



KTU LECTURE NOTES



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CLASSICAL SETS AND FUZZY SETS

LECTURE 4

November 26, 2017

Classical sets (Crisp sets)

- A *set* is defined as a collection of objects, which share certain characteristics.
- A *classical set* is a collection of distinct objects.
- Each individual entity in a set is called a *member* or an *element* of the set.
- Collection of elements in the universe(U) is called *whole set*.
- Number of elements in U is called *cardinal number*.
- Collection of elements within a set are called *subsets*.
- Classical set is defined as the U is spitted in to two groups: *members and nonmembers*.

No partial membership exists.

Defining a set

- 1 The list of all the members of a set may be given.

$$A = \{2, 4, 6, 8, 10\}$$

- 2 The properties of the set elements may be specified.

$$A = \{x \mid x \text{ is prime number} < 20\}$$

- 3 The formula for the definition of a set may be mentioned.

$$A = \{x_i = \frac{x_i + 1}{5}, i=1 \text{ to } 10, \text{ where } x_i = 1\}$$

- 4 The set may be defined on the basis of the results of a logical operation.

$$A = \{x \mid x \text{ is an element belonging to } P \text{ AND } Q\}$$

- 5 There exist a membership function, which may also be used to define a set.

$$\mu_A(x) = \begin{cases} 1 & \text{if } x \in A \\ 0 & \text{if } x \notin A \end{cases}$$

Operations on classical sets

1 Union

$$A \cup B = \{x | x \in A \text{ or } x \in B\}$$

2 Intersection

$$A \cap B = \{x | x \in A \text{ and } x \in B\}$$

3 Complement

$$\overline{A} = \{x | x \notin A, x \in X\}$$

4 Difference (Subtraction)

$$A \setminus B \text{ or } (A - B) = \{x | x \in A \text{ and } x \notin B\} = A - (A \cap B)$$

$$B \setminus A \text{ or } (B - A) = \{x | x \in B \text{ and } x \notin A\} = B - (B \cap A)$$

Properties of classical sets

■ Commutativity

$$A \cup B = B \cup A; A \cap B = B \cap A$$

■ Associativity

$$A \cup (B \cup C) = (A \cup B) \cup C; A \cap (B \cap C) = (A \cap B) \cap C$$

■ Distributivity

$$A \cup (B \cap C) = (A \cup B) \cap (A \cup C); A \cap (B \cup C) = (A \cap B) \cup (A \cap C)$$

■ Idempotency

$$A \cup A = A; A \cap A = A$$

■ Transitivity

$$A \subseteq B \subseteq C, \text{ then } A \subseteq C$$

■ Identity

$$A \cup \phi = A; A \cap \phi = \phi$$

$$A \cup X = X; A \cap X = X$$

■ Involution

$$\overline{\overline{A}} = A$$

■ Law of excluded middle

$$A \cup \overline{A} = X$$

■ Law of contradiction

$$A \cap \overline{A} = \phi$$

■ DeMorgan's law

$$\overline{A \cap B} = \overline{A} \cup \overline{B}$$

$$\overline{A \cup B} = \overline{A} \cap \overline{B}$$

Function mapping of classical sets

- *Mapping* is a rule of correspondence between set theoretic forms and function theoretic forms.
- A classical set is represented by its characteristic function, $\chi(x)$, where x is the element in the universe.

1 Union ($A \cup B$)

$$\chi_{A \cup B}(x) = \chi_A(x) \vee \chi_B(x) = \max\{\chi_A(x), \chi_B(x)\}$$

2 Intersection ($A \cap B$)

$$\chi_{A \cap B}(x) = \chi_A(x) \wedge \chi_B(x) = \min\{\chi_A(x), \chi_B(x)\}$$

3 Complement (\overline{A})

$$\chi_{\overline{A}}(x) = 1 - \chi_A(x)$$

4 Containment

$$\text{If } A \subseteq B, \text{ then } \chi_A(x) \leq \chi_B(x)$$

Introduction to fuzzy logic

- *Fuzzy logic* is a form of multi-valued logic to deal with reasoning that is approximate rather than precise.
- Fuzzy logic variables may have a truth value that ranges between *0* and *1* and is not constrained to the two truth values of classical proposition logic.
- *0.0* represents *absolute falseness* and *1.0* represents *absolute truth*.
- Fuzzy set is characterized by $(0.0, 0, 1.0)$.

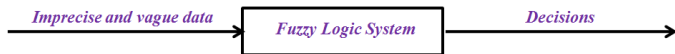


Figure 2.1: A fuzzy logic system accepting imprecise data and providing a decision

Example

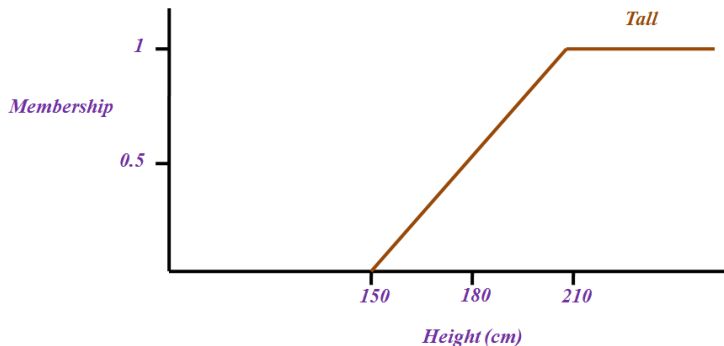


Figure 2.2: Graph showing membership functions for fuzzy set "tall"

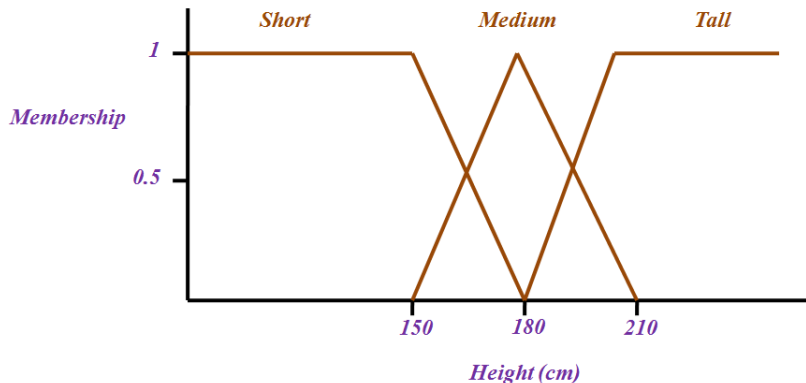


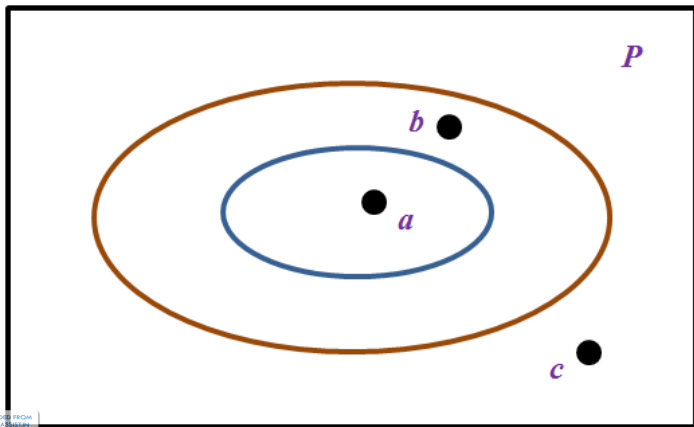
Figure 2.3: Graph showing membership functions for fuzzy sets "short", "medium", "tall"

└ Fuzzy logic

└ Boundary region of a fuzzy set

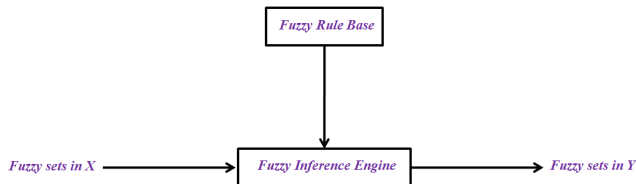
Boundary region of a fuzzy set

X – universe of discourse



Configuration of a pure fuzzy system

- Fuzziness describes the ambiguity of an event and randomness describes the uncertainty in the occurrence of an event.
- Fuzzy logic consists of *fuzzy inference engine* to perform approximate reasoning.
- Fuzzy sets form the building blocks for fuzzy *IF–THEN rules*.
- A *fuzzy system* is a set of fuzzy rules that converts inputs to outputs.



