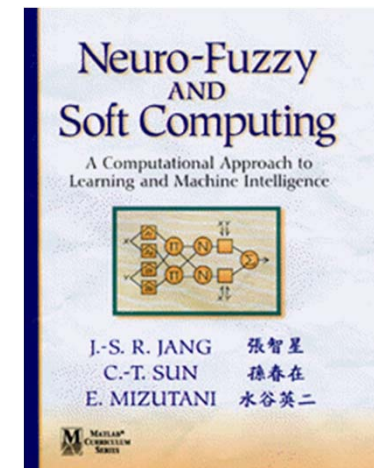


Sistem Cerdas : PTK – Pasca Sarjana - UNY

Fuzzy Rules & Fuzzy Reasoning

👉 Pengampu: Fatchul Arifin

Referensi:



Jyh-Shing Roger Jang et al., *Neuro-Fuzzy and Soft Computing: A Computational Approach to Learning and Machine Intelligence*, First Edition, Prentice Hall, 1997

Fuzzy if-then rules (3.3) (cont.)

– Orthogonality

A term set $T = t_1, \dots, t_n$ of a linguistic variable x on the universe X is **orthogonal** if:

$$\sum_{i=1}^n \mu_{t_i}(x) = 1, \quad \forall x \in X$$

Where the t_i 's are convex & normal fuzzy sets defined on X .

Fuzzy if-then rules (3.3) (cont.)

🔦 General format:

- If x is A then y is B (where A & B are linguistic values defined by fuzzy sets on universes of discourse X & Y).
 - “ x is A ” is called the antecedent or premise
 - “ y is B ” is called the consequence or conclusion
- Examples:
 - If pressure is high, then volume is small.
 - If the road is slippery, then driving is dangerous.
 - If a tomato is red, then it is ripe.
 - If the speed is high, then apply the brake a little.

Fuzzy if-then rules (3.3) (cont.)

- Meaning of fuzzy if-then-rules ($A \Rightarrow B$)
 - It is a relation between two variables x & y ; therefore it is a binary fuzzy relation R defined on $X * Y$
 - There are two ways to interpret $A \Rightarrow B$:
 - A coupled with B
 - A entails B

if A is coupled with B then:

$$\mathbf{R = A \Rightarrow B = A * B = \int_{X*Y} \mu_A(x) \overset{\sim}{*} \mu_B(y) / (x, y)}$$

where $\overset{\sim}{*}$ is a T - normoperator.

Fuzzy if-then rules (3.3) (cont.)

If A entails B then:

$$R = A \Rightarrow B = \neg A \cup B \text{ (material implication)}$$

$$R = A \Rightarrow B = \neg A \cup (A \cap B) \text{ (propositional calculus)}$$

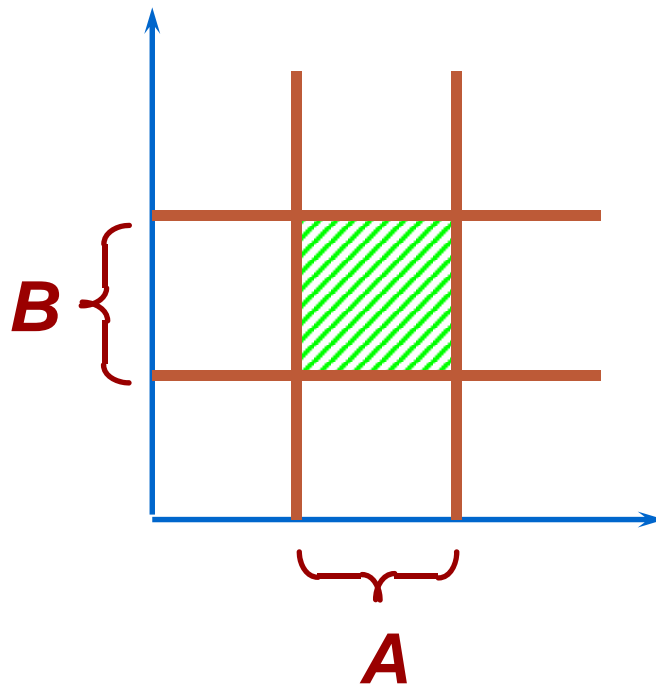
$$R = A \Rightarrow B = (\neg A \cap \neg B) \cup B \text{ (extended propositional calculus)}$$

$$\mu_R(\mathbf{x}, \mathbf{y}) = \sup \left\{ c; \mu_A(\mathbf{x}) * c \leq \mu_B(\mathbf{y}), 0 \leq c \leq 1 \right\}$$

