

FUZZY SET THEORY

Vishakha Kelkar

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Classical Set Theory

- Define: An element having certain property belongs to a set or not.
- Set 'A' contained in an universal space 'X'
- We can state whether 'x' is or is not an element of set 'A'
- These sets are called **crisp sets**

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Classical Set Theory

Classical set theory enumerates all its elements using

$$A = \{ a_1, a_2, a_3, a_4, \dots, a_n \}$$

If the elements a_i ($i = 1, 2, 3, \dots, n$) of a set A are subset of universal set X , then set A can be represented for all elements $x \in X$ by its **characteristic function**

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

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Classical Set Theory

$$\left. \begin{aligned} A : X &\rightarrow [0, 1] \\ A(x) &= 1, \text{ } x \text{ is a member of } A \\ A(x) &= 0, \text{ } x \text{ is not a member of } A \end{aligned} \right\} \text{Eq.(1)}$$

Alternatively, the set A can be represented for all elements $x \in X$ by its characteristic function $\mu_A(x)$ defined as

$$\mu_A(x) = \begin{cases} 1 & \text{if } x \in X \\ 0 & \text{otherwise} \end{cases} \quad \text{Eq.(2)}$$

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Member function of Crisp Set

Crisp set: membership of element X of set A is defined by

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



Example: Set of heights from 5 to 7 feet

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Fuzzy sets

Crisp set theory is not capable of representing descriptions and classifications in many cases; In fact, Crisp set does not provide adequate representation for most cases.

The proposition of Fuzzy Sets are motivated by the need to capture and represent real world data with **uncertainty** due to imprecise measurement.

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Fuzzy sets

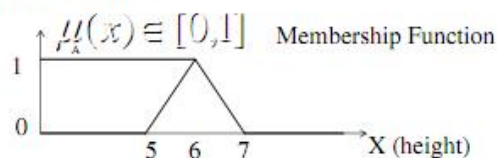
- $\mu_A(x)$ is also called as Characteristic Function of the set 'A'
- For crisp sets this can have only two values 0 OR 1
- Real world problems can not be characterised by such function eg tall person , worm temperature
- The characteristic function can be generalised Which will be called as 'membership function'

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Fuzzy sets

Fuzzy set: Contain objects that satisfy imprecise properties of membership

Example : The set of heights in the region around 6 feet



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Fuzzy sets

Name	Height, cm	Degree of Membership	
		<i>Crisp</i>	<i>Fuzzy</i>
Chris	208	1	1.00
Mark	205	1	1.00
John	198	1	0.98
Tom	181	1	0.82
David	179	0	0.78
Mike	172	0	0.24
Bob	167	0	0.15
Steven	158	0	0.06
Bill	155	0	0.01
Peter	152	0	0.00

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Fuzzy sets

- Many decision-making and problem-solving tasks are too complex to be understood quantitatively, however, **people succeed by using knowledge that is imprecise rather than precise.**
- Fuzzy set theory resembles human reasoning in its use of approximate information and uncertainty to generate decisions.
- It was specifically designed to mathematically represent uncertainty and vagueness and provide formalized tools for dealing with the imprecision intrinsic to many problems.

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Fuzzy sets

- Since knowledge can be expressed in a more natural way by using fuzzy sets, many engineering and decision problems can be greatly simplified.

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Fuzzy sets

A fuzzy set **A** is written as a set of pairs $\{x, A(x)\}$ as

$$A = \{\{x, A(x)\}\}, \quad x \text{ in the set } X$$

where **x** is an element of the universal space **X**, and

A(x) is the value of the function **A** for this element.

The value **A(x)** is the **membership grade** of the element **x** in a fuzzy set **A**.

Define a fuzzy set “ small “ for numbers up to 12.

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Fuzzy sets

Example : Set **SMALL** in set **X** consisting of natural numbers \leq to 12.