Fuzzy sets

- Fuzzy sets may be viewed as an extension and generalization of the basic concepts of crisp sets.
- It allows partial membership.
- A *fuzzy set* is a set having degrees of membership between *1 and 0*.
- Member of one fuzzy set can also be member of other fuzzy sets in the same universe.
- Vagueness is introduced in fuzzy set by eliminating the sharp boundaries that divide members from non-members in the group.
- Possibility distribution: A fuzzy set A in the universe of discourse U can be defined as a set of ordered pairs and it is given by,

$$A = \{(x, \mu_A(x)) | x \in U\}$$

where, $\mu_A(x)$ is the degree of membership of x in A . $\mu_A(x) \in [0, 1]$

- When U is discrete and finite, fuzzy set A is given as:

$$A = \left\{ \frac{\mu_A(x_1)}{x_1} + \frac{\mu_A(x_2)}{x_2} + \frac{\mu_A(x_3)}{x_3} + \dots \right\} = \left\{ \sum_{i=1}^n \frac{\mu_A(x_i)}{x_i} \right\}$$

- n is a finite value.
- The *summation symbol* $(+)$ indicates the collection of each element.
- When U is continuous and infinite, fuzzy set A is given as:

$$A = \left\{ \int \frac{\mu_A(x)}{x} \right\}$$

- The *integral sign* (\int) is a continuous function—theoretic union for continuous variables.

- A fuzzy set is *universal fuzzy set* if and only if the value of membership function is 1 for all members.

$$\mu_U(x) = 1$$

- The universal fuzzy set can also be called *whole fuzzy set*.
- Two fuzzy sets A and B are equal if,

$$\mu_A(x) = \mu_B(x) \text{ for all } x \in U$$

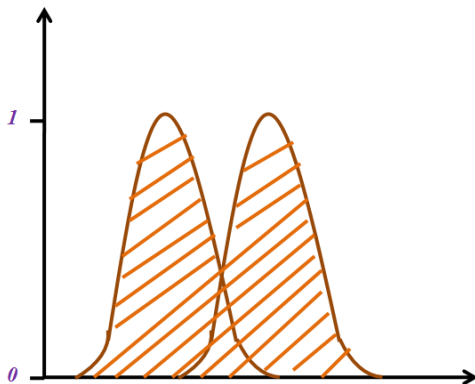
- A fuzzy set A is an *empty fuzzy set* if and only if value of membership function is 0 for all members.

$$\mu_\phi(x) = 0$$

Fuzzy set operations

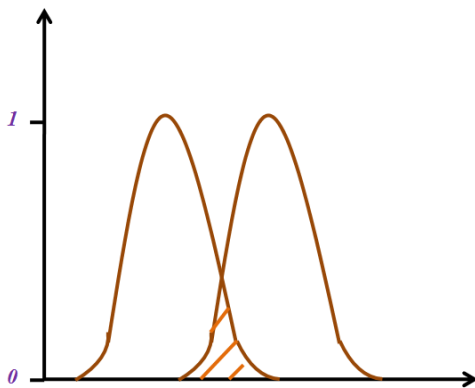
■ Union

$$\mu_{A \cup B}(x) = \max[\mu_A(x), \mu_B(x)] = \mu_A(x) \vee \mu_B(x), \text{ for all } x \in U$$



■ Intersection

$$\mu_{A \cap B}(x) = \min[\mu_A(x), \mu_B(x)] = \mu_A(x) \cap \mu_B(x), \text{ for all } x \in U$$



■ Complement

$$\mu_{\overline{A}} = 1 - \mu_A(x), \text{ for all } x \in U$$

■ Algebraic sum

$$\mu_{A+B}(x) = \mu_A(x) + \mu_B(x) - \mu_A(x) \cdot \mu_B(x)$$

■ Algebraic product

$$\mu_{A.B}(x) = \mu_A(x) \cdot \mu_B(x)$$

■ Bounded sum

$$\mu_{A \oplus B}(x) = \min[1, \mu_A(x) + \mu_B(x)]$$

■ Bounded difference

$$\mu_{A \ominus B}(x) = \max[0, \mu_A(x) - \mu_B(x)]$$

Properties of fuzzy sets

■ Commutativity

$$A \cup B = B \cup A; A \cap B = B \cap A$$

■ Associativity

$$A \cup (B \cup C) = (A \cup B) \cup C; A \cap (B \cap C) = (A \cap B) \cap C$$

■ Distributivity

$$A \cup (B \cap C) = (A \cup B) \cap (A \cup C); A \cap (B \cup C) = (A \cap B) \cup (A \cap C)$$

■ Idempotency

$$A \cup A = A; A \cap A = A$$

■ Transitivity

$$A \subseteq B \subseteq C, \text{ then } A \subseteq C$$

■ Identity

$$A \cup \phi = A; A \cap \phi = \phi$$
$$A \cup X = X; A \cap X = X$$

■ Involution

$$\overline{\overline{A}} = A$$

■ DeMorgan's law

$$\overline{A \cap B} = \overline{A} \cup \overline{B}$$

$$\overline{A \cup B} = \overline{A} \cap \overline{B}$$

Problems

(1) Find the power set and cardinality of the given set $X = \{2, 4, 6\}$. Also find cardinality of power set.

(1) Find the power set and cardinality of the given set $X = \{2, 4, 6\}$. Also find cardinality of power set.

Set X contains 3 elements, so,

$$n_X = 3$$

The power set of X is,

$$P(X) = \{\emptyset, \{2\}, \{4\}, \{6\}, \{2, 4\}, \{2, 6\}, \{4, 6\}, \{2, 4, 6\}\}$$

The cardinality of power set $P(X)$ is,

$$n_{P(X)} = 2^{n_X} = 2^3 = 8$$

(2) Consider 2 given fuzzy sets,

$$A = \left\{ \frac{1}{2} + \frac{0.3}{4} + \frac{0.5}{6} + \frac{0.2}{8} \right\}$$
$$B = \left\{ \frac{0.5}{2} + \frac{0.4}{4} + \frac{0.1}{6} + \frac{1}{8} \right\}$$

Perform,

(a) Union

(b) Intersection

(c) Complement

(d) Difference

(a) Union

$$\begin{aligned}
 A \cup B &= \max\{\mu_A(x), \mu_B(x)\} \\
 &= \left\{ \frac{1}{2} + \frac{0.4}{4} + \frac{0.5}{6} + \frac{1}{8} \right\}
 \end{aligned}$$

(b) Intersection

$$\begin{aligned}
 A \cap B &= \min\{\mu_A(x), \mu_B(x)\} \\
 &= \left\{ \frac{0.5}{2} + \frac{0.3}{4} + \frac{0.1}{6} + \frac{0.2}{8} \right\}
 \end{aligned}$$

(c) Complement

$$\begin{aligned}
 \overline{A} &= 1 - \mu_A(x) = \left\{ \frac{0}{2} + \frac{0.7}{4} + \frac{0.5}{6} + \frac{0.8}{8} \right\} \\
 \overline{B} &= 1 - \mu_B(x) = \left\{ \frac{0.5}{2} + \frac{0.6}{4} + \frac{0.9}{6} + \frac{0}{8} \right\}
 \end{aligned}$$

