

# Fuzzy Relations, Rules and Inferences

# Fuzzy Relations

# Crisp relations

To understand the fuzzy relations, it is better to discuss first **crisp relation**.

Suppose,  $A$  and  $B$  are two (crisp) sets. Then Cartesian product denoted as  $A \times B$  is a collection of order pairs, such that

$$A \times B = \{(a, b) | a \in A \text{ and } b \in B\}$$

Note :

$$(1) A \times B \neq B \times A$$

$$(2) |A \times B| = |A| \times |B|$$

(3)  $A \times B$  provides a mapping from  $a \in A$  to  $b \in B$ .

The mapping so mentioned is called a **relation**.

# Crisp relations

## Example 1:

Consider the two crisp sets  $A$  and  $B$  as given below.  $A = \{1, 2, 3, 4\}$   
 $B = \{3, 5, 7\}$ .

Then,  $A \times B = \{(1, 3), (1, 5), (1, 7), (2, 3), (2, 5), (2, 7), (3, 3), (3, 5), (3, 7), (4, 3), (4, 5), (4, 7)\}$

Let us define a relation  $R$  as  $R = \{(a, b) | b = a + 1, (a, b) \in A \times B\}$

Then,  $R = \{(2, 3), (4, 5)\}$  in this case.

We can represent the relation  $R$  in a matrix form as follows.

$$R = \begin{matrix} & \begin{matrix} 3 & 5 & 7 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \end{matrix}$$

# Operations on crisp relations

Suppose,  $R(x, y)$  and  $S(x, y)$  are the two relations define over two crisp sets  $x \in A$  and  $y \in B$

**Union:**

$$R(x, y) \cup S(x, y) = \max(R(x, y), S(x, y));$$

**Intersection:**

$$R(x, y) \cap S(x, y) = \min(R(x, y), S(x, y));$$

**Complement:**

$$\overline{R(x, y)} = 1 - R(x, y)$$

# Example: Operations on crisp relations

Example:

Suppose,  $R(x, y)$  and  $S(x, y)$  are the two relations define over two crisp sets  $x \in A$  and  $y \in B$

$$R = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \text{ and } S = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix};$$

1  $R \cup S$

2  $R \cap S$

3  $R$

# Composition of two crisp relations

Given  $R$  is a relation on  $X, Y$  and  $S$  is another relation on  $Y, Z$ .  
Then  $R \circ S$  is called a composition of relation on  $X$  and  $Z$  which is defined as follows.

$$R \circ S = \{(x, z) | (x, y) \in R \text{ and } (y, z) \in S \text{ and } \forall y \in Y\}$$

## Max-Min Composition

Given the two relation matrices  $R$  and  $S$ , the **max-min composition** is defined as  $T = R \circ S$  ;

$$T(x, z) = \max\{\min\{R(x, y), S(y, z) \text{ and } \forall y \in Y\}\}$$

# Composition: Composition

## Example:

Given

$$X = \{1, 3, 5\}; Y = \{1, 3, 5\}; R = \{(x, y) | y = x + 2\}; S = \{(x, y) | x < y\}$$

Here,  $R$  and  $S$  is on  $X \times Y$ .

Thus, we have

$$R = \{(1, 3), (3, 5)\}$$

$$S = \{(1, 3), (1, 5), (3, 5)\}$$

$$R = \begin{matrix} & \begin{matrix} 1 & 3 & 5 \end{matrix} \\ \begin{matrix} 1 \\ 3 \\ 5 \end{matrix} & \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix} \end{matrix} \text{ and } S =$$

$$\begin{matrix} & \begin{matrix} 1 & 3 & 5 \end{matrix} \\ \begin{matrix} 1 \\ 3 \\ 5 \end{matrix} & \begin{bmatrix} 0 & 1 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix} \end{matrix}$$

Using max-min composition  $R \circ S =$

$$\begin{matrix} & \begin{matrix} 1 & 3 & 5 \end{matrix} \\ \begin{matrix} 1 \\ 3 \\ 5 \end{matrix} & \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \end{matrix}$$



# Fuzzy relations

- Fuzzy relation is a fuzzy set defined on the Cartesian product of crisp set  $X_1, X_2, \dots, X_n$
- Here,  $n$ -tuples  $(x_1, x_2, \dots, x_n)$  may have varying degree of memberships within the relationship.
- The membership values indicate the strength of the relation between the tuples.

