

EE496 : COMPUTATIONAL INTELLIGENCE

FS05: FUZZY RULE BASES

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Approximate Reasoning with Fuzzy Rules

General schema

Rule 1: **if** X is M_1 , **then** Y is N_1

Rule 2: **if** X is M_2 , then Y is N_2

...

Rule r : if X is M_r , then Y is N_r

Fact: X is M'

Conclusion: Y is N'

Given r **if-then rules** and fact **X is M'** , we conclude **Y is N'** .

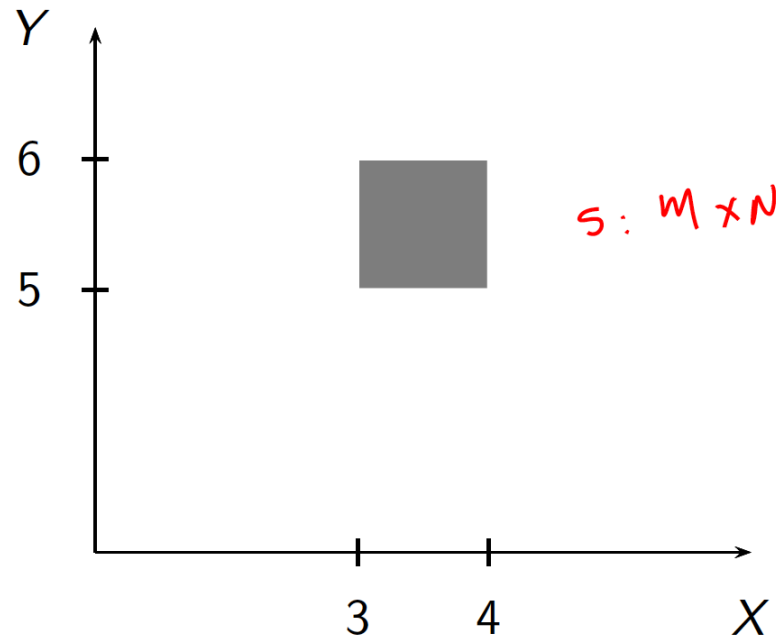
Typically used in **fuzzy controllers**.

Approximate Reasoning

Disjunctive Imprecise Rule

Imprecise rule: if $X = [3, 4]$ then $Y = [5, 6]$.

Interpretation: values coming from $[3, 4] \times [5, 6]$.

