,1] \times [0,1] \rightarrow [0,1]$  satisfies axioms 1-9 for fuzzy implications  
for a certain fuzzy complement  $c(\cdot)$   $\Leftrightarrow$

$\exists$  strictly monotone increasing, continuous function  $f: [0,1] \rightarrow [0, \infty)$  with

- $f(0) = 0$
- $\forall a, b \in [0,1]: \text{Imp}(a, b) = f^{-1}(f(1) - f(a) + f(b))$
- $\forall a \in [0,1]: c(a) = f^{-1}(f(1) - f(a))$

**Proof:** Smets & Magrez (1987). ■

**examples:** (in tutorial)

## choosing an „appropriate“ fuzzy implication ...

**apt quotation:** (Klir & Yuan 1995, p. 312)

„To select an appropriate fuzzy implication for approximate reasoning under each particular situation is a difficult problem.“

**guideline:**

GMP, GMT, GHS should be compatible with MP, MT, HS  
for fuzzy implication in calculations with relations:

$$B(y) = \sup \{ t(A(x), \text{Imp}(A(x), B(y))) : x \in X \}$$

**example:**

Gödel implication for t-norm = bounded difference