Example:

$X = \{ \text{typhoid, viral, cold} \}$  and  $Y = \{ \text{running nose, high temp, shivering} \}$

The fuzzy relation  $R$  is defined as

	<i>runningnose</i>	<i>hightemperature</i>	<i>shivering</i>
<i>typhoid</i>	0.1	0.9	0.8
<i>viral</i>	0.2	0.9	0.7
<i>cold</i>	0.9	0.4	0.6

# Fuzzy Cartesian product

Suppose

$A$  is a fuzzy set on the universe of discourse  $X$  with  $\mu_A(x) | x \in X$

$B$  is a fuzzy set on the universe of discourse  $Y$  with  $\mu_B(y) | y \in Y$

Then  $R = A \times B \subset X \times Y$  ; where  $R$  has its membership function given by  $\mu_R(x, y) = \mu_{A \times B}(x, y) = \min\{\mu_A(x), \mu_B(y)\}$

Example :

$A = \{(a_1, 0.2), (a_2, 0.7), (a_3, 0.4)\}$  and  $B = \{(b_1, 0.5), (b_2, 0.6)\}$

$$R = A \times B = \begin{array}{cc} & \begin{matrix} b_1 & b_2 \end{matrix} \\ \begin{matrix} a_1 \\ a_2 \\ a_3 \end{matrix} & \left[ \begin{array}{cc} 0.2 & 0.2 \\ 0.5 & 0.6 \\ 0.4 & 0.4 \end{array} \right] \end{array}$$

# Operations on Fuzzy relations

Let  $R$  and  $S$  be two fuzzy relations on  $A \times B$ .

**Union:**

$$\mu_{R \cup S}(a, b) = \max\{\mu_R(a, b), \mu_S(a, b)\}$$

**Intersection:**

$$\mu_{R \cap S}(a, b) = \min\{\mu_R(a, b), \mu_S(a, b)\}$$

**Complement:**

$$\mu_{\bar{R}}(a, b) = 1 - \mu_R(a, b)$$

**Composition**

$$T = R \circ S$$
$$\mu_{R \circ S} = \max_{y \in Y} \{\min(\mu_R(x, y), \mu_S(y, z))\}$$

# Operations on Fuzzy relations: Examples

Example:

$$X = (x_1, x_2, x_3); Y = (y_1, y_2); Z = (z_1, z_2, z_3);$$

$$R = \begin{matrix} & \begin{matrix} y_1 & y_2 \end{matrix} \\ \begin{matrix} x_1 \\ x_2 \\ x_3 \end{matrix} & \begin{bmatrix} 0.5 & 0.1 \\ 0.2 & 0.9 \\ 0.8 & 0.6 \end{bmatrix} \end{matrix}$$

$$S = \begin{matrix} & \begin{matrix} z_1 & z_2 & z_3 \end{matrix} \\ \begin{matrix} y_1 \\ y_2 \end{matrix} & \begin{bmatrix} 0.6 & 0.4 & 0.7 \\ 0.5 & 0.8 & 0.9 \end{bmatrix} \end{matrix}$$

$$R \circ S = \begin{matrix} & \begin{matrix} z_1 & z_2 & z_3 \end{matrix} \\ \begin{matrix} x_1 \\ x_2 \\ x_3 \end{matrix} & \begin{bmatrix} 0.5 & 0.4 & 0.5 \\ 0.5 & 0.8 & 0.9 \\ 0.6 & 0.6 & 0.7 \end{bmatrix} \end{matrix}$$

$$\begin{aligned} \mu_{R \circ S}(x_1, y_1) &= \max\{\min(x_1, y_1), \min(y_1, z_1), \min(x_1, y_2), \min(y_2, z_1)\} \\ &= \max\{\min(0.5, 0.6), \min(0.1, 0.5)\} = \max\{0.5, 0.1\} = 0.5 \text{ and so on.} \end{aligned}$$

## Fuzzy relation : An example

Consider the following two sets  $P$  and  $D$ , which represent a set of paddy plants and a set of plant diseases. More precisely

$P = \{P_1, P_2, P_3, P_4\}$  a set of four varieties of paddy plants

$D = \{D_1, D_2, D_3, D_4\}$  of the four various diseases affecting the plants

In addition to these, also consider another set  $S = \{S_1, S_2, S_3, S_4\}$  be the common symptoms of the diseases.