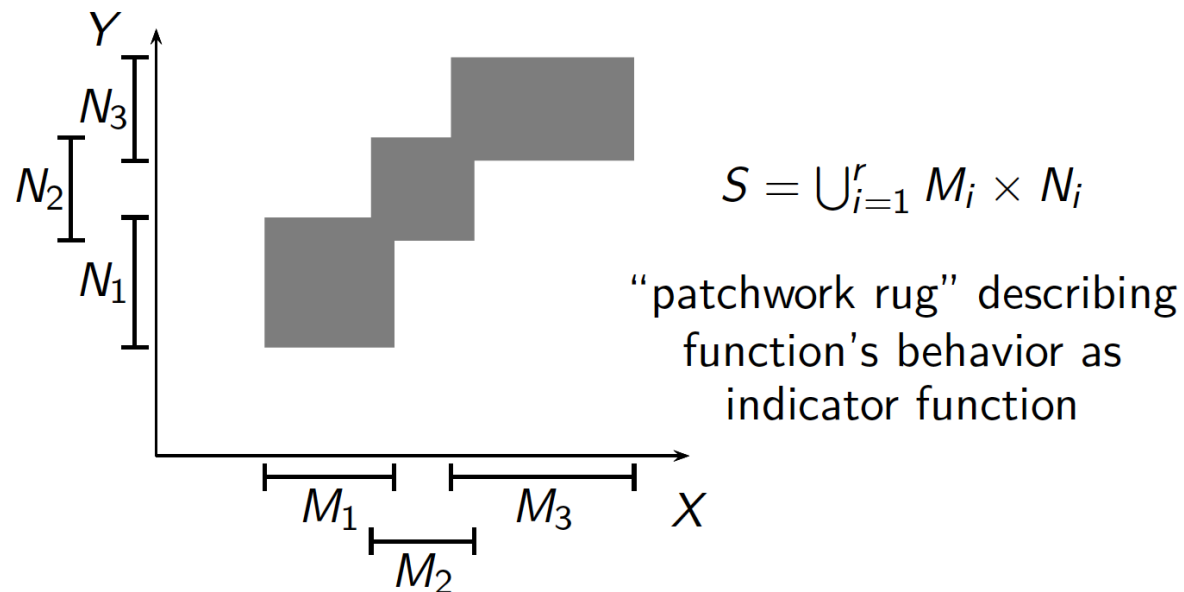


Approximate Reasoning

Disjunctive Imprecise Rules

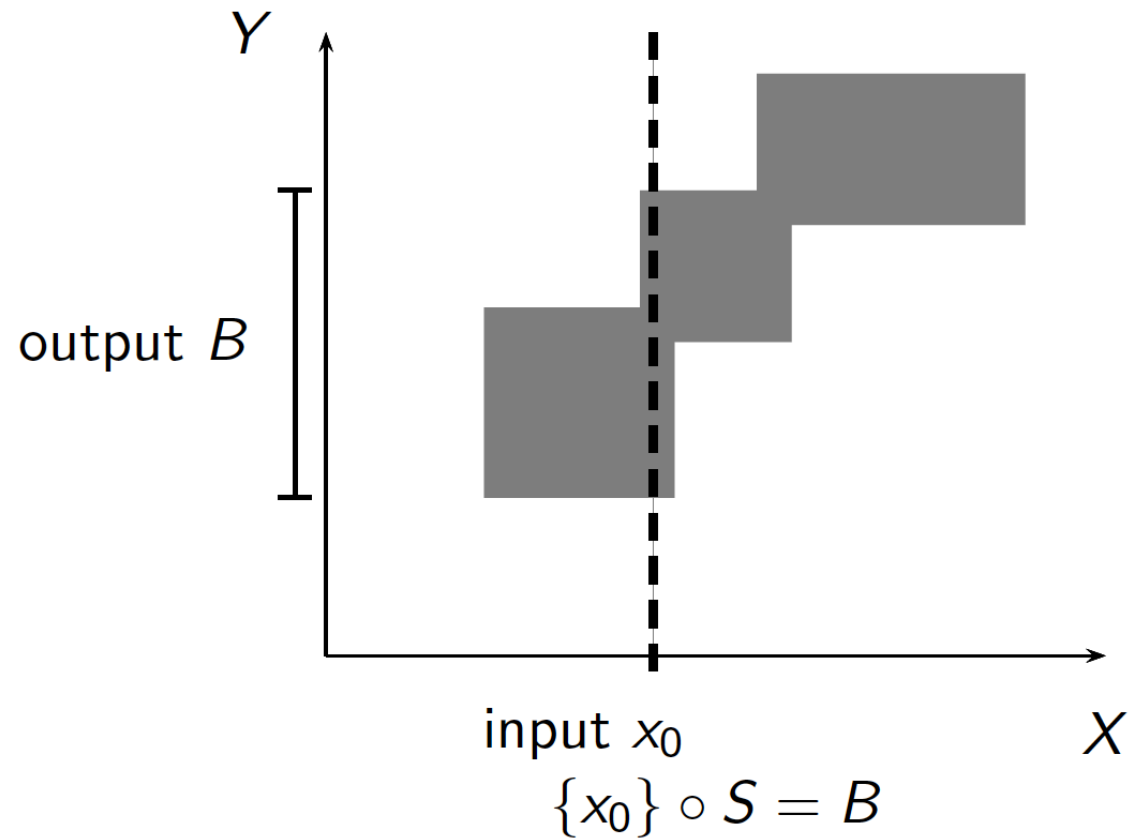
Several imprecise rules: if $X = M_1$ then $Y = N_1$,
if $X = M_2$ then $Y = N_2$, if $X = M_3$ then $Y = N_3$.

Interpretation: rule 1 as well as rule 2 as well as rule 3 hold true.