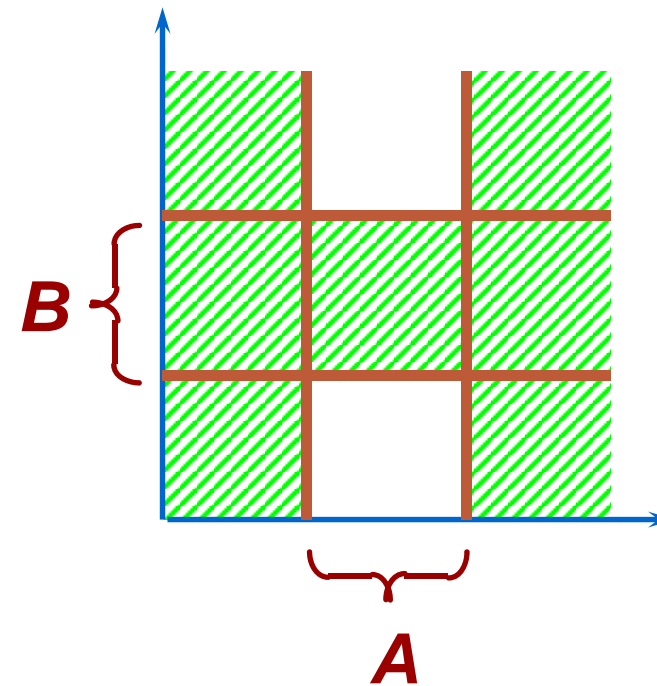
Fuzzy if-then rules (3.3) (cont.)

Two ways to interpret “If x is A then y is B ”:

A coupled with B



A entails B



Fuzzy if-then rules (3.3) (cont.)

- Note that R can be viewed as a fuzzy set with a two-dimensional MF

$$\mu_R(x, y) = f(\mu_A(x), \mu_B(y)) = f(a, b)$$

With $a = \mu_A(x)$, $b = \mu_B(y)$ and f called the fuzzy implication function provides the membership value of (x, y)

Fuzzy if-then rules (3.3) (cont.)

– Case of “A coupled with B”

$$R_m = A * B = \int_{X*Y} \mu_A(x) \wedge \mu_B(y) / (x, y)$$

(minimum operator proposed by Mamdani, 1975)

$$R_p = A * B = \int_{X*Y} \mu_A(\mathbf{x}) \mu_B(\mathbf{y}) / (\mathbf{x}, \mathbf{y})$$

(product proposed by Larsen, 1980)

$$\begin{aligned} R_{bp} = A * B &= \int_{X*Y} \mu_A(\mathbf{x}) \otimes \mu_B(\mathbf{y}) / (\mathbf{x}, \mathbf{y}) \\ &= \int_{X*Y} \mathbf{0} \vee (\mu_A(\mathbf{x}) + \mu_B(\mathbf{y}) - \mathbf{1}) / (\mathbf{x}, \mathbf{y}) \end{aligned}$$

(bounded product operator)

Fuzzy if-then rules (3.3) (cont.)

- Case of “A coupled with B” (cont.)

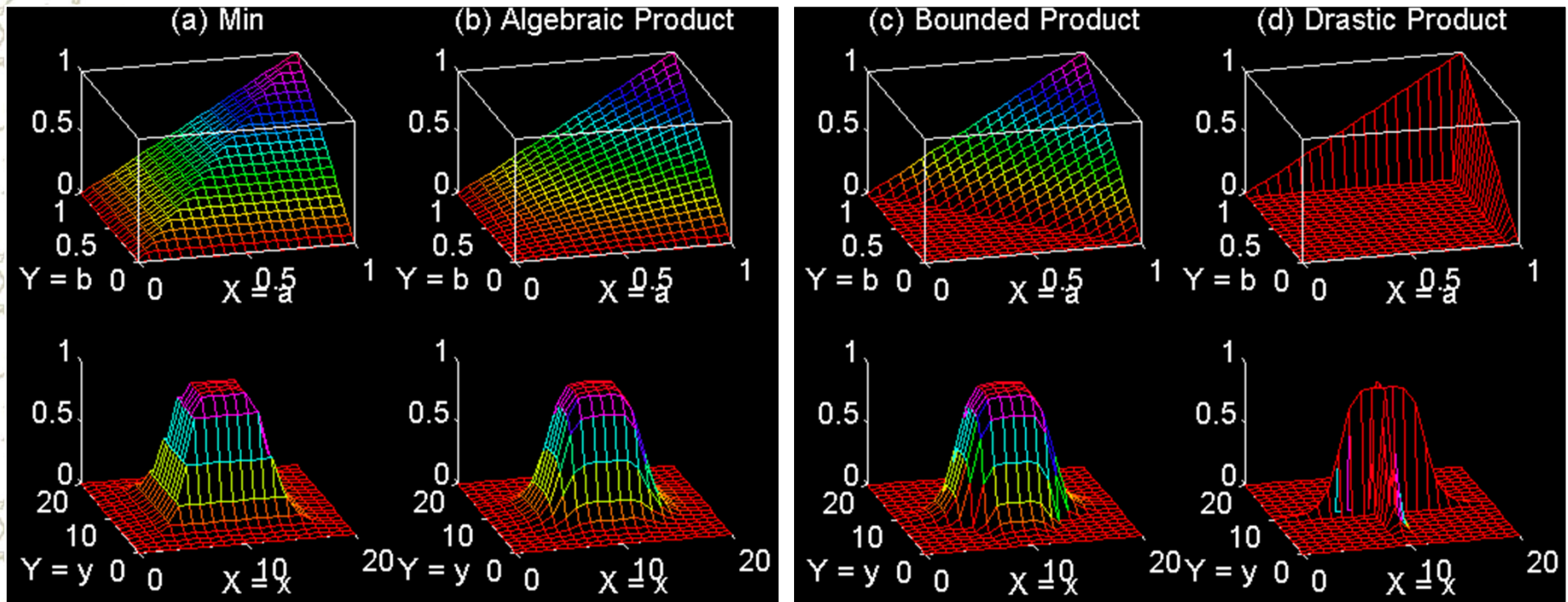
$$R_{dp} = A * B = \int_{X*Y} \mu_A(x) \dot{\wedge} \mu_B(y) / (x, y)$$