Let,  $R$  be a relation on  $P \times D$ , representing which plant is susceptible to which diseases, then  $R$  can be stated as

$$R = \begin{array}{c} P_1 \\ P_2 \\ P_3 \\ P_4 \end{array} \begin{bmatrix} D_1 & D_2 & D_3 & D_4 \\ 0.6 & 0.6 & 0.9 & 0.8 \\ 0.1 & 0.2 & 0.9 & 0.8 \\ 0.9 & 0.3 & 0.4 & 0.8 \\ 0.9 & 0.8 & 0.4 & 0.2 \end{bmatrix}$$

## Fuzzy relation : An example

Also, consider  $T$  be the another relation on  $D \times S$ , which is given by

$$S = \begin{array}{c} D_1 \\ D_2 \\ D_3 \\ D_4 \end{array} \begin{array}{cccc} S_1 & S_2 & S_3 & S_4 \\ \left[ \begin{array}{cccc} 0.1 & 0.2 & 0.7 & 0.9 \\ 1.0 & 1.0 & 0.4 & 0.6 \\ 0.0 & 0.0 & 0.5 & 0.9 \\ 0.9 & 1.0 & 0.8 & 0.2 \end{array} \right] \end{array}$$

Obtain the association of plants with the different symptoms of the disease using **max-min composition**.

$$R \circ S = \begin{array}{c} P_1 \\ P_2 \\ P_3 \\ P_4 \end{array} \begin{array}{cccc} S_1 & S_2 & S_3 & S_4 \\ \left[ \begin{array}{cccc} 0.8 & 0.8 & 0.8 & 0.9 \\ 0.8 & 0.8 & 0.8 & 0.9 \\ 0.8 & 0.8 & 0.8 & 0.9 \\ 0.8 & 0.8 & 0.7 & 0.9 \end{array} \right] \end{array}$$

# Fuzzy Propositions

# Two-valued logic vs. Multi-valued logic

- The basic assumption upon which crisp logic is based - that every proposition is either TRUE or FALSE.
- The classical two-valued logic can be extended to multi-valued logic.
- As an example, three valued logic to denote true(1), false(0) and indeterminacy ( $\frac{1}{2}$ ).



# Two-valued logic vs. Multi-valued logic

Different operations with three-valued logic can be extended as shown in the following truth table:

<b>a</b>	<b>b</b>	$\wedge$	$\vee$	$\neg a$	$\implies$	$=$
0	0	0	0	1	1	1
0	$\frac{1}{2}$	0	$\frac{1}{2}$	1	1	$\frac{1}{2}$
0	1	0	1	1	1	0
$\frac{1}{2}$	0	0	$\frac{1}{2}$	$\frac{1}{2}$	$\frac{1}{2}$	$\frac{1}{2}$
$\frac{1}{2}$	$\frac{1}{2}$	$\frac{1}{2}$	$\frac{1}{2}$	$\frac{1}{2}$	$\frac{1}{2}$	1
$\frac{1}{2}$	1	$\frac{1}{2}$	1	$\frac{1}{2}$	1	$\frac{1}{2}$
1	0	0	1	1	0	0
1	$\frac{1}{2}$	$\frac{1}{2}$	1	1	$\frac{1}{2}$	$\frac{1}{2}$
1	1	1	1	1	1	1

Fuzzy connectives used in the above table are:

AND ( $\wedge$ ), OR ( $\vee$ ), NOT ( $\neg$ ), IMPLICATION ( $\implies$ ) and EQUAL ( $=$ ).