Approximate Reasoning:

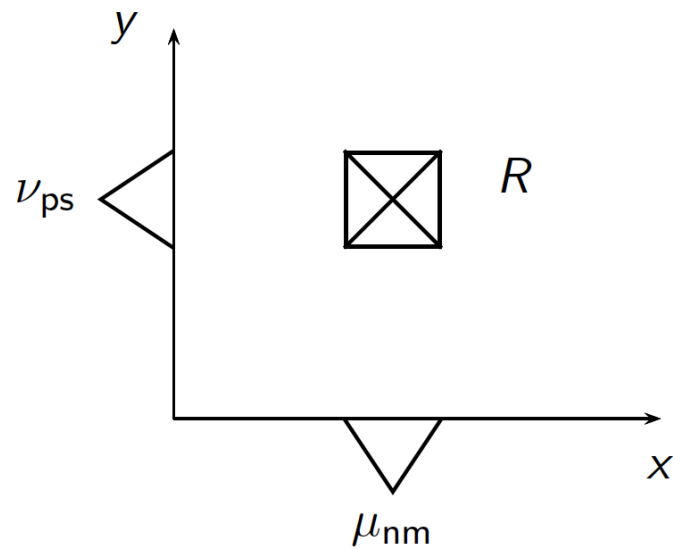
Conclusion



Approximate Reasoning

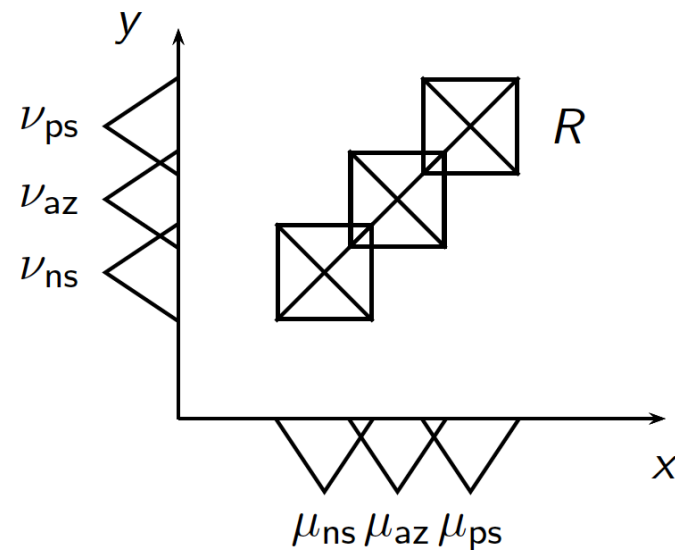
Disjunctive Fuzzy Rules

one fuzzy rule:
if $X = nm$ **then** $Y = ps$



$$R = \mu_{nm} \times \nu_{ps}$$

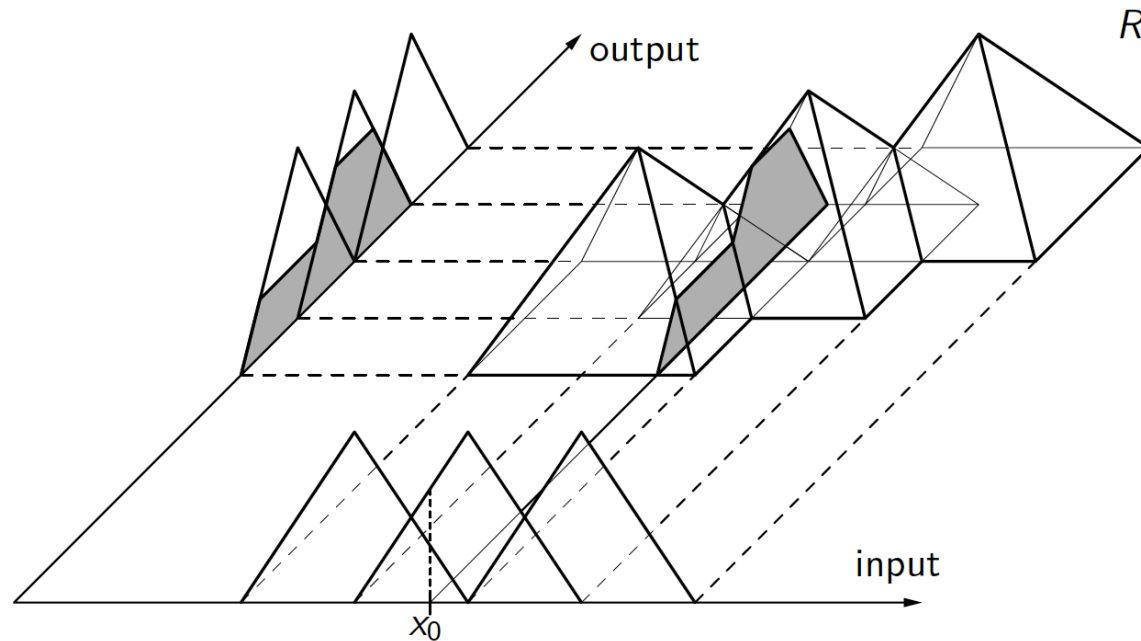
several fuzzy rules:
 $ns \rightarrow ns'$, $az \rightarrow az'$, $ps \rightarrow ps'$