Assume: $SMALL(1) = 1$, $SMALL(2) = 1$, $SMALL(3) = 0.9$, $SMALL(4) = 0.6$,
 $SMALL(5) = 0.4$, $SMALL(6) = 0.3$, $SMALL(7) = 0.2$, $SMALL(8) = 0.1$,
 $SMALL(u) = 0$ for $u \geq 9$.

Then, following the notations described in the definition above :

Set SMALL = $\{\{1, 1\}, \{2, 1\}, \{3, 0.9\}, \{4, 0.6\}, \{5, 0.4\}, \{6, 0.3\}, \{7, 0.2\},$
 $\{8, 0.1\}, \{9, 0\}, \{10, 0\}, \{11, 0\}, \{12, 0\}\}$

Note that a fuzzy set can be defined precisely by associating with each **x** ,
its grade of membership in **SMALL**.

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Graphical interpretation of fuzzy set small

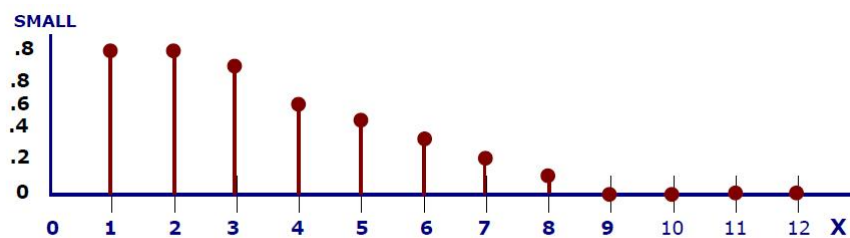


Fig Graphic Interpretation of Fuzzy Sets SMALL

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Fuzzy sets

- Non-Crisp Representation to represent the notion of a tall person.

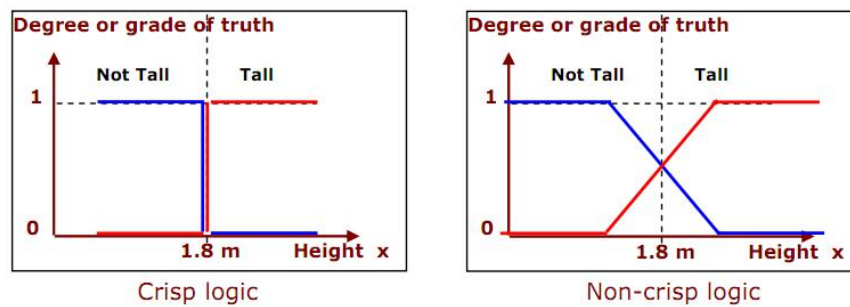


Fig. 1 Set Representation – Degree or grade of truth

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Representation of fuzzy sets

$$A = \mu_1/x_1 + \mu_2/x_2 + \mu_3/x_3 + \dots$$

The symbol / here does not denote division, nor does the symbol + denote summation. The summation symbol is used to connect the terms and thus it means a union of single-term subsets.

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Example

Let the values of temperature in °C under consideration be

$$T = \{0, 5, 10, 15, 20, 25, 30, 35, 40\}.$$

Then, the term *hot* can be defined by a fuzzy set as follows

$$\text{HOT} = \{(0,0), (5,0.1), (10,0.3), (15,0.5), (20,0.6), (25,0.7), (30,0.8), (35,0.9), (40,1.0)\}.$$

This fuzzy set reflects the point of view that 0 °C is not hot at all, 5, 10, and 15 °C are somewhat hot, and 40 °C is indeed hot. Another person could have defined the set differently.