# Three-valued logic

Fuzzy connectives defined for such a three-valued logic better can be stated as follows:

Symbol	Connective	Usage	Definition
$\neg$	NOT	$\neg P$	$1 - T(P)$
$\vee$	OR	$P \vee Q$	$\max\{T(P), T(Q)\}$
$\wedge$	AND	$P \wedge Q$	$\min\{T(P), T(Q)\}$
$\implies$	IMPLICATION	$(P \implies Q)$ or $(\neg P \vee Q)$	$\max\{(1 - T(P)), T(Q)\}$
$=$	EQUALITY	$(P = Q)$ or $[(P \implies Q) \wedge (Q \implies P)]$	$1 -  T(P) - T(Q) $

# Fuzzy proposition

## Example 1:

P : Ram is honest

- 1  $T(P) = 0.0$  : Absolutely false
- 2  $T(P) = 0.2$  : Partially false
- 3  $T(P) = 0.4$  : May be false or not false
- 4  $T(P) = 0.6$  : May be true or not true
- 5  $T(P) = 0.8$  : Partially true
- 6  $T(P) = 1.0$  : Absolutely true.

## Example 2 :Fuzzy proposition

P : Mary is efficient ;  $T(P) = 0.8$ ;

Q : Ram is efficient ;  $T(Q) = 0.6$

1 **Mary is not efficient.**

$$T(\neg P) = 1 - T(P) = 0.2$$

2 **Mary is efficient and so is Ram.**

$$T(P \wedge Q) = \min\{T(P), T(Q)\} = 0.6$$

3 **Either Mary or Ram is efficient**

$$T(P \vee Q) = \max\{T(P), T(Q)\} = 0.8$$

4 **If Mary is efficient then so is Ram**

$$T(P \Rightarrow Q) = \max\{1 - T(P), T(Q)\} = 0.6$$

# Fuzzy proposition vs. Crisp proposition

- The fundamental difference between crisp (classical) proposition and fuzzy propositions is in the range of their truth values.
- While each classical proposition is required to be either true or false, the truth or falsity of fuzzy proposition is a matter of degree.
- The degree of truth of each fuzzy proposition is expressed by a value in the interval  $[0,1]$  both inclusive.

# Canonical representation of Fuzzy proposition

- Suppose,  $X$  is a universe of discourse of five persons.  
Intelligent of  $x \in X$  is a fuzzy set as defined below.

Intelligent:  $\{(x_1, 0.3), (x_2, 0.4), (x_3, 0.1), (x_4, 0.6), (x_5, 0.9)\}$

- We define a fuzzy proposition as follows:

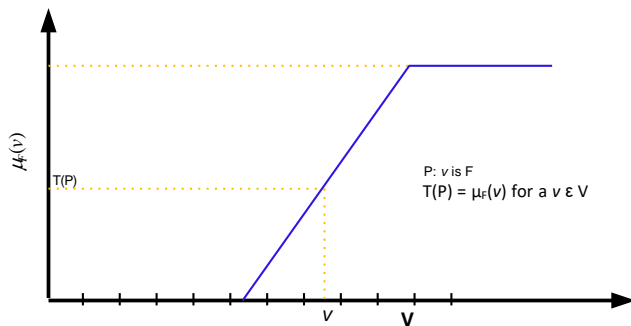
$P : x$  is intelligent

- The canonical form of fuzzy proposition of this type,  $P$  is expressed by the sentence  $P : v$  is  $F$ .
- Predicate in terms of fuzzy set.

$P : \mathbf{v}$  is  $\mathbf{F}$  ; where  $v$  is an element that takes values  $v$  from some universal set  $V$  and  $F$  is a fuzzy set on  $V$  that represents a fuzzy predicate.

- In other words, given, a particular element  $v$ , this element belongs to  $F$  with membership grade  $\mu_F(v)$ .

# Graphical interpretation of fuzzy proposition



- For a given value  $v$  of variable  $V$  in proposition  $P$ ,  $T(P)$  denotes the degree of truth of proposition  $P$ .