$$\text{where : } f(a, b) = a \dot{\wedge} b = \begin{cases} a & \text{if } b = 1 \\ b & \text{if } a = 1 \\ 0 & \text{if otherwise} \end{cases}$$

Example for $\mu_A(x) = bell(x;4,3,10)$ and $\mu_B(y) = bell(y;4,3,10)$

(Drastic operator)

Fuzzy if-then rules (3.3) (cont.)



A coupled with B

Fuzzy if-then rules (3.3) (cont.)

– Case of “A entails B”

$$\mathbf{R}_a = \neg A \cup B = \int_{\mathbf{X} * \mathbf{Y}} \mathbf{1} \wedge (1 - \mu_A(\mathbf{x}) + \mu_B(\mathbf{y})) / (\mathbf{x}, \mathbf{y})$$

$$\text{where : } \mathbf{f}_a(\mathbf{a}, \mathbf{b}) = \mathbf{1} \wedge (1 - \mathbf{a} + \mathbf{b})$$

(Zadeh's arithmetic rule by using bounded sum operator for union)

$$\mathbf{R}_{mm} = \neg A \cup (A \cap B) = \int_{\mathbf{X} * \mathbf{Y}} (1 - \mu_A(\mathbf{x})) \vee (\mu_A(\mathbf{x}) \wedge \mu_B(\mathbf{y})) / (\mathbf{x}, \mathbf{y})$$

$$\text{where : } \mathbf{f}_m(\mathbf{a}, \mathbf{b}) = (1 - \mathbf{a}) \vee (\mathbf{a} \wedge \mathbf{b})$$

(Zadeh's max-min rule)

Fuzzy if-then rules (3.3) (cont.)

- Case of “A entails B” (cont.)

$$\mathbf{R}_s = \neg \mathbf{A} \cup \mathbf{B} = \int_{\mathbf{X} * \mathbf{Y}} (1 - \mu_{\mathbf{A}}(\mathbf{x})) \vee \mu_{\mathbf{B}}(\mathbf{y}) / (\mathbf{x}, \mathbf{y})$$