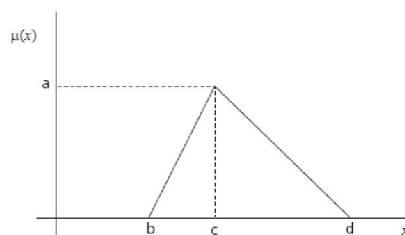
Example

Let $X = \{x_1, x_2, x_3, x_4\}$. One can define a fuzzy set as:

$$A = 0.8/x_1 + 0.4/x_2 + 0.1/x_3 + 0.9/x_4.$$

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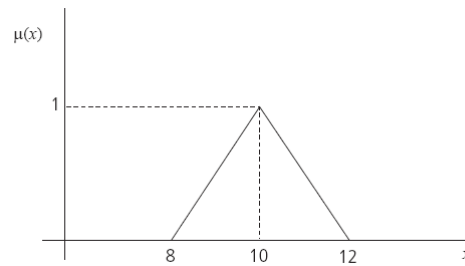
Triangular MF



$$\begin{aligned} \mu(x) &= a(b-x)/(b-c) ; & b &\leq x < c \\ &= a(d-x)/(d-c) ; & c &\leq x < d \\ &= 0 ; & \text{otherwise} \end{aligned}$$

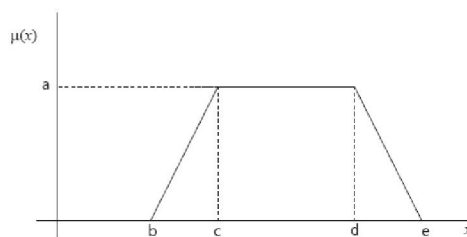
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Vague expression Around ten



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Trapezoidal MF



$$\begin{aligned} \mu(x) &= a (x-b)/(c-b) ; & b \leq x < c \\ &= a ; & c \leq x \leq d \\ &= a (e-x)/(e-d) ; & d < x \leq e \\ &= 0 ; & \text{otherwise} \end{aligned}$$

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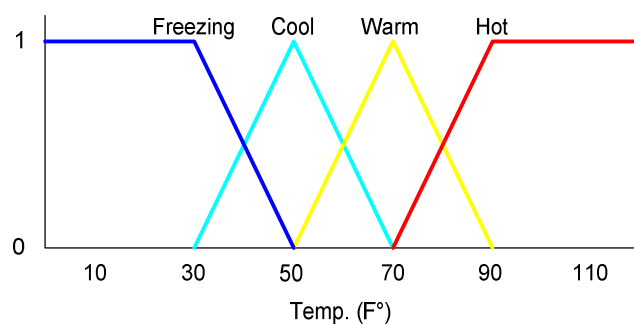
Fuzzy Logic to represent Linguistic variables

- Fuzzy logic:
 - A way to represent variation or imprecision in logic
 - A way to make use of natural language in logic
 - Approximate reasoning
- Humans say things like "If it is sunny and warm today, I will drive fast"
- Linguistic variables:
 - Temp: {freezing, cool, warm, hot}
 - Cloud Cover: {overcast, partly cloudy, sunny}
 - Speed: {slow, fast}

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Membership Functions

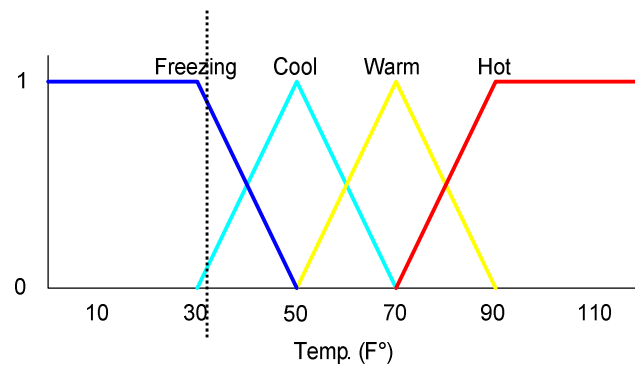
- Temp: {Freezing, Cool, Warm, Hot}
- Degree of Truth or "Membership"