(d) Difference

$$A|B = A \cap \overline{B} = \left\{ \frac{0.5}{2} + \frac{0.3}{4} + \frac{0.5}{6} + \frac{0}{8} \right\}$$
$$B|A = B \cap \overline{A} = \left\{ \frac{0}{2} + \frac{0.4}{4} + \frac{0.1}{6} + \frac{0.8}{8} \right\}$$

(3) Consider 2 given fuzzy sets,

$$B_1 = \left\{ \frac{1}{1.0} + \frac{0.75}{1.5} + \frac{0.3}{2.0} + \frac{0.15}{2.5} + \frac{0}{3.0} \right\}$$

$$B_2 = \left\{ \frac{1}{1.0} + \frac{0.6}{1.5} + \frac{0.2}{2.0} + \frac{0.1}{2.5} + \frac{0}{3.0} \right\}$$

Perform,

(a) $B_1 \cup B_2$

(b) $B_1 \cap B_2$

(c) $\overline{B_1}$

(d) $\overline{B_2}$

(e) $B_1 | B_2$

(f) $\overline{B_1 \cup B_2}$

(g) $\overline{B_1 \cap B_2}$

(h) $B_1 \cap \overline{B_1}$

(i) $B_1 \cup \overline{B_1}$

(j) $B_2 \cap \overline{B_2}$

(k) $B_2 \cup \overline{B_2}$

$$(a) B_1 \cup B_2 = \left\{ \frac{1}{1.0} + \frac{0.75}{1.5} + \frac{0.3}{2.0} + \frac{0.15}{2.5} + \frac{0}{3.0} \right\}$$

$$(b) B_1 \cap B_2 = \left\{ \frac{1}{1.0} + \frac{0.6}{1.5} + \frac{0.2}{2.0} + \frac{0.1}{2.5} + \frac{0}{3.0} \right\}$$

$$(c) \overline{B_1} = \left\{ \frac{0}{1.0} + \frac{0.25}{1.5} + \frac{0.7}{2.0} + \frac{0.85}{2.5} + \frac{1}{3.0} \right\}$$

$$(d) \overline{B_2} = \left\{ \frac{0}{1.0} + \frac{0.4}{1.5} + \frac{0.8}{2.0} + \frac{0.9}{2.5} + \frac{1}{3.0} \right\}$$

$$(e) B_1|B_2 = B_1 \cap \overline{B_2} = \left\{ \frac{0}{1.0} + \frac{0.4}{1.5} + \frac{0.3}{2.0} + \frac{0.15}{2.5} + \frac{0}{3.0} \right\}$$

$$(f) \overline{B_1 \cup B_2} = \overline{B_1} \cap \overline{B_2} = \left\{ \frac{0}{1.0} + \frac{0.25}{1.5} + \frac{0.7}{2.0} + \frac{0.85}{2.5} + \frac{1}{3.0} \right\}$$

$$(g) \overline{B_1 \cap B_2} = \overline{B_1} \cup \overline{B_2} = \left\{ \frac{0}{1.0} + \frac{0.4}{1.5} + \frac{0.8}{2.0} + \frac{0.9}{2.5} + \frac{1}{3.0} \right\}$$

$$\begin{aligned}
 (h) \quad B_1 \cap \overline{B_1} &= \left\{ \frac{0}{1.0} + \frac{0.25}{1.5} + \frac{0.3}{2.0} + \frac{0.15}{2.5} + \frac{0}{3.0} \right\} \\
 (i) \quad B_1 \cup \overline{B_1} &= \left\{ \frac{1}{1.0} + \frac{0.75}{1.5} + \frac{0.7}{2.0} + \frac{0.85}{2.5} + \frac{1}{3.0} \right\} \\
 (j) \quad B_2 \cap \overline{B_2} &= \left\{ \frac{0}{1.0} + \frac{0.4}{1.5} + \frac{0.2}{2.0} + \frac{0.1}{2.5} + \frac{0}{3.0} \right\} \\
 (k) \quad B_2 \cup \overline{B_2} &= \left\{ \frac{1}{1.0} + \frac{0.6}{1.5} + \frac{0.8}{2.0} + \frac{0.9}{2.5} + \frac{1}{3.0} \right\}
 \end{aligned}$$

(4) It is necessary to compare two sensors based upon their detection levels and gain settings. The table of gain settings and sensor detection levels with a standard item being monitored providing typical membership values to represent the detection levels for each sensor is given in table:

Gain setting	Detection level of sensor 1	Detection level of sensor 2
0	0	0
10	0.2	0.35
20	0.35	0.25
30	0.65	0.8
40	0.85	0.95
50	1	1

Perform union, intersection, complement and difference over sensor 1 and sensor 2.

Given the universe of discourse,

$$X = \{0, 10, 20, 30, 40, 50\}$$

The membership functions for the two sensors in the discrete form as,

$$D_1 = \left\{ \frac{0}{0} + \frac{0.2}{10} + \frac{0.35}{20} + \frac{0.65}{30} + \frac{0.85}{40} + \frac{1}{50} \right\}$$

$$D_2 = \left\{ \frac{0}{0} + \frac{0.35}{10} + \frac{0.25}{20} + \frac{0.8}{30} + \frac{0.95}{40} + \frac{1}{50} \right\}$$

$$D_1 \Rightarrow \text{Sensor1}$$

$$D_2 \Rightarrow \text{Sensor2}$$

(a) Union

$$D_1 \cup D_2 = \max[D_1, D_2] = \left\{ \frac{0}{0} + \frac{0.35}{10} + \frac{0.35}{20} + \frac{0.8}{30} + \frac{0.95}{40} + \frac{1}{50} \right\}$$

(b) Intersection

$$D_1 \cap D_2 = \min[D_1, D_2] = \left\{ \frac{0}{0} + \frac{0.2}{10} + \frac{0.25}{20} + \frac{0.65}{30} + \frac{0.85}{40} + \frac{1}{50} \right\}$$

(c) Complement

$$\overline{D_1} = 1 - D_1 = \left\{ \frac{1}{0} + \frac{0.8}{10} + \frac{0.65}{20} + \frac{0.35}{30} + \frac{0.15}{40} + \frac{0}{50} \right\}$$

$$\overline{D_2} = 1 - D_2 = \left\{ \frac{1}{0} + \frac{0.65}{10} + \frac{0.75}{20} + \frac{0.2}{30} + \frac{0.05}{40} + \frac{0}{50} \right\}$$

(d) Difference

$$D_1|D_2 = D_1 \cap \overline{D_2} = \left\{ \frac{0}{0} + \frac{0.2}{10} + \frac{0.35}{20} + \frac{0.2}{30} + \frac{0.05}{40} + \frac{0}{50} \right\}$$

$$D_2|D_1 = D_2 \cap \overline{D_1} = \left\{ \frac{0}{0} + \frac{0.35}{10} + \frac{0.25}{20} + \frac{0.35}{30} + \frac{0.15}{40} + \frac{0}{50} \right\}$$

(5) Consider 2 given fuzzy sets,

$$A = \left\{ \frac{0.2}{1} + \frac{0.3}{2} + \frac{0.4}{3} + \frac{0.5}{4} \right\}$$

$$B = \left\{ \frac{0.1}{1} + \frac{0.2}{2} + \frac{0.2}{3} + \frac{0.1}{4} \right\}$$

Find,

- (a) algebraic sum
- (b) algebraic product
- (c) bounded sum
- (d) bounded difference