$$R = \mu_{ns} \times \nu_{ns'} \cup \mu_{az} \times \nu_{az'} \cup \mu_{ps} \times \nu_{ps'}$$

Approximate Reasoning

Conclusion



3 fuzzy rules.

Every pyramid is specified by 1 fuzzy rule (Cartesian product).

Input x_0 leads to gray-shaded fuzzy output $\{x_0\} \circ R$.

Disjunctive or Conjunctive?

Fuzzy relation R employed in reasoning is obtained as follows.

For each rule i, we determine relation R_i by

$$R_i(x, y) = \min[M_i(x), N_i(y)]$$

for all $x \in X, y \in Y$.

Then, R is defined by union of R_i , i.e.

$$R = \bigcup_{1 \leq i \leq r} R_i.$$

That is, if-then rules are treated disjunctive.

If-then rules can be also treated conjunctive by

$$R = \bigcap_{1 \leq i \leq r} R_i.$$

Disjunctive or Conjunctive?

Decision depends on intended use and how R_i are obtained.

For both interpretations, two possible ways of applying composition:

$$B'_1 = A' \circ \left(\bigcup_{1 \leq i \leq r} R_i \right)$$

$$B'_3 = \bigcup_{1 \leq i \leq r} A' \circ R_i$$

$$B'_2 = A' \circ \left(\bigcap_{1 \leq i \leq r} R_i \right)$$

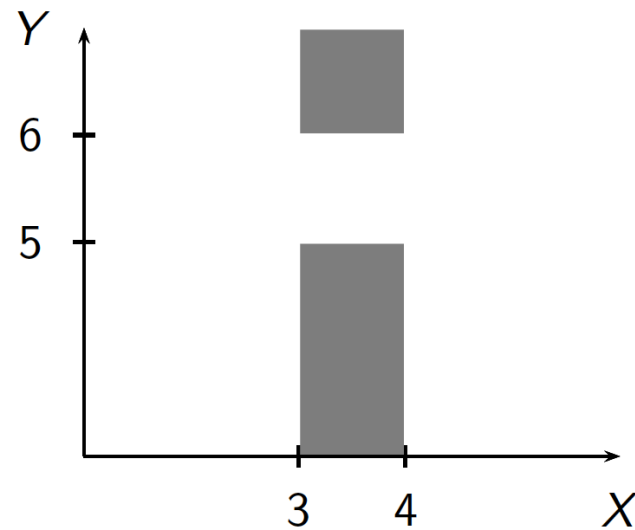
$$B'_4 = \bigcap_{1 \leq i \leq r} A' \circ R_i$$