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Membership Functions

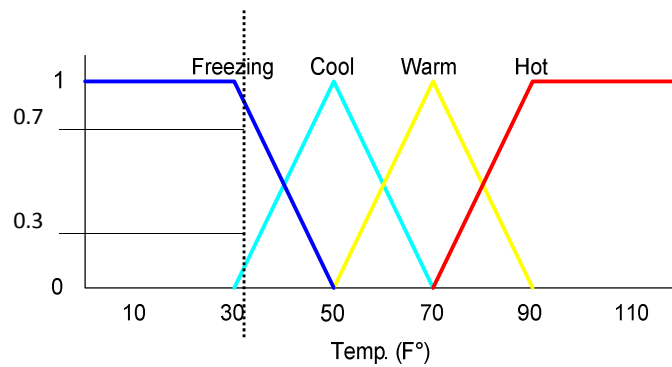
- How cool is 36 F° ?



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Membership Functions

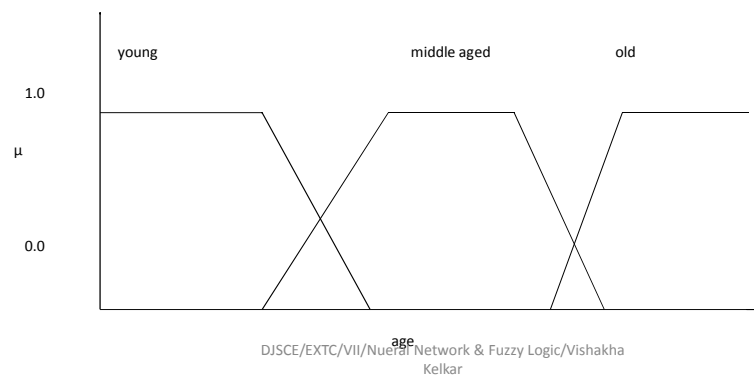
- How cool is 36 F° ?
- It is 30% Cool and 70% Freezing



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Membership Functions

Define membership functions "Young", "Middle aged" and "Old"



Fussy set operations

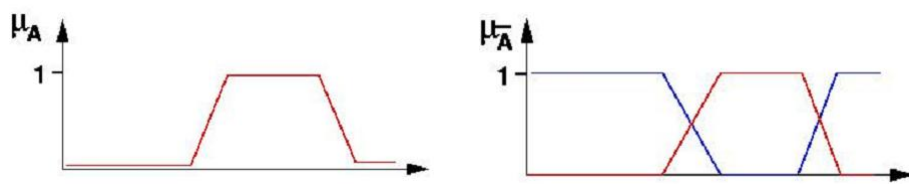
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Compliment of A

COMPLEMENT

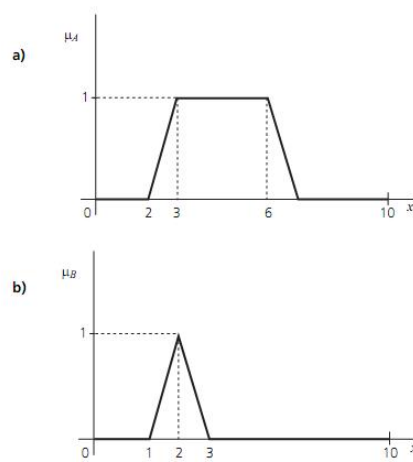
The absolute complement of a fuzzy set A is denoted by \bar{A} and its membership function is defined by

$$\mu_{\bar{A}}(x) = 1 - \mu_A(x) \text{ for all } x \in X$$



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Fussy set operations



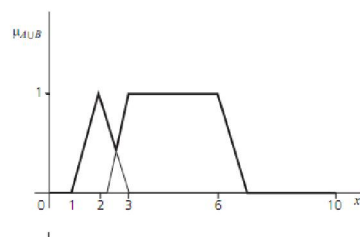
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Fuzzy union

UNION

The union of two fuzzy sets A and B is a fuzzy set whose membership function is defined by

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



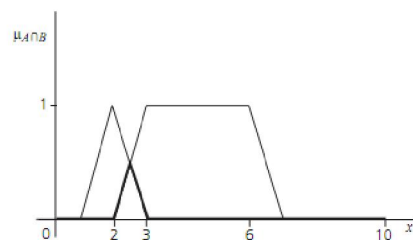
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Fuzzy Intersection