(a) Algebraic sum

$$\begin{aligned}
 \mu_{A+B}(X) &= [\mu_A(x) + \mu_B(x)] - [\mu_A(x) \cdot \mu_B(x)] \\
 &= \left\{ \frac{0.3}{1} + \frac{0.5}{2} + \frac{0.6}{3} + \frac{1.5}{4} \right\} - \left\{ \frac{0.02}{1} + \frac{0.06}{2} + \frac{0.08}{3} + \frac{0.5}{4} \right\} \\
 &= \left\{ \frac{0.28}{1} + \frac{0.44}{2} + \frac{0.52}{3} + \frac{1}{4} \right\}
 \end{aligned}$$

(b) Algebraic product

$$\begin{aligned}
 \mu_{A \cdot B}(X) &= \mu_A(x) \cdot \mu_B(x) \\
 &= \left\{ \frac{0.02}{1} + \frac{0.06}{2} + \frac{0.08}{3} + \frac{0.5}{4} \right\}
 \end{aligned}$$

(c) Bounded sum

$$\begin{aligned}\mu_{A \oplus B}(X) &= \min[1, \mu_A(x) + \mu_B(x)] \\ &= \min\{1, \{\frac{0.3}{1} + \frac{0.5}{2} + \frac{0.6}{3} + \frac{1.5}{4}\}\} \\ &= \{\frac{0.3}{1} + \frac{0.5}{2} + \frac{0.6}{3} + \frac{1}{4}\}\end{aligned}$$

(d) Bounded difference

$$\begin{aligned}\mu_{A \ominus B}(X) &= \max[0, \mu_A(x) - \mu_B(x)] \\ &= \max\{0, \{\frac{0.1}{1} + \frac{0.1}{2} + \frac{0.2}{3} + \frac{0.5}{4}\}\} \\ &= \{\frac{0.1}{1} + \frac{0.1}{2} + \frac{0.2}{3} + \frac{0.5}{4}\}\end{aligned}$$

(6) Show the following fuzzy sets satisfy DeMorgan's law:

$$\mu_A(x) = \frac{1}{1 + 5x}$$
$$\mu_B(x) = \left(\frac{1}{1 + 5x}\right)^{1/2}$$

DeMorgan's law:

$$\overline{A \cup B} = \overline{A} \cap \overline{B}$$

We have,

$$\begin{aligned}\mu_{A \cup B}(x) &= \max[\mu_A(x), \mu_B(x)] \\ &= \frac{\mu_A(x) + \mu_B(x) + |\mu_A(x) - \mu_B(x)|}{2} \\ \mu_{A \cap B}(x) &= \min[\mu_A(x), \mu_B(x)] \\ &= \frac{\mu_A(x) + \mu_B(x) - |\mu_A(x) - \mu_B(x)|}{2}\end{aligned}$$

$$\begin{aligned}
 A \cup B &= \mu_{A \cup B}(x) \\
 &= \frac{\mu_A(x) + \mu_B(x) + |\mu_A(x) - \mu_B(x)|}{2} \\
 &= \frac{\mu_A(x) + \mu_B(x) + |-\mu_B(x) + \mu_A(x)|}{2} \\
 &= \frac{(\because \mu_B(x) > \mu_A(x))}{\mu_A(x) + \mu_B(x) + [\mu_B(x) - \mu_A(x)]} \\
 &= \frac{\mu_A(x) + \mu_B(x) + \mu_B(x) - \mu_A(x)}{2} \\
 &= \frac{2 \times \mu_B(x)}{2} = \mu_B = \left(\frac{1}{1+5x}\right)^{1/2}
 \end{aligned}$$

$$\begin{aligned}\overline{A \cup B} &= 1 - \mu_{A \cup B}(x) = 1 - \left(\frac{1}{1 + 5x}\right)^{1/2} \\ \overline{A} &= 1 - \mu_A(x) = 1 - \frac{1}{1 + 5x} \\ \overline{B} &= 1 - \mu_B(x) = 1 - \left(\frac{1}{1 + 5x}\right)^{1/2}\end{aligned}$$

$$\begin{aligned}
 \overline{A} \cap \overline{B} &= \mu_{\overline{A} \cap \overline{B}}(x) \\
 &= \frac{\mu_{\overline{A}}(x) + \mu_{\overline{B}}(x) - |\mu_{\overline{A}}(x) - \mu_{\overline{B}}(x)|}{2} \\
 &= \frac{\mu_{\overline{A}}(x) + \mu_{\overline{B}}(x) - [\mu_{\overline{A}}(x) - \mu_{\overline{B}}(x)]}{2} \\
 &\quad (\because \mu_{\overline{A}}(x) > \mu_{\overline{B}}(x)) \\
 &= \frac{\mu_{\overline{A}}(x) + \mu_{\overline{B}}(x) - \mu_{\overline{A}}(x) + \mu_{\overline{B}}(x)}{2} \\
 &= \frac{2 \times \mu_{\overline{B}}(x)}{2} = \mu_{\overline{B}} = 1 - \left(\frac{1}{1+5x}\right)^{1/2} \\
 \overline{A \cup B} &= \overline{A} \cap \overline{B} = 1 - \left(\frac{1}{1+5x}\right)^{1/2}
 \end{aligned}$$

Hence, DeMorgan's law is satisfied.

CLASSICAL RELATIONS AND FUZZY RELATIONS

LECTURE 5

November 26, 2017

Relations

- *Relations* represents mapping between sets and connectives in logic.
- A *classical binary relation* represents the presence or absence of a connection between the elements of two sets.
- An ordered r -tuple is an ordered sequence of r -elements is expressed in the form $(a_1, a_2, a_3, \dots, a_r)$.
- For $r = 2$, the r -tuple is called an *ordered pair*.

Cartesian product of relation

- For crisp sets,

$$A_1, A_2, \dots, A_r$$

the set of all r -tuples,

$$(a_1, a_2, a_3, \dots, a_r)$$

where,

$$a_1 \in A_1, a_2 \in A_2, \dots, a_r \in A_r,$$

is called *Cartesian product* and is denoted by,

$$A_1 \times A_2 \times \dots \times A_r$$

- If all elements are identical, then the Cartesian product is denoted as, A^r .

Classical Relation

- An r -ary relation over A_1, A_2, \dots, A_r is a *subset* of the Cartesian product $A_1 \times A_2 \times \dots \times A_r$.

r	r -ary relation
2	binary
3	ternary
4	quaternary
5	quinary

- Consider two universes X and Y , their Cartesian product is given by,

$$X \times Y = \{(x, y) | x \in X, y \in Y\}$$

- The *characteristic function*, χ , gives the strength of the relationship between ordered pair of elements in each universe.

$$\chi_{X \times Y}(x, y) = \begin{cases} 1, & (x, y) \in X \times Y \\ 0, & (x, y) \notin X \times Y \end{cases}$$

- When the sets are finite, then the relation is represented by a matrix called *relation matrix*.

Example

- Consider,

$$X = \{p, q, r\}$$

$$Y = \{2, 4, 6\}$$

Cartesian product of these two sets, $X \times Y$, is,

$$\{(p, 2), (p, 4), (p, 6), (q, 2), (q, 4), (q, 6), (r, 2), (r, 4), (r, 6)\}$$

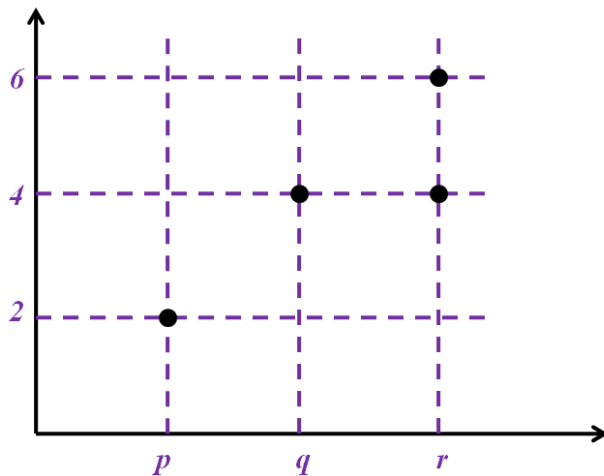
From this set one may select a subset such that,