# Fuzzy Implications



# Fuzzy rule

- A fuzzy implication (also known as fuzzy If-Then rule, fuzzy rule, or fuzzy conditional statement) assumes the form :

**If  $x$  is  $A$  then  $y$  is  $B$**

where,  $A$  and  $B$  are two linguistic variables defined by fuzzy sets  $A$  and  $B$  on the universe of discourses  $X$  and  $Y$ , respectively.

- Often,  $x$  **is**  $A$  is called the **antecedent** or premise, while  $y$  **is**  $B$  is called the **consequence** or conclusion.

# Fuzzy implication : Example 1

- If pressure is High then temperature is Low
- If mango is Yellow then mango is Sweet else mango is Sour
- If road is Good then driving is Smooth else traffic is High
- The fuzzy implication is denoted as  $R : A \rightarrow B$
- In essence, it represents a binary fuzzy relation  $R$  on the (Cartesian) product of  $A \times B$

## Fuzzy implication : Example 2

- Suppose,  $P$  and  $T$  are two universes of discourses representing pressure and temperature, respectively as follows.
- $P = \{1, 2, 3, 4\}$  and  $T = \{10, 15, 20, 25, 30, 35, 40, 45, 50\}$
- Let the linguistic variable **High temperature** and **Low pressure** are given as
- $T_{HIGH} = \{(20, 0.2), (25, 0.4), (30, 0.6), (35, 0.6), (40, 0.7), (45, 0.8), (50, 0.8)\}$
- $P_{LOW} = (1, 0.8), (2, 0.8), (3, 0.6), (4, 0.4)$

## Fuzzy implications : Example 2

- Then the fuzzy implication **If temperature is High then pressure is Low** can be defined as

$$R : T_{HIGH} \rightarrow P_{LOW}$$

where,  $R =$

	1	2	3	4
20	0.2	0.2	0.2	0.2
25	0.4	0.4	0.4	0.4
30	0.6	0.6	0.6	0.4
35	0.6	0.6	0.6	0.4
40	0.7	0.7	0.6	0.4
45	0.8	0.8	0.6	0.4
50	0.8	0.8	0.6	0.4

**Note :** If temperature is 40 then what about low pressure?

# Zadeh max-min Rule to compute R

With the above mentioned implications, there are a number of fuzzy implication functions that are popularly followed in fuzzy rule-based system.

**Zadeh's max-min rule :**

$$R_{mm} = \bar{A} \cup (A \cap B) = \int_{X \times Y} (1 - \mu_A(x)) \vee (\mu_A(x) \wedge \mu_B(y)) |_{(x,y)}$$

or

$$f_{mm}(a, b) = (1 - a) \vee (a \wedge b)$$

## Example 3: Zadeh's Max-Min rule

If **x is A then y is B** with the implication of Zadeh's max-min rule can be written equivalently as :

$$R_{mm} = (A \times B) \cup (\bar{A} \times Y)$$

**Here**,  $Y$  is the universe of discourse with membership values for all  $y \in Y$  is 1, that is ,  $\mu_Y(y) = 1 \forall y \in Y$ .

**Suppose**  $X = \{a, b, c, d\}$  and  $Y = \{1, 2, 3, 4\}$

and  $A = \{(a, 0.0), (b, 0.8), (c, 0.6), (d, 1.0)\}$

$B = \{(1, 0.2), (2, 1.0), (3, 0.8), (4, 0.0)\}$  are two fuzzy sets.

We are to determine  $R_{mm} = (A \times B) \cup (\bar{A} \times Y)$

### Example 3: Zadeh's min-max rule:

The computation of  $R_{mm} = (A \times B) \cup (\bar{A} \times Y)$  is as follows:

$$A \times B = \begin{array}{c} a \\ b \\ c \\ d \end{array} \begin{array}{cccc} 1 & 2 & 3 & 4 \\ \left[ \begin{array}{cccc} 0 & 0 & 0 & 0 \\ 0.2 & 0.8 & 0.8 & 0 \\ 0.2 & 0.6 & 0.6 & 0 \\ 0.2 & 1.0 & 0.8 & 0 \end{array} \right] \end{array} \text{ and}$$

$$\bar{A} \times Y = \begin{array}{c} a \\ b \\ c \\ d \end{array} \begin{array}{cccc} 1 & 2 & 3 & 4 \\ \left[ \begin{array}{cccc} 1 & 1 & 1 & 1 \\ 0.2 & 0.2 & 0.2 & 0.2 \\ 0.4 & 0.4 & 0.4 & 0.4 \\ 0 & 0 & 0 & 0 \end{array} \right] \end{array}$$