INTERSECTION

The intersection of two fuzzy sets A and B is a fuzzy set whose membership function is defined by

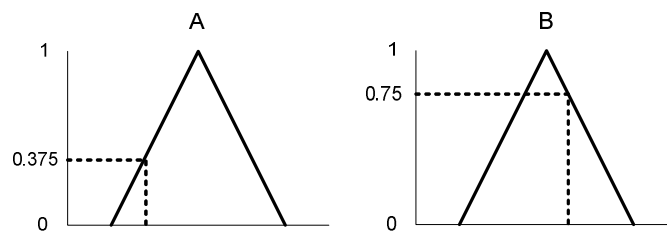
$$\mu_{A \cap B}(x) = \min[\mu_A(x), \mu_B(x)].$$



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Fuzzy Union

- $A \cup B \triangleq \max(A, B)$
- $A \cup B = C$ "Quality C is the union of Quality A and B"

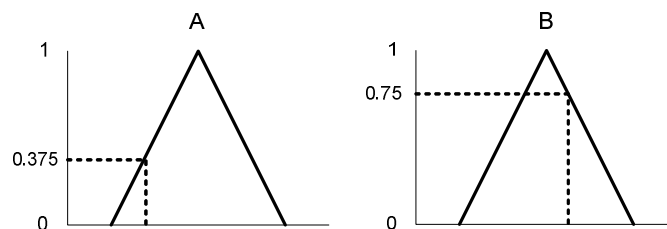


$$(A \vee B = C) \Rightarrow (C = 0.75)$$

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Fuzzy Conjunction

- $A \wedge B \triangleq \min(A, B)$
- $A \wedge B = C$ "Quality C is the conjunction of Quality A and B"



$$(A \wedge B = C) \Rightarrow (C = 0.375)$$

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Properties of Fuzzy Sets

- Equality of two fuzzy sets
- Inclusion of one set into another fuzzy set
- Cardinality of a fuzzy set
- An empty fuzzy set
- α -cuts (alpha-cuts)

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Properties of Fuzzy Sets

Commutativity

$$\begin{aligned} A \cup B &= B \cup A, \\ \tilde{\tilde{A}} \cap \tilde{\tilde{B}} &= \tilde{\tilde{B}} \cap \tilde{\tilde{A}}. \end{aligned}$$

Associativity

$$\begin{aligned} A \cup (B \cap C) &= (A \cup B) \cap C, \\ \tilde{\tilde{A}} \cap (\tilde{\tilde{B}} \cap \tilde{\tilde{C}}) &= (\tilde{\tilde{A}} \cap \tilde{\tilde{B}}) \cap \tilde{\tilde{C}}. \end{aligned}$$

Distributivity

$$\begin{aligned} A \cup (B \cap C) &= (A \cup B) \cap (A \cup C), \\ \tilde{\tilde{A}} \cap (\tilde{\tilde{B}} \cap \tilde{\tilde{C}}) &= (\tilde{\tilde{A}} \cap \tilde{\tilde{B}}) \cup (\tilde{\tilde{A}} \cap \tilde{\tilde{C}}). \end{aligned}$$

▼

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Properties of Fuzzy Sets

Idempotency

$$\begin{aligned} \tilde{A} \cup \tilde{A} &= \tilde{A}, \\ \tilde{A} \cap \tilde{A} &= \tilde{A}. \end{aligned}$$

Identity

$$\begin{aligned} \tilde{A} \cup \phi &= \tilde{A} \quad \text{and} \quad \tilde{A} \cap X = \tilde{A}, \\ \tilde{A} \cap \phi &= \phi \quad \text{and} \quad \tilde{A} \cup X = \tilde{X}. \end{aligned}$$

Transitivity

$$\text{If } \tilde{A} \subset \tilde{B} \subset \tilde{C} \text{ then } \tilde{A} \subset \tilde{C}$$