Approximate Reasoning

Conjunctive Imprecise Rules

if $X = [3, 4]$ **then** $Y = [5, 6]$

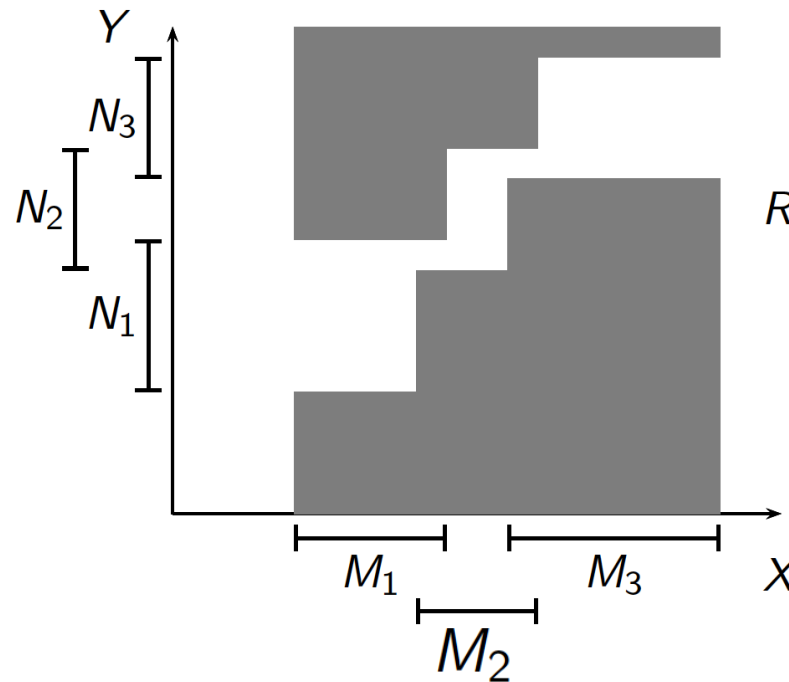
Gray-shaded values are impossible, white ones are possible.



Approximate Reasoning

Conjunctive Imprecise Rules

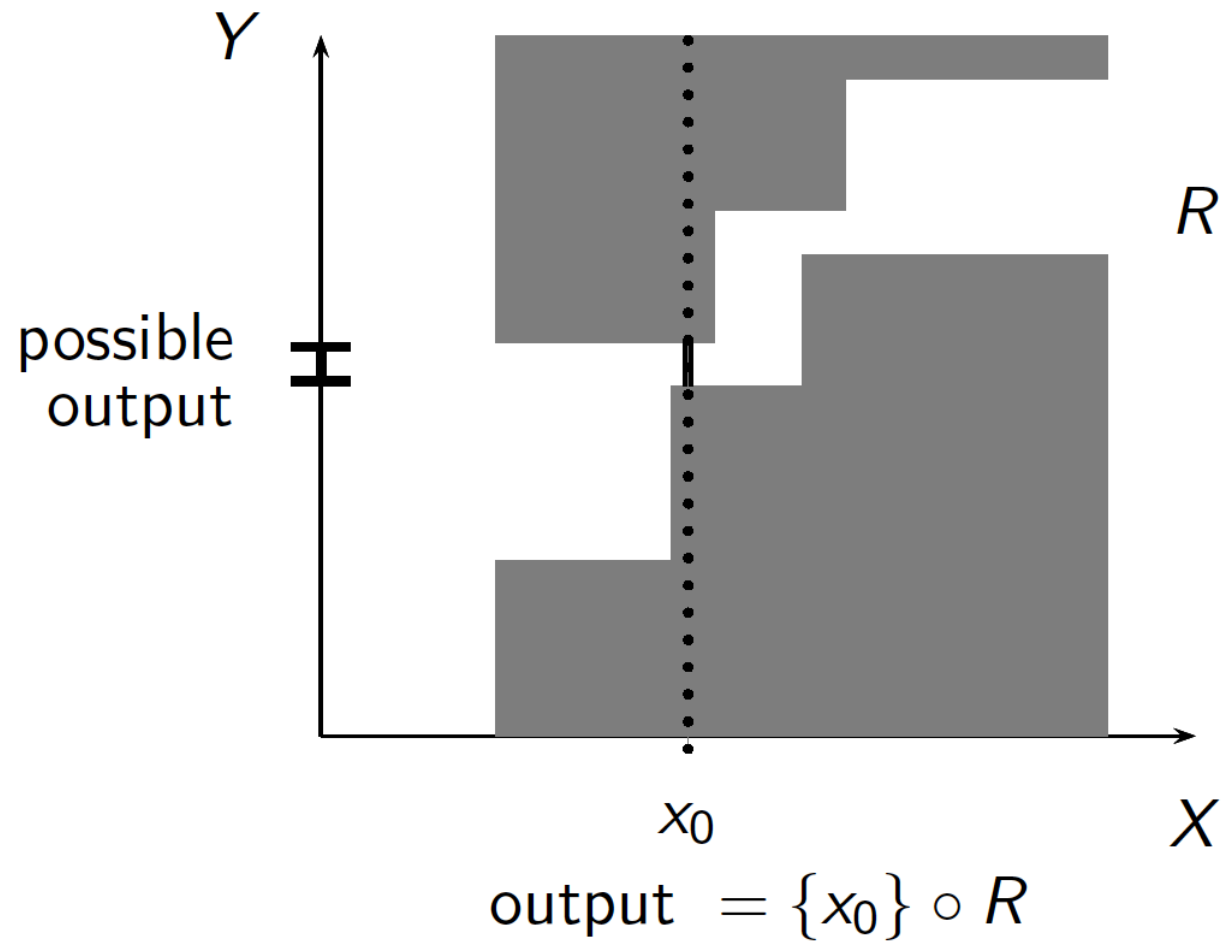
Several imprecise rules: if $X = M_1$ then $Y = N_1$,
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still possible are
$$R = \bigcap_{i=1}^r (M_i \times N_i) \cup (M_i^c \times Y)$$