## Example 3: Zadeh's min-max rule:

Therefore,

$$R_{mm} = (A \times B) \cup (\bar{A} \times Y) =$$

	1	2	3	4
$a$	1	1	1	1
$b$	0.2	0.8	0.8	0.2
$c$	0.4	0.6	0.6	0.4
$d$	0.2	1.0	0.8	0



## Example 3 :

$$X = \{a, b, c, d\}$$

$$Y = \{1, 2, 3, 4\}$$

$$\text{Let, } A = \{(a, 0.0), (b, 0.8), (c, 0.6), (d, 1.0)\}$$

$$B = \{(1, 0.2), (2, 1.0), (3, 0.8), (4, 0.0)\}$$

Determine the implication relation :

**If  $x$  is  $A$  then  $y$  is  $B$**

Here,  $A \times B =$

$$\begin{array}{c} a \\ b \\ c \\ d \end{array} \begin{bmatrix} 1 & 2 & 3 & 4 \\ 0 & 0 & 0 & 0 \\ 0.2 & 0.8 & 0.8 & 0 \\ 0.2 & 0.6 & 0.6 & 0 \\ 0.2 & 1.0 & 0.8 & 0 \end{bmatrix}$$

## Example 3 :

$$\text{and } \bar{A} \times Y =$$

	1	2	3	4
$a$	1	1	1	1
$b$	0.2	0.2	0.2	0.2
$c$	0.4	0.4	0.4	0.4
$d$	0	0	0	0

$$R_{mm} = (A \times B) \cup (\bar{A} \times Y) =$$

	1	2	3	4
$a$	1	1	1	1
$b$	0.2	0.8	0.8	0.2
$c$	0.4	0.6	0.6	0.4
$d$	0.2	1.0	0.8	0

This  $R$  represents **If  $x$  is  $A$  then  $y$  is  $B$**

## Example 3 :

IF  $x$  is  $A$  THEN  $y$  is  $B$  ELSE  $y$  is  $C$ .

The relation  $R$  is equivalent to

$$R = (A \times B) \cup (\bar{A} \times C)$$

The membership function of  $R$  is given by

$$\mu_R(x, y) = \max[\min\{\mu_A(x), \mu_B(y)\}, \min\{\mu_{\bar{A}}(x), \mu_C(y)\}]$$

## Example 4:

$$X = \{a, b, c, d\}$$

$$Y = \{1, 2, 3, 4\}$$

$$A = \{(a, 0.0), (b, 0.8), (c, 0.6), (d, 1.0)\}$$

$$B = \{(1, 0.2), (2, 1.0), (3, 0.8), (4, 0.0)\}$$

$$C = \{(1, 0), (2, 0.4), (3, 1.0), (4, 0.8)\}$$

Determine the implication relation :

**If  $x$  is  $A$  then  $y$  is  $B$  else  $y$  is  $C$**

Here,  $A \times B =$

	1	2	3	4
$a$	0	0	0	0
$b$	0.2	0.8	0.8	0
$c$	0.2	0.6	0.6	0
$d$	0.2	1.0	0.8	0

## Example 4:

and  $\bar{A} \times C =$

$$\begin{array}{c} a \\ b \\ c \\ d \end{array} \begin{array}{c} 1 \quad 2 \quad 3 \quad 4 \\ \left[ \begin{array}{cccc} 0 & 0.4 & 1.0 & 0.8 \\ 0 & 0.2 & 0.2 & 0.2 \\ 0 & 0.4 & 0.4 & 0.4 \\ 0 & 0 & 0 & 0 \end{array} \right] \end{array}$$

$$R = \begin{array}{c} a \\ b \\ c \\ d \end{array} \begin{array}{c} 1 \quad 2 \quad 3 \quad 4 \\ \left[ \begin{array}{cccc} 0 & 0.4 & 1.0 & 0.8 \\ 0.2 & 0.8 & 0.8 & 0.2 \\ 0.2 & 0.6 & 0.6 & 0.4 \\ 0.2 & 1.0 & 0.8 & 0 \end{array} \right] \end{array}$$

# Fuzzy Inferences

# Fuzzy inferences

Let's start with propositional logic. We know the following in propositional logic.