Involution

$$\overline{\tilde{A}} = \tilde{A}.$$

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Equality

- Fuzzy set A is considered equal to a fuzzy set B , IF AND ONLY IF (iff):

$$\mu_A(x) = \mu_B(x), \forall x \in X$$

$$A = 0.3/1 + 0.5/2 + 1/3$$

$$B = 0.3/1 + 0.5/2 + 1/3$$

therefore $A = B$

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Inclusion

- Inclusion of one fuzzy set into another fuzzy set. Fuzzy set $A \subseteq X$ is included in (is a subset of) another fuzzy set, $B \subseteq X$:

$$\mu_A(x) \leq \mu_B(x), \forall x \in X$$

Consider $X = \{1, 2, 3\}$ and sets A and B

$$A = 0.3/1 + 0.5/2 + 1/3;$$

$$B = 0.5/1 + 0.55/2 + 1/3$$

then A is a subset of B , or $A \subseteq B$

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Cardinality

- Cardinality of a non-fuzzy set, Z , is the number of elements in Z . BUT the cardinality of a fuzzy set A , the so-called SIGMA COUNT, is expressed as a SUM of the values of the membership function of A , $\mu_A(x)$:

$$card_A = \mu_A(x_1) + \mu_A(x_2) + \dots + \mu_A(x_n) = \sum \mu_A(x_i), \quad \text{for } i=1..n$$

Consider $X = \{1, 2, 3\}$ and sets A and B

$$A = 0.3/1 + 0.5/2 + 1/3;$$

$$B = 0.5/1 + 0.55/2 + 1/3$$

$$card_A = 1.8$$

$$card_B = 2.05$$

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Empty Fuzzy Set

- A fuzzy set A is empty, IF AND ONLY IF:

$$\mu_A(x) = 0, \forall x \in X$$

Consider $X = \{1, 2, 3\}$ and set A

$$A = 0/1 + 0/2 + 0/3$$

then A is *empty*

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Alpha-cut

- An α -cut or α -level set of a fuzzy set $A \subseteq X$ is an ORDINARY SET $A_\alpha \subseteq X$, such that:

$$A_\alpha = \{x \in X \mid \mu_A(x) \geq \alpha\}$$

Consider $X = \{1, 2, 3\}$ and set A

$$A = 0.3/1 + 0.5/2 + 1/3$$

then $A_{0.5} = \{2, 3\}$,

$$A_{0.1} = \{1, 2, 3\},$$

$$A_1 = \{3\}$$

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Fuzzy Sets Core and Support

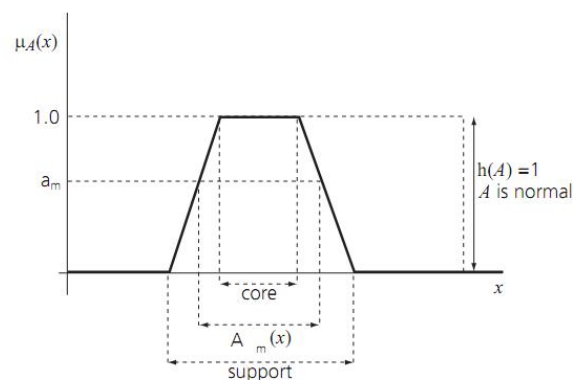
- Assume A is a fuzzy subset of X :
- the **support** of A is the crisp subset of X consisting of all elements with membership grade:

$$\text{supp}(A) = \{x \mid \mu_A(x) > 0 \text{ and } x \in X\}$$
- the **core** of A is the crisp subset of X consisting of all elements with membership grade:

$$\text{core}(A) = \{x \mid \mu_A(x) = 1 \text{ and } x \in X\}$$

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Features of Fuzzy Membership Function Core and Support



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Fuzzy Set Math Operations

- $aA = \{a\mu_A(x), \forall x \in X\}$
 Let $a = 0.5