“corridor” describing
function’s behavior

Approximate Reasoning with Crisp Input



Example: Fuzzy Relation

Classes of cars $X = \{s, m, h\}$ (small, medium, high quality).

Possible maximum speeds $Y = \{140, 160, 180, 200, 220\}$ (in km/h).

For any $(x, y) \in X \times Y$, fuzzy relation ρ states possibility that maximum speed of car of class x is y .

ρ	140	160	180	200	220
s	1	.5	.1	0	0
m	0	.5	1	.5	0
h	0	0	.4	.8	1

Fuzzy Relational Equations

Given μ_1, \dots, μ_r of X and v_1, \dots, v_r of Y and r rules if μ_i then v_i .

What is a **fuzzy relation** ϱ that fits the rule system?

One solution is to find a relation ϱ such that

$$\forall i \in \{1, \dots, r\} : v_i = \mu_i \circ \varrho,$$

$$\mu \circ \varrho : Y \rightarrow [0, 1], y \rightarrow \sup_{x \in X} \min\{\mu(x), \varrho(x, y)\}.$$

Solution of a Relational Equation

Single Rule Largest Solution

Theorem

- i) Let “if A then B” be a rule with $\mu_A \in F(X)$ and $\nu_B \in F(Y)$.
- ii) Then the relational equation $\nu_B = \mu_A \circ \varrho$ can be solved iff the Gödel relation $\varrho_{A \sim B}$ is a solution.

$\varrho_{A \sim B} : X \times Y \rightarrow [0, 1]$ is defined by

$$\begin{aligned} (x, y) &\rightarrow 1 && \text{if } \mu_A(x) \leq \nu_B(y), \\ &\rightarrow \nu_B(y) && \text{otherwise.} \end{aligned}$$

- ii) If ϱ is a solution, then the set of solutions

$$R = \{\varrho_S \in F(X \times Y) \mid \nu_B = \mu_A \circ \varrho_S\}$$

has the following property:

$$\text{If } \varrho_{S'}, \varrho_{S''} \in R, \text{ then } \varrho_{S' \cup S''} \in R.$$

- iii) If $\varrho_{A \sim B}$ is a solution, then $\varrho_{A \sim B}$ is the largest solution w.r.t. \subseteq .

Example

Single Rule Largest Solution

$$\mu_A = (.9 \quad 1 \quad .7)$$

$$\nu_B = (1 \quad .4 \quad .8 \quad .7)$$

$$\mu_{A \odot B}(x,y) \rightarrow \begin{cases} 1 & \text{if } \mu_A(x) \leq \nu_B(y) \\ \nu_B(y) & \text{otherwise} \end{cases}$$

$$\varrho_{A \odot B} = \begin{pmatrix} 1 & .4 & .8 & .7 \\ 1 & .4 & .8 & .7 \\ 1 & .4 & 1 & 1 \end{pmatrix}$$

$$\begin{matrix} (0.9, 1) \rightarrow 1 & (0.9, 0.8) \rightarrow 0.8 \end{matrix}$$

.9	1	.7	1	.4	.8	.7
1			1	.4	.8	.7
.7			1	.4	1	1
.9	1	.7	1	.4	.8	.7

$$\varrho_1 = \begin{pmatrix} 0 & 0 & 0 & .7 \\ 1 & .4 & .8 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$(0.9, 1, 0.7) \odot \varrho_1 = (1 \quad 0.4 \quad 0.8 \quad 0.7) \quad \checkmark \text{ } \varrho_1 \text{ is a solution}$$

$$\varrho_2 = \begin{pmatrix} 0 & .4 & .8 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & .7 \end{pmatrix}$$

$$\mu \vee \varrho_2 = \begin{pmatrix} 0 & .4 & .8 & .7 \\ 1 & .4 & .8 & 0 \\ 0 & 0 & 0 & .7 \end{pmatrix}$$

$\varrho_{A \odot B}$ largest solution, ϱ_1, ϱ_2 are two minimal solutions.