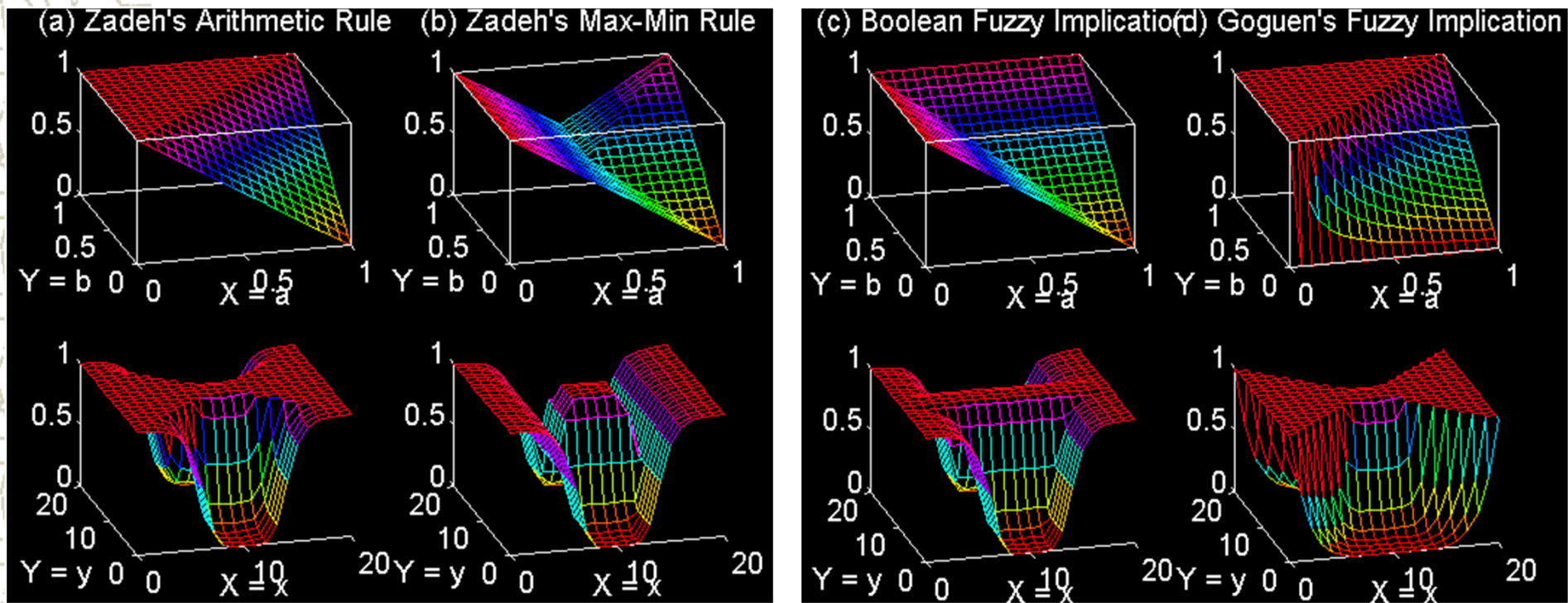
$$\text{where : } \mathbf{f}_s(\mathbf{a}, \mathbf{b}) = (1 - \mathbf{a}) \vee \mathbf{b}$$

(Boolean fuzzy implication with max for union)

$$\mathbf{R}_{\Delta} = \int_{\mathbf{X} * \mathbf{Y}} (\mu_{\mathbf{A}}(\mathbf{x}) \lesssim \mu_{\mathbf{B}}(\mathbf{y})) / (\mathbf{x}, \mathbf{y})$$

$$\text{where : } \mathbf{a} \lesssim \mathbf{b} = \begin{cases} \mathbf{1} & \text{if } \mathbf{a} \leq \mathbf{b} \\ \mathbf{b} / \mathbf{a} & \text{otherwise} \end{cases}$$

(Goguen's fuzzy implication with algebraic product for T-norm)



A entails B

Fuzzy Reasoning (3.4)

🔦 Definition

- Known also as approximate reasoning
- It is an inference procedure that derives conclusions from a set of fuzzy if-then-rules & known facts

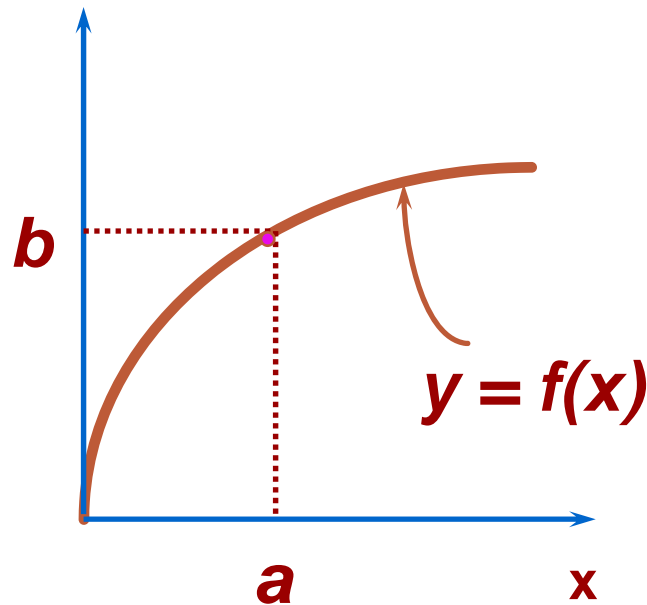
Fuzzy Reasoning (3.4) (cont.)