1 Modus Ponens :  $P, P \Rightarrow Q, \quad \Leftrightarrow Q$

2 Modus Tollens :  $P \Rightarrow Q, \neg Q \quad \Leftrightarrow, \neg P$

3 Chain rule :  $P \Rightarrow Q, Q \Rightarrow R \quad \Leftrightarrow, P \Rightarrow R$

# Inferring procedures in Fuzzy logic

Two important inferring procedures are used in fuzzy systems :

- **Generalized Modus Ponens (GMP)**

If  $x$  is  $A$  Then  $y$  is  $B$

$x$  is  $A'$

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$y$  is  $B'$

- **Generalized Modus Tollens (GMT)**

If  $x$  is  $A$  Then  $y$  is  $B$

$y$  is  $B'$

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$x$  is  $A'$

$x$  is  $A$



# Fuzzy inferring procedures

- Here,  $A$ ,  $B$ ,  $A'$  and  $B'$  are fuzzy sets.
- To compute the membership function  $A'$  and  $B'$  the max-min composition of fuzzy sets  $B$  and  $A'$ , respectively with  $R(x, y)$  (which is the known implication relation) is to be used.
- Thus,

$$B' = A' \circ R(x, y) \qquad \mu_{B'}(y) = \max[\min(\mu_{A'}(x), \mu_R(x, y))]$$

$$A' = B' \circ R(x, y) \qquad \mu_{A'}(x) = \max[\min(\mu_{B'}(y), \mu_R(x, y))]$$

# Generalized Modus Ponens

## Generalized Modus Ponens (GMP)

$P$  : If  $x$  is  $A$  then  $y$  is  $B$

Let us consider two sets of variables  $x$  and  $y$  be  
 $X = \{x_1, x_2, x_3\}$  and  $Y = \{y_1, y_2\}$ , respectively.

Also, let us consider the following.

$A = \{(x_1, 0.5), (x_2, 1), (x_3, 0.6)\}$

$B = \{(y_1, 1), (y_2, 0.4)\}$

Then, given a fact expressed by the proposition  $x$  is  $A'$ ,  
where  $A' = \{(x_1, 0.6), (x_2, 0.9), (x_3, 0.7)\}$   
derive a conclusion in the form  $y$  is  $B$  (using generalized modus ponens (GMP)).

## Example: Generalized Modus Ponens

If  $x$  is  $A$  Then  $y$  is  $B$

$x$  is  $A'$

---

$y$  is  $B'$

We are to find  $B' = A' \circ R(x, y)$  where  $R(x, y) = \max\{A \times B, \bar{A} \times Y\}$

$$A \times B = \begin{matrix} & y_1 & y_2 \\ \begin{matrix} x_1 \\ x_2 \\ x_3 \end{matrix} & \begin{bmatrix} 0.5 & 0.4 \\ 1 & 0.4 \\ 0.6 & 0.4 \end{bmatrix} \end{matrix} \text{ and } \bar{A} \times Y = \begin{matrix} & y_1 & y_2 \\ \begin{matrix} x_1 \\ x_2 \\ x_3 \end{matrix} & \begin{bmatrix} 0.5 & 0.5 \\ 0 & 0 \\ 0.4 & 0.4 \end{bmatrix} \end{matrix}$$