Solution space forms upper semilattice.

Solution of a Set of Relational Equations

Multiple rules, Largest Solution

Generalization of this result to system of r relational equations:

Theorem

Let $v_{Bi} = \mu_{Ai} \circ \varrho$ for $i = 1, \dots, r$ be a system of relational equations.

i) There is a solution iff

$\bigcap_{i=1}^r \varrho_{Ai \sim Bi}$, is a solution. (intersection of the largest solutions)

ii) If $\bigcap_{i=1}^r \varrho_{Ai \sim Bi}$ is a solution, then this solution is the biggest solution w.r.t. \subseteq .

Remark: if there is no solution, then Gödel relation is at least a good approximation.

Solving a System of Relational Equations

Single Rule Smallest Solution

Sometimes it is a good choice not to use the largest but one of the smallest solutions.

i.e. the Cartesian product $\varrho_{A \times B}(x, y) = \min\{\mu_A(x), v_B(y)\}$.

If a solution of $v_B = \mu_A \circ \varrho$ exists, then $\varrho_{A \times B}$ is a solution, too. (smallest solution)

Theorem

Let $\mu_A \in F(X)$, $v_B \in F(Y)$. Furthermore, let $\varrho \in F(X \times Y)$ be a fuzzy relation which satisfies the relational equation $v_B = \mu_A \circ \varrho$.

Then $v_B = \mu_A \circ \varrho_{A \times B}$ holds.

(i.e. if a solution exists then cartesian product is a solution)

Solving a System of Relational Equations

Multiple rules, Smallest Solution

Using Cartesian product

$\mu_{A_i} = \nu_{B_i} \circ \varrho$, $1 \leq i \leq r$ can be reasonably solved with $A \times B$ by

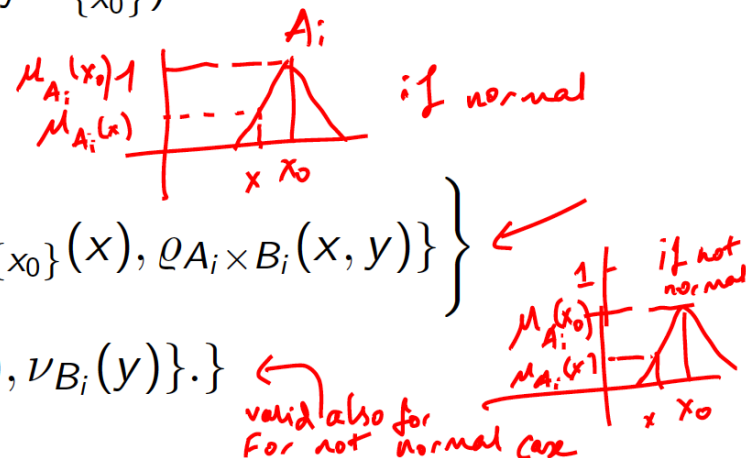
$$\varrho = \max \{ \varrho_{A_i \times B_i} \mid 1 \leq i \leq r \}.$$

For crisp value $x_0 \in X$ (represented by $\mathbb{1}_{\{x_0\}}$):

$$\nu(y) = (\mathbb{1}_{\{x_0\}} \circ \varrho)(y)$$

$$= \max_{1 \leq i \leq r} \left\{ \sup_{x \in X} \min \{ \mathbb{1}_{\{x_0\}}(x), \varrho_{A_i \times B_i}(x, y) \} \right\}$$

$$= \max_{1 \leq i \leq r} \{ \min \{ \mu_{A_i}(x_0), \nu_{B_i}(y) \} \}$$



That is Mamdani-Assilian fuzzy control (to be discussed).