💡 Compositional rule of inference

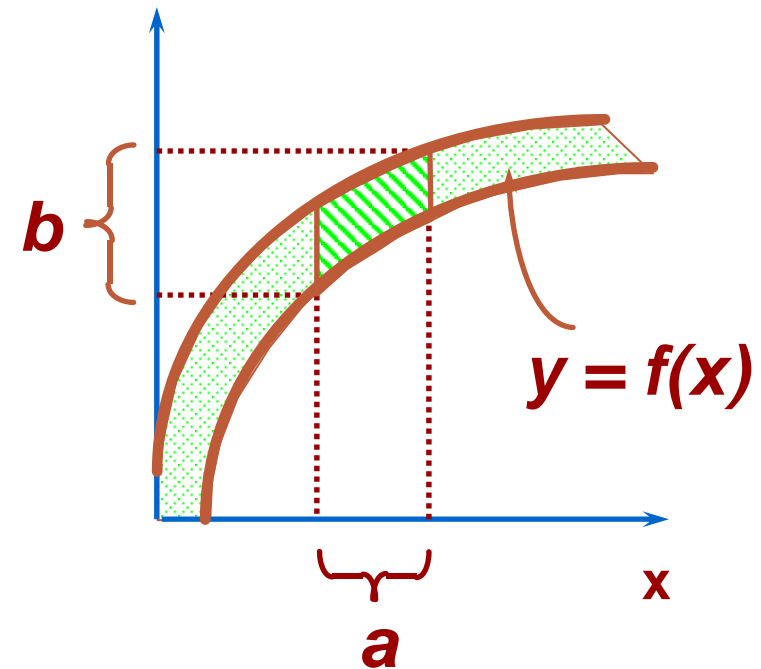
- Idea of composition (cylindrical extension & projection)
- Computation of b given a & f is the goal of the composition
 - Image of a point is a point
 - Image of an interval is an interval

Fuzzy Reasoning (3.4) (cont.)

Derivation of $y = b$ from $x = a$ and $y = f(x)$:



a and b : points
 $y = f(x)$: a curve



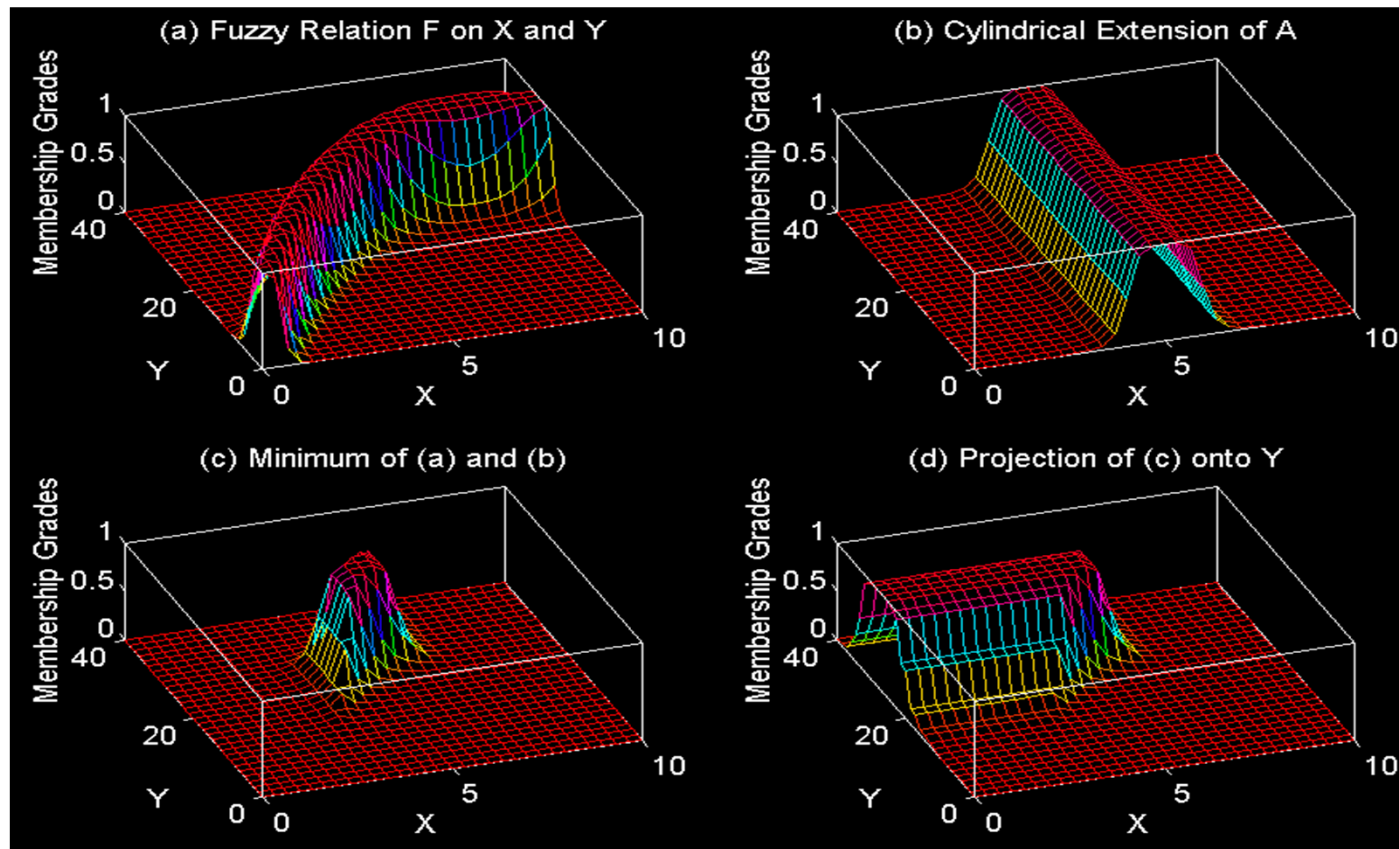
a and b : intervals
 **$y = f(x)$: an interval-valued
 function**