$$R = \{(p, 2), (q, 4), (r, 4), (r, 6)\}$$

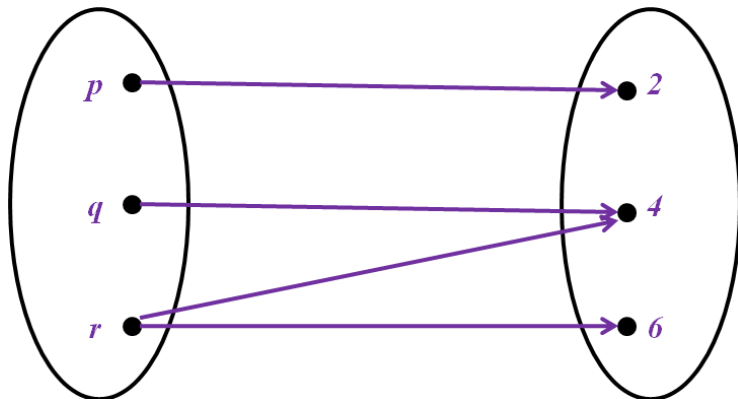
Relation matrix is,

$$\begin{array}{c} \begin{matrix} & 2 & 4 & 6 \end{matrix} \\ \begin{matrix} p \\ q \\ r \end{matrix} \left[\begin{array}{ccc} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{array} \right] \end{array}$$

Coordinate diagram of a relation



Mapping representation of a relation



Cardinality of classical relation

- When the cardinality of,

$$X = n_X \text{ and}$$

$$Y = n_Y,$$

then the cardinality of relation R between the two universe is,

$$n_{X \times Y} = n_X \times n_Y$$

- The cardinality of the power set is given by,

$$n_{P(X \times Y)} = 2^{(n_X n_Y)}$$

Operations on classical relations

■ Union

$$R \cup S \longrightarrow \chi_{R \cup S}(x, y); \chi_{R \cup S}(x, y) = \max[\chi_R(x, y), \chi_S(x, y)]$$

■ Intersection

$$R \cap S \longrightarrow \chi_{R \cap S}(x, y); \chi_{R \cap S}(x, y) = \min[\chi_R(x, y), \chi_S(x, y)]$$

■ Complement

$$\overline{R} \longrightarrow \chi_{\overline{R}}(x, y); \chi_{\overline{R}}(x, y) = 1 - \chi_R(x, y)$$

■ Containment

$$R \subset S \longrightarrow \chi_R(x, y); \chi_R(x, y) \leq \chi_S(x, y)$$

■ Identity

$$\phi \longrightarrow \phi_R \text{ and } X \longrightarrow E_R$$

Properties of crisp relations

- Commutativity
- Associativity
- Distributivity
- Involution
- Idempotency
- Excluded middle laws
- DeMorgan's law

Composition of classical relations

- The operation executed on two compatible binary relations to get a single binary relation is called *composition*.
- Let R be a relation that maps elements from X to Y and S be a relation that maps elements from Y to Z . R and S are compatible if,

$$R \subseteq X \times Y \text{ and } S \subseteq Y \times Z$$

- The composition between the two relations is denoted by $R \circ S$.

Example

- Consider the universal sets,

$$X = \{a_1, a_2, a_3\}$$

$$Y = \{b_1, b_2, b_3\}$$

$$Z = \{c_1, c_2, c_3\}$$

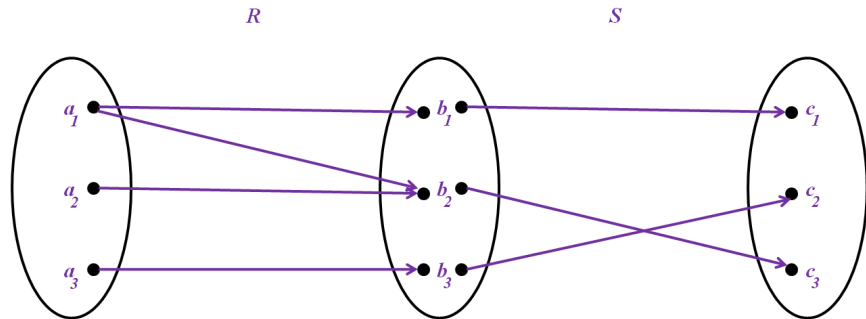
Let the relations R and S be formed as,

$$R = X \times Y = \{(a_1, b_1), (a_1, b_2), (a_2, b_2), (a_3, b_3)\}$$

$$S = Y \times Z = \{(b_1, c_1), (b_2, c_3), (b_3, c_2)\}$$

$$T = R \circ S = \{(a_1, c_1), (a_2, c_3), (a_3, c_2), (a_1, c_3)\}$$

Illustration of relation R and S



- The composition operations are of two types:
 - 1 Max–min composition
 - 2 Max–product composition

Max–min composition

- The max–min composition is defined by the function theoretic expression as:

$$T = R \circ S$$
$$\chi_T(x, z) = \vee \{ \chi_R(x, y) \wedge \chi_S(y, z) \}$$

Max-product composition

- The max-product composition is defined by the function theoretic expression as:

$$T = R \circ S$$
$$\chi_T(x, z) = \vee \{ \chi_R(x, y) \cdot \chi_S(y, z) \}$$

Few properties of composition operation

Associative	$(R \circ S) \circ M = R \circ (S \circ M)$
Commutative	$R \circ S = S \circ R$
Inverse	$(R \circ S)^{-1} = S^{-1} \circ R^{-1}$

Fuzzy Relations

- *Fuzzy relations* relate elements of one universe to those of another universe through the Cartesian product of the two universes.
- Based on the concept that everything is related to some extent or unrelated.
- A fuzzy relation is a fuzzy set defined on the Cartesian product of classical sets,

$$\{X_1, X_2, \dots, X_n\}$$

where tuples,

$$(x_1, x_2, \dots, x_n)$$

may have varying degrees of membership,

$$\mu_R(x_1, x_2, \dots, x_n)$$

within the relation.

$$R(X_1, X_2, \dots, X_n) = \int_{X_1 \times X_2 \times \dots \times X_n} \frac{\mu_R(x_1, x_2, \dots, x_n)}{(x_1, x_2, \dots, x_n)}, x_i \in X_i$$

Fuzzy matrix

■ Let,

$$X = \{x_1, x_2, \dots, x_n\} \text{ and } Y = \{y_1, y_2, \dots, y_m\}$$

Fuzzy relation $R(x, y)$ can be expressed as an $n \times m$ matrix as:

$$R(x, y) = \begin{bmatrix} \mu_R(x_1, y_1) & \mu_R(x_1, y_2) & \dots & \mu_R(x_1, y_m) \\ \mu_R(x_2, y_1) & \mu_R(x_2, y_2) & \dots & \mu_R(x_2, y_m) \\ \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \\ \mu_R(x_n, y_1) & \mu_R(x_n, y_2) & \dots & \mu_R(x_n, y_m) \end{bmatrix}$$

■ The matrix representing a fuzzy relation is called *Fuzzy matrix*.

Fuzzy graph

- *Fuzzy graph* is a graphical representation of binary fuzzy relation.
- Each element in X and Y corresponds to a node in the fuzzy graph.
- The connection links are established between the nodes by the elements of $X \times Y$ with nonzero membership grades in $R(X, Y)$.
- The links may also be present in the form of arcs.
- Links are labeled with the membership values as $\mu_R(x_i, y_j)$.

Bipartite graph

- When $x \neq y$, the link connecting the two nodes is an undirected binary graph called *bipartite graph*.
- Each of the sets X and Y can be represented by a set of nodes such that nodes corresponding to one set are clearly differentiated from the nodes representing the other set.