Note: For  $A \times B$ ,  $\mu_{A \times B}(x, y) = \min(\mu_A(x), \mu_B(y))$

## Example: Generalized Modus Ponens

$$R(x, y) = (A \times B) \cup (\bar{A} \times y) = \begin{matrix} & y_1 & y_2 \\ \begin{matrix} x_1 \\ x_2 \\ x_3 \end{matrix} & \begin{bmatrix} 0.5 & 0.5 \\ 1 & 0.4 \\ 0.6 & 0.4 \end{bmatrix} \end{matrix}$$

**Now,**  $A' = \{(x_1, 0.6), (x_2, 0.9), (x_3, 0.7)\}$

Therefore,  $B' = A' \circ R(x, y) =$

$$[0.6 \quad 0.9 \quad 0.7] \circ \begin{bmatrix} 0.5 & 0.5 \\ 1 & 0.4 \\ 0.6 & 0.4 \end{bmatrix} = [0.9 \quad 0.5]$$

Thus we derive that  $y$  is  $B'$  where  $B' = \{(y_1, 0.9), (y_2, 0.5)\}$

## Example: Generalized Modus Tollens

### Generalized Modus Tollens (GMT)

P: If  $x$  is  $A$  Then  $y$  is  $B$

Q:  $y$  is  $B'$

---

$x$  is  $A'$

## Example: Generalized Modus Tollens

- Let sets of variables  $x$  and  $y$  be  $X = \{x_1, x_2, x_3\}$  and  $y = \{y_1, y_2\}$ , respectively.
- Assume that a proposition **If  $x$  is  $A$  Then  $y$  is  $B$**  given where  $A = \{(x_1, 0.5), (x_2, 1.0), (x_3, 0.6)\}$  and  $B = \{(y_1, 0.6), (y_2, 0.4)\}$
- Assume now that a fact expressed by a proposition  **$y$  is  $B$**  is given where  $B = \{(y_1, 0.9), (y_2, 0.7)\}$ .
- From the above, we are to conclude that  **$x$  is  $A'$** . That is, we are to determine  $A'$

## Example: Generalized Modus Tollens

- We first calculate  $R(x, y) = (A \times B) \cup (\bar{A} \times y)$

$$R(x, y) = \begin{matrix} & \begin{matrix} y_1 & y_2 \end{matrix} \\ \begin{matrix} x_1 \\ x_2 \\ x_3 \end{matrix} & \begin{bmatrix} 0.5 & 0.5 \\ 1 & 0.4 \\ 0.6 & 0.4 \end{bmatrix} \end{matrix}$$

- Next, we calculate  $A' = B' \circ R(x, y)$

$$A' = [0.9 \quad 0.7] \circ \begin{matrix} & \begin{matrix} y_1 & y_2 \end{matrix} \\ \begin{matrix} x_1 \\ x_2 \\ x_3 \end{matrix} & \begin{bmatrix} 0.5 & 0.5 \\ 1 & 0.4 \\ 0.6 & 0.4 \end{bmatrix} \end{matrix} = [0.5 \quad 0.9 \quad 0.6]$$

- Hence, we calculate that  $x$  is  $A'$  where  $A' = [(x_1, 0.5), (x_2, 0.9), (x_3, 0.6)]$

# Practice

Apply the fuzzy GMP rule to deduce **Rotation is quite slow**

Given that :

- If temperature is High then rotation is Slow.
- temperature is Very High

Let,

$X = \{30, 40, 50, 60, 70, 80, 90, 100\}$  be the set of temperatures.

$Y = \{10, 20, 30, 40, 50, 60\}$  be the set of rotations per minute.



# Practice

The fuzzy set High(H), Very High (VH), Slow(S) and Quite Slow (QS) are given below.

$$H = \{(70, 1), (80, 1), (90, 0.3)\}$$

$$VH = \{(90, 0.9), (100, 1)\}$$

$$S = \{(30, 0.8), (40, 1.0), (50, 0.6)\}$$

$$QS = \{(10, 1), (20, 0.8)\}$$

- 1 If temperature is High then the rotation is Slow.

$$R = (H \times S) \cup (\bar{H} \times Y)$$

- 2 temperature is Very High

Thus, to deduce "rotation is Quite Slow", we make use the composition rule  $QS = VH \circ R(x, y)$