Fuzzy Reasoning (3.4) (cont.)

- The extension principle is a special case of the compositional rule of inference
 - F is a fuzzy relation on $X*Y$, A is a fuzzy set of X & the goal is to determine the resulting fuzzy set B
 - Construct a cylindrical extension $c(A)$ with base A
 - Determine $c(A) \wedge F$ (using minimum operator)
 - Project $c(A) \wedge F$ onto the y-axis which provides B

Fuzzy Reasoning (3.4) (cont.)

a is a fuzzy set and $y = f(x)$ is a fuzzy relation:



Fuzzy Reasoning (3.4) (cont.)

✿ Given A , $A \Rightarrow B$, infer B

A = “today is sunny”

$A \Rightarrow B$: day = sunny then sky = blue

infer: “sky is blue”

- illustration

Premise 1 (fact): x is A

Premise 2 (rule): if x is A then y is B

Consequence: y is B

Fuzzy Reasoning (3.4) (cont.)

Approximation

A' = “ today is more or less sunny”

B' = “ sky is more or less blue”

- illustration

Premise 1 (fact): x is A'

Premise 2 (rule): if x is A then y is B

Consequence: y is B'

(approximate reasoning or fuzzy reasoning!)

Fuzzy Reasoning (3.4) (cont.)

Definition of fuzzy reasoning

Let A , A' and B be fuzzy sets of X , X , and Y , respectively. Assume that the fuzzy implication $A \Rightarrow B$ is expressed as a fuzzy relation R on $X \times Y$. Then the fuzzy set B induced by “ x is A' ” and the fuzzy rule “if x is A then y is B ” is defined by:

$$\mu_{B'}(y) = \max_x \min[\mu_{A'}(x), \mu_R(x, y)]$$

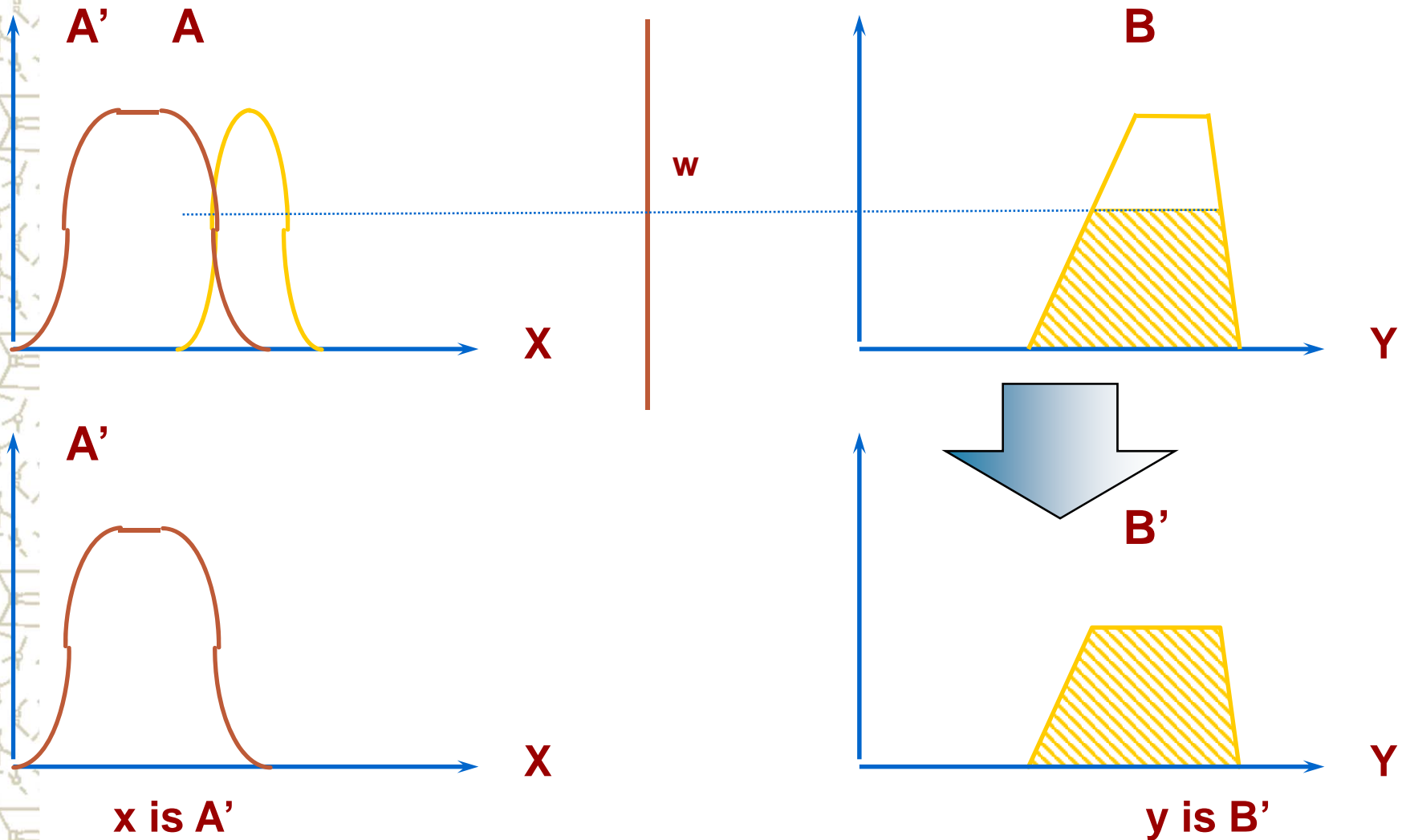
Single rule with single antecedent

Rule : if x is A then y is B

Fact: x is A'

Conclusion: y is B' ($\mu_{B'}(y) = [\bigvee_x (\mu_{A'}(x) \wedge \mu_A(x))] \wedge \mu_B(y)$)

Fuzzy Reasoning (3.4) (cont.)



– Single rule with multiple antecedents

Premise 1 (fact): x is A' and y is B'

Premise 2 (rule): if x is A and y is B then z is C

Conclusion: z is C'