Directed graph

- When $x = y$, a node is connected to itself and directed links are used.
- Only one set of nodes corresponding to set X is used.

Domain and range

- The domain of a binary fuzzy relation $R(x, y)$ is the fuzzy set, $dom\ R(x, y)$, having the membership function as,

$$\mu_{domain} R(x) = \max_{y \in Y} \mu_R(x, y), \forall x \in X$$

- The range of a binary fuzzy relation $R(x, y)$ is the fuzzy set, $ran\ R(x, y)$, having the membership function as,

$$\mu_{range} R(y) = \max_{x \in X} \mu_R(x, y), \forall y \in Y$$

Example

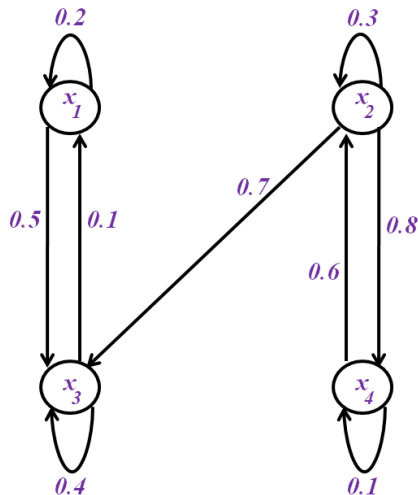
- Consider,

$$X = \{x_1, x_2, x_3, x_4\}$$

Binary fuzzy relation on X as,

	x_1	x_2	x_3	x_4
x_1	0.2	0	0.5	0
x_2	0	0.3	0.7	0.8
x_3	0.1	0	0.4	0
x_4	0	0.6	0	1

Simple fuzzy graph or Directed graph



Example

■ Let,

$$X = \{x_1, x_2, x_3, x_4\} \text{ and } Y = \{y_1, y_2, y_3, y_4\}$$

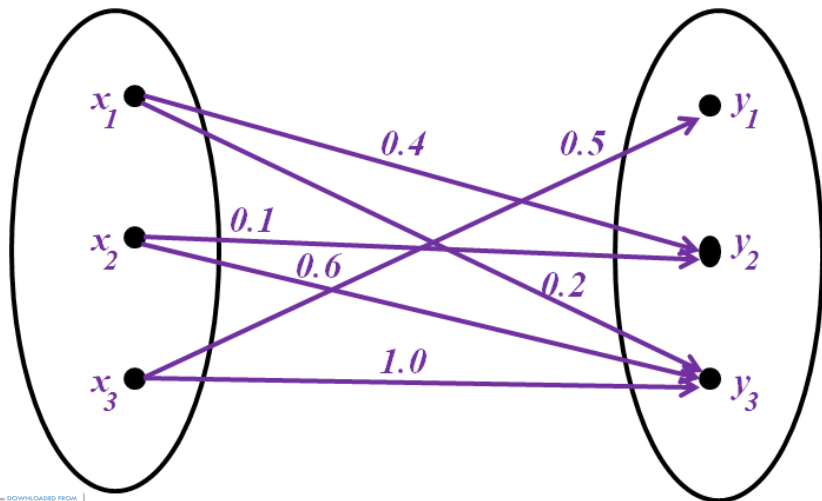
Let R be a relation from X and Y given by,

$$R = \frac{0.2}{(x_1, y_3)} + \frac{0.4}{(x_1, y_2)} + \frac{0.1}{(x_2, y_2)} + \frac{0.6}{(x_2, y_3)} + \frac{1.0}{(x_3, y_3)} + \frac{0.5}{(x_3, y_1)}$$

Fuzzy matrix for relation R is,

$$\begin{array}{c} \begin{array}{ccc} & y_1 & y_2 & y_3 \\ \begin{array}{c} x_1 \\ x_2 \\ x_3 \end{array} & \left[\begin{array}{ccc} 0 & 0.4 & 0.2 \\ 0 & 0.1 & 0.6 \\ 0.5 & 0 & 1.0 \end{array} \right] \end{array}$$

Bipartite graph



Operations on fuzzy relations

■ Union

$$\mu_{R \cup S}(x, y) = \max\{\mu_R(x, y), \mu_S(x, y)\}$$

■ Intersection

$$\mu_{R \cap S}(x, y) = \min\{\mu_R(x, y), \mu_S(x, y)\}$$

■ Complement

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

■ Containment

$$R \subset S \implies \mu_R(x, y) \leq \mu_S(x, y)$$

■ Inverse

$$R^{-1}(y, x) = R(x, y) \text{ for all pairs } (y, x) \in Y \times X$$

■ Projection

$$\mu_{[R \downarrow Y]}(x, y) = \max_x \mu_R(x, y)$$

Properties of fuzzy relations

- Commutativity
- Associativity
- Distributivity
- Identity
- Idempotency
- DeMorgan's law

Fuzzy composition

- Let A be a fuzzy set on universe X and B be a fuzzy set on universe Y .
- the cartesian product over A and B results in fuzzy relation R .ie,

$$A \times B = R$$

where

$$R \subset X \times Y$$

- The membership function is given by,

$$\mu_R(x, y) = \mu_{A \times B}(x, y) = \min[\mu_A(x), \mu_B(y)]$$

Fuzzy composition techniques

■ Max–min composition

- Let R be fuzzy relation on cartesian space $X \times Y$ and S be fuzzy relation on cartesian space $Y \times Z$.
- Max–min composition of $R(X, Y)$ and $S(Y, Z)$,

$$\begin{aligned}\mu_T(x, z) &= \mu_{R \circ S}(x, z) \\ &= \max_{y \in Y} \{ \min[\mu_R(x, y), \mu_S(y, z)] \} \\ &= \vee_{y \in Y} [\mu_R(x, y) \wedge \mu_S(y, z)] \forall x \in X, z \in Z\end{aligned}$$

■ Min-max composition

- Let R be fuzzy relation on cartesian space $X \times Y$ and S be fuzzy relation on cartesian space $Y \times Z$.
- Max-min composition of $R(X, Y)$ and $S(Y, Z)$,

$$\begin{aligned}\mu_T(x, z) &= \mu_{R \circ S}(x, z) \\ &= \min_{y \in Y} \{ \max[\mu_R(x, y), \mu_S(y, z)] \} \\ &= \bigwedge_{y \in Y} [\mu_R(x, y) \vee \mu_S(y, z)] \forall x \in X, z \in Z\end{aligned}$$

■ Max-product composition

- Let R be fuzzy relation on cartesian space $X \times Y$ and S be fuzzy relation on cartesian space $Y \times Z$.
- Max-min composition of $R(X, Y)$ and $S(Y, Z)$,

$$\begin{aligned}\mu_T(x, z) &= \mu_{R \circ S}(x, z) \\ &= \max_{y \in Y} \{ \mu_R(x, y) \cdot \mu_S(y, z) \} \\ &= \bigvee_{y \in Y} [\mu_R(x, y) \cdot \mu_S(y, z)] \forall x \in X, z \in Z\end{aligned}$$

Properties of fuzzy composition

$$1 \quad R \circ S = S \circ R$$

$$2 \quad (R \circ S)^{-1} = S^{-1} \circ R^{-1}$$

$$3 \quad (R \circ S) \circ M = R \circ (S \circ M)$$

Tolerance and Equivalence Relations

- The three characteristic properties of relations are: *reflexivity*, *symmetry* and *transitivity*.
 - 1 A relation is said to be *reflexive* if every vertex in the graph originates a single loop.
 - 2 A relation is said to be *symmetric* if for every edge pointing from vertex i to vertex j , there is an edge pointing in the opposite direction, i.e., from vertex j to i .
 - 3 A relation is said to be *transitive* if for every pair of edges— one pointing from vertex i to vertex j and the other pointing from vertex j to vertex k , then there is an edge pointing from vertex i to vertex k .

Classical Equivalence Relation

- Let relation R on a universe X be a relation from X to X , is an *equivalence relation* if the following 3 properties are satisfied:

1 Reflexivity

$$\chi_R(x_i, x_i) = 1 \text{ or } (x_i, x_i) \in R$$

2 Symmetry

$$\begin{aligned} \chi_R(x_i, x_j) &= \chi_R(x_j, x_i) \\ \text{ie, } (x_i, x_j) \in R &\Rightarrow (x_j, x_i) \in R \end{aligned}$$

3 Transitivity

$$\chi_R(x_i, x_j) \text{ and } \chi_R(x_j, x_k) = 1, \text{ so, } \chi_R(x_i, x_k) = 1$$

Classical Tolerance Relation

- A *tolerance relation* R_1 on universe X is one where the only properties of reflexivity and symmetry are satisfied.
- It can also be called as *proximity relation*.
- An equivalence relation, R , can be formed from tolerance relation R_1 by $(n - 1)$ compositions within itself:

$$R_1^{n-1} = R_1 \circ R_1 \circ \dots \circ R_1 = R$$

where n is the cardinality of the set that defines R_1 .

Fuzzy Equivalence Relation

- The *equivalence relation* can also be called *similarity relation*.
- Let relation R on a universe X be a relation from X to X , is an *fuzzy equivalence relation* if the following 3 properties are satisfied:

1 Reflexivity

$$\mu_R(x_i, x_i) = 1 \forall x \in X$$

If this is not the case for few $x \in X$, then $R(X, X)$ is said to be *irreflexive*.

2 Symmetry

$$\mu_R(x_i, x_j) = \mu_R(x_j, x_i) \forall x_i, x_j \in X$$

If this is not the case for few $x_i, x_j \in X$, then $R(X, X)$ is called *asymmetric*.

3 Transitivity

$$\mu_R(x_i, x_j) = \lambda_1 \text{ and } \mu_R(x_j, x_k) = \lambda_2 \Rightarrow \mu_R(x_i, x_k) = \lambda$$

Fuzzy Tolerance Relation

- A binary fuzzy relation that possesses the properties of reflexivity and symmetry is called *fuzzy tolerance relation*.
- It can also be called as *resemblance relation*.
- A fuzzy equivalence relation, R , can be formed from fuzzy tolerance relation R_1 by $(n - 1)$ compositions within itself:

$$R_1^{n-1} = R_1 \circ R_1 \circ \dots \circ R_1 = R$$

where n is the cardinality of the set that defines R_1 .

Non-interactive Fuzzy sets

- The independent events in probability theory are analogous to non-interactive fuzzy sets in fuzzy theory.
- Defining fuzzy set A on the cartesian space $X = X_1 \times X_2$. Set A is separable into two non-interactive fuzzy sets called *orthogonal projections*, if and only if,

$$A = OPr_{X_1}(A) \times OPr_{X_2}(A)$$

where,

$$\mu_{OPr_{X_1}(A)}(x_1) = \max_{x_2 \in X_2} \mu_A(x_1, x_2), \forall x_1 \in X_1$$

$$\mu_{OPr_{X_2}(A)}(x_2) = \max_{x_1 \in X_1} \mu_A(x_1, x_2), \forall x_2 \in X_2$$

Problems

(1) Consider the following two fuzzy sets:

$$A = \left\{ \frac{0.3}{x_1} + \frac{0.7}{x_2} + \frac{1}{x_3} \right\} \text{ and}$$
$$B = \left\{ \frac{0.4}{y_1} + \frac{0.9}{y_2} \right\}$$

Perform the cartesian product over these given fuzzy sets.

$$\mu_R(x_1, y_1) = \min[\mu_A(x_1), \mu_B(y_1)] = \min[0.3, 0.4] = 0.3$$

$$\mu_R(x_1, y_2) = \min[\mu_A(x_1), \mu_B(y_2)] = \min[0.3, 0.9] = 0.3$$

$$\mu_R(x_2, y_1) = \min[\mu_A(x_2), \mu_B(y_1)] = \min[0.7, 0.4] = 0.4$$

$$\mu_R(x_2, y_2) = \min[\mu_A(x_2), \mu_B(y_2)] = \min[0.7, 0.9] = 0.7$$

$$\mu_R(x_3, y_1) = \min[\mu_A(x_3), \mu_B(y_1)] = \min[1, 0.4] = 0.4$$

$$\mu_R(x_3, y_2) = \min[\mu_A(x_3), \mu_B(y_2)] = \min[1, 0.9] = 0.9$$

$$\mu_R(x_1, y_1) = \min[\mu_A(x_1), \mu_B(y_1)] = \min[0.3, 0.4] = 0.3$$

$$\mu_R(x_1, y_2) = \min[\mu_A(x_1), \mu_B(y_2)] = \min[0.3, 0.9] = 0.3$$

$$\mu_R(x_2, y_1) = \min[\mu_A(x_2), \mu_B(y_1)] = \min[0.7, 0.4] = 0.4$$

$$\mu_R(x_2, y_2) = \min[\mu_A(x_2), \mu_B(y_2)] = \min[0.7, 0.9] = 0.7$$

$$\mu_R(x_3, y_1) = \min[\mu_A(x_3), \mu_B(y_1)] = \min[1, 0.4] = 0.4$$

$$\mu_R(x_3, y_2) = \min[\mu_A(x_3), \mu_B(y_2)] = \min[1, 0.9] = 0.9$$

$$R = A \times B = \begin{matrix} & \begin{matrix} y_1 & y_2 \end{matrix} \\ \begin{matrix} x_1 \\ x_2 \\ x_3 \end{matrix} & \begin{bmatrix} 0.3 & 0.3 \\ 0.4 & 0.7 \\ 0.4 & 0.9 \end{bmatrix} \end{matrix}$$

(2) *Two fuzzy relations are given by,*

$$R = \begin{matrix} & y_1 & y_2 \\ \begin{matrix} x_1 \\ x_2 \end{matrix} & \begin{bmatrix} 0.6 & 0.3 \\ 0.2 & 0.9 \end{bmatrix} \end{matrix} \text{ and}$$

$$S = \begin{matrix} & z_1 & z_2 & z_3 \\ \begin{matrix} y_1 \\ y_2 \end{matrix} & \begin{bmatrix} 1 & 0.5 & 0.3 \\ 0.8 & 0.4 & 0.7 \end{bmatrix} \end{matrix}$$

Obtain fuzzy relation T as a composition between the fuzzy relations R and S .

(a) Max-min composition

$$\begin{aligned}
 \mu_T(x_1, z_1) &= \max\{\min[\mu_R(x_1, y_1), \mu_S(y_1, z_1)], \\
 &\quad \min[\mu_R(x_1, y_2), \mu_S(y_2, z_1)]\} \\
 &= \max\{\min[0.6, 1], \min[0.3, 0.8]\} \\
 &= \max\{0.6, 0.3\} = 0.6
 \end{aligned}$$

$$\begin{aligned}
 \mu_T(x_1, z_2) &= \max\{\min[\mu_R(x_1, y_1), \mu_S(y_1, z_2)], \\
 &\quad \min[\mu_R(x_1, y_2), \mu_S(y_2, z_2)]\} \\
 &= \max\{\min[0.6, 0.5], \min[0.3, 0.4]\} \\
 &= \max\{0.5, 0.3\} = 0.5
 \end{aligned}$$

$$\begin{aligned}
 \mu_T(x_1, z_3) &= \max\{\min[\mu_R(x_1, y_1), \mu_S(y_1, z_3)], \\
 &\quad \min[\mu_R(x_1, y_2), \mu_S(y_2, z_3)]\} \\
 &= \max\{\min[0.6, 0.3], \min[0.3, 0.7]\} \\
 &= \max\{0.3, 0.3\} = 0.3
 \end{aligned}$$