Premise 2: $A * B \rightarrow C$

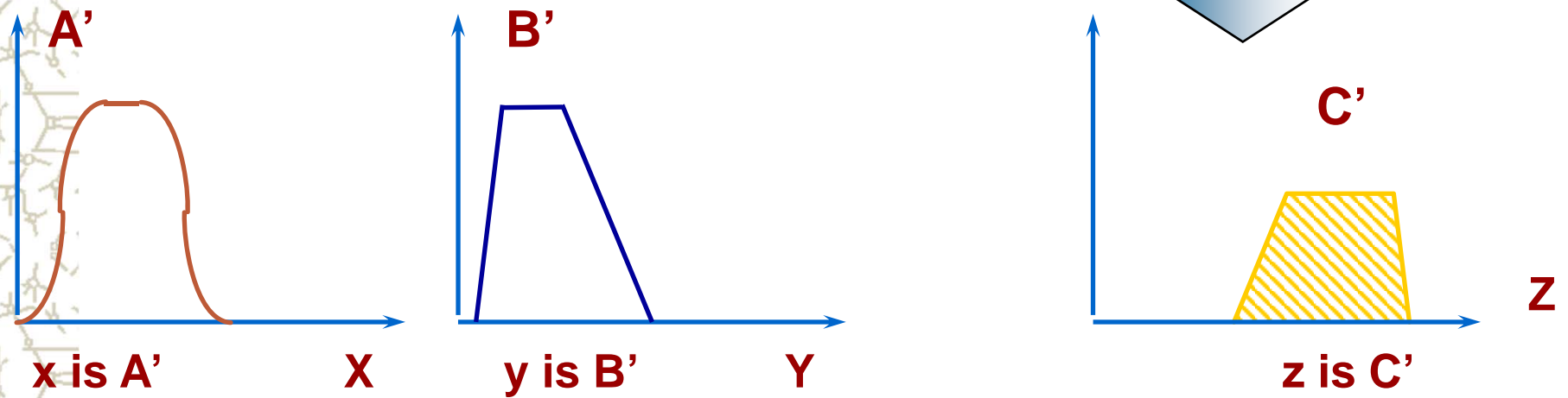
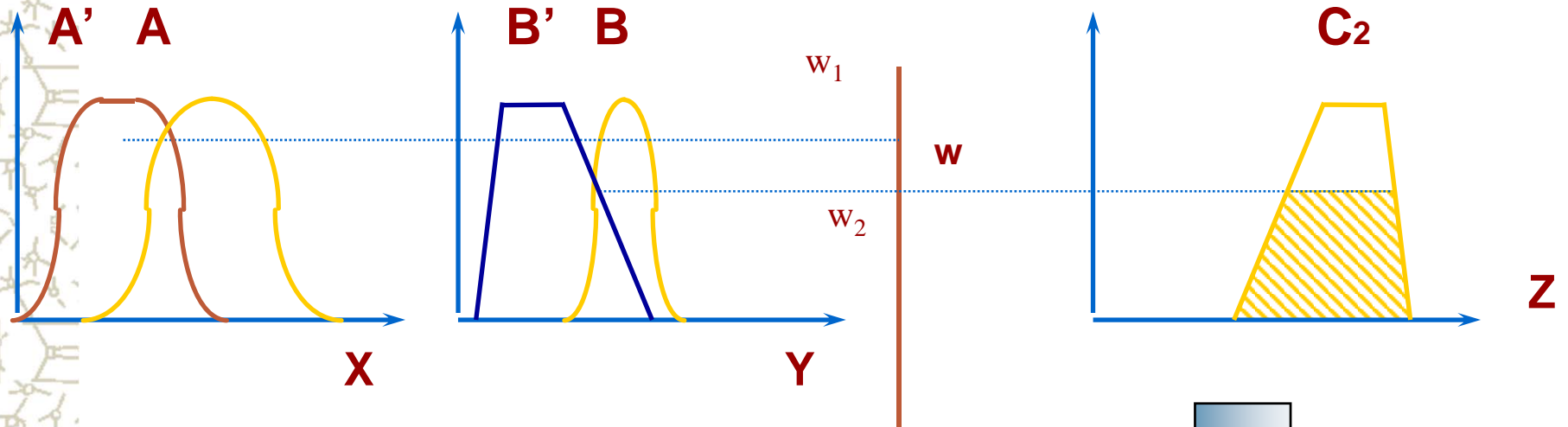
$$\mathbf{R}_{\text{mamdani}}(A, B, C) = (A * B) * C = \int_{\mathbf{X} * \mathbf{Y} * \mathbf{Z}} \mu_A(\mathbf{x}) \wedge \mu_B(\mathbf{y}) \wedge \mu_C(\mathbf{z}) / (\mathbf{x}, \mathbf{y}, \mathbf{z})$$

$$C' = \underbrace{(A' * B')}_{\text{premise 1}} \circ \underbrace{(A * B \rightarrow C)}_{\text{premise 2}}$$

$$\begin{aligned} \mu_{C'}(\mathbf{z}) &= \bigvee_{\mathbf{x}, \mathbf{y}} [\mu_{A'}(\mathbf{x}) \wedge \mu_{B'}(\mathbf{y})] \wedge [\mu_A(\mathbf{x}) \wedge \mu_B(\mathbf{y}) \wedge \mu_C(\mathbf{z})] \\ &= \bigvee_{\mathbf{x}, \mathbf{y}} \{ \mu_{A'}(\mathbf{x}) \wedge \mu_{B'}(\mathbf{y}) \wedge \mu_A(\mathbf{x}) \wedge \mu_B(\mathbf{y}) \} \wedge \mu_C(\mathbf{z}) \\ &= \underbrace{\bigvee_{\mathbf{x}} [\mu_{A'}(\mathbf{x}) \wedge \mu_A(\mathbf{x})]}_{\mathbf{w}_1} \wedge \underbrace{\bigvee_{\mathbf{y}} [\mu_{B'}(\mathbf{y}) \wedge \mu_B(\mathbf{y})]}_{\mathbf{w}_2} \wedge \mu_C(\mathbf{z}) \\ &= (\mathbf{w}_1 \wedge \mathbf{w}_2) \wedge \mu_C(\mathbf{z}) \end{aligned}$$



T-norm



Fuzzy Reasoning (3.4) (cont.)

– Multiple rules with multiple antecedents

Premise 1 (fact): x is A' and y is B'

Premise 2 (rule 1): if x is A₁ and y is B₁ then z is C₁

Premise 3 (rule 2): If x is A₂ and y is B₂ then z is C₂

Consequence (conclusion): z is C'

$$R_1 = A_1 * B_1 \rightarrow C_1$$

$$R_2 = A_2 * B_2 \rightarrow C_2$$

Since the max-min composition operator \circ is distributive over the union operator, it follows:

$$C' = (A' * B') \circ (R_1 \cup R_2) = [(A' * B') \circ R_1] \cup [(A' * B') \circ R_2] = C'_1 \cup C'_2$$

Where C'₁ & C'₂ are the inferred fuzzy set for rules 1 & 2 respectively