$$\begin{aligned}\mu_T(x_2, z_1) &= \max\{\min[\mu_R(x_2, y_1), \mu_S(y_1, z_1)], \\ &\quad \min[\mu_R(x_2, y_2), \mu_S(y_2, z_1)]\} \\ &= \max\{\min[0.2, 1], \min[0.9, 0.8]\} \\ &= \max\{0.2, 0.8\} = 0.8\end{aligned}$$

$$\begin{aligned}\mu_T(x_2, z_2) &= \max\{\min[\mu_R(x_2, y_1), \mu_S(y_1, z_2)], \\ &\quad \min[\mu_R(x_2, y_2), \mu_S(y_2, z_2)]\} \\ &= \max\{\min[0.2, 0.5], \min[0.9, 0.4]\} \\ &= \max\{0.2, 0.4\} = 0.4\end{aligned}$$

$$\begin{aligned}\mu_T(x_2, z_3) &= \max\{\min[\mu_R(x_2, y_1), \mu_S(y_1, z_3)], \\ &\quad \min[\mu_R(x_2, y_2), \mu_S(y_2, z_3)]\} \\ &= \max\{\min[0.2, 0.3], \min[0.9, 0.7]\} \\ &= \max\{0.2, 0.7\} = 0.7\end{aligned}$$

$$T = R \circ S = \begin{matrix} & & z_1 & z_2 & z_3 \\ \begin{matrix} x_1 \\ x_2 \end{matrix} & \left[\begin{array}{ccc} 0.6 & 0.5 & 0.3 \\ 0.8 & 0.4 & 0.7 \end{array} \right] \end{matrix}$$

(b) Max-product composition

$$\begin{aligned}
 \mu_T(x_1, z_1) &= \max\{[\mu_R(x_1, y_1) \bullet \mu_S(y_1, z_1)], \\
 &\quad [\mu_R(x_1, y_2) \bullet \mu_S(y_2, z_1)]\} \\
 &= \max\{[0.6 \times 1], [0.3 \times 0.8]\} \\
 &= \max\{0.6, 0.24\} = 0.6
 \end{aligned}$$

$$\begin{aligned}
 \mu_T(x_1, z_2) &= \max\{[\mu_R(x_1, y_1) \bullet \mu_S(y_1, z_2)], \\
 &\quad [\mu_R(x_1, y_2) \bullet \mu_S(y_2, z_2)]\} \\
 &= \max\{[0.6 \times 0.5], [0.3 \times 0.4]\} \\
 &= \max\{0.3, 0.12\} = 0.3
 \end{aligned}$$

$$\begin{aligned}
 \mu_T(x_1, z_3) &= \max\{[\mu_R(x_1, y_1) \bullet \mu_S(y_1, z_3)], \\
 &\quad [\mu_R(x_1, y_2) \bullet \mu_S(y_2, z_3)]\} \\
 &= \max\{[0.6 \times 0.3], [0.3 \times 0.7]\} \\
 &= \max\{0.18, 0.21\} = 0.21
 \end{aligned}$$

$$\begin{aligned}\mu_T(x_2, z_1) &= \max\{[\mu_R(x_2, y_1) \bullet \mu_S(y_1, z_1)], \\ &\quad [\mu_R(x_2, y_2) \bullet \mu_S(y_2, z_1)]\} \\ &= \max\{[0.2 \times 1], [0.9 \times 0.8]\} \\ &= \max\{0.2, 0.72\} = 0.72\end{aligned}$$

$$\begin{aligned}\mu_T(x_2, z_2) &= \max\{[\mu_R(x_2, y_1) \bullet \mu_S(y_1, z_2)], \\ &\quad [\mu_R(x_2, y_2) \bullet \mu_S(y_2, z_2)]\} \\ &= \max\{[0.2 \times 0.5], [0.9 \times 0.4]\} \\ &= \max\{0.1, 0.36\} = 0.36\end{aligned}$$

$$\begin{aligned}\mu_T(x_2, z_3) &= \max\{[\mu_R(x_2, y_1) \bullet \mu_S(y_1, z_3)], \\ &\quad [\mu_R(x_2, y_2) \bullet \mu_S(y_2, z_3)]\} \\ &= \max\{[0.2 \times 0.3], [0.9 \times 0.7]\} \\ &= \max\{0.06, 0.63\} = 0.63\end{aligned}$$

$$T = R \circ S = \begin{matrix} & \begin{matrix} z_1 & z_2 & z_3 \end{matrix} \\ \begin{matrix} x_1 \\ x_2 \end{matrix} & \begin{bmatrix} 0.6 & 0.3 & 0.21 \\ 0.72 & 0.36 & 0.63 \end{bmatrix} \end{matrix}$$

(3) *For a speed control of DC motor, the membership functions of series resistance, armature current and speed are given as follows:*

$$\begin{aligned}
 SR &= \left\{ \frac{0.4}{30} + \frac{0.6}{60} + \frac{1.0}{100} + \frac{0.1}{120} \right\} \\
 I &= \left\{ \frac{0.2}{20} + \frac{0.3}{40} + \frac{0.6}{60} + \frac{0.8}{80} + \frac{1.0}{100} + \frac{0.2}{120} \right\} \\
 N &= \left\{ \frac{0.35}{500} + \frac{0.67}{1000} + \frac{0.97}{1500} + \frac{0.25}{1800} \right\}
 \end{aligned}$$

Compute relation T for relating series resistance to motor speed. Perform max–min composition only.

$$R = SR \times I = \begin{array}{c} 30 \\ 60 \\ 100 \\ 120 \end{array} \begin{array}{c} 20 \quad 40 \quad 60 \quad 80 \quad 100 \quad 120 \\ \left[\begin{array}{cccccc} 0.2 & 0.3 & 0.4 & 0.4 & 0.4 & 0.2 \\ 0.2 & 0.3 & 0.6 & 0.6 & 0.6 & 0.2 \\ 0.2 & 0.3 & 0.6 & 0.8 & 1.0 & 0.2 \\ 0.1 & 0.1 & 0.1 & 0.1 & 0.1 & 0.1 \end{array} \right] \end{array}$$

$$S = I \times N = \begin{array}{c} 20 \\ 40 \\ 60 \\ 80 \\ 100 \\ 120 \end{array} \begin{array}{c} 500 \quad 1000 \quad 1500 \quad 1800 \\ \left[\begin{array}{cccc} 0.2 & 0.2 & 0.2 & 0.2 \\ 0.3 & 0.3 & 0.3 & 0.25 \\ 0.35 & 0.6 & 0.6 & 0.25 \\ 0.35 & 0.67 & 0.8 & 0.25 \\ 0.35 & 0.67 & 0.97 & 0.25 \\ 0.2 & 0.2 & 0.2 & 0.2 \end{array} \right] \end{array}$$

$$T = R \circ S = \begin{array}{cc} & \begin{array}{cccc} 500 & 1000 & 1500 & 1800 \end{array} \\ \begin{array}{c} 30 \\ 60 \\ 100 \\ 120 \end{array} & \left[\begin{array}{cccc} 0.35 & 0.4 & 0.4 & 0.25 \\ 0.35 & 0.6 & 0.6 & 0.25 \\ 0.35 & 0.67 & 0.97 & 0.25 \\ 0.1 & 0.1 & 0.1 & 0.1 \end{array} \right] \end{array}$$

(4) Which of the following are equivalence relations?

<i>No.</i>	<i>Set</i>	<i>Relation on the set</i>
(a)	<i>People</i>	<i>is the brother of</i>
(b)	<i>People</i>	<i>has the same parents as</i>
(c)	<i>Points on a map</i>	<i>is connected by a road to</i>
(d)	<i>Lines in plane geometry</i>	<i>is perpendicular to</i>
(e)	<i>Positive Integers</i>	<i>for some integer k equals 10^k times</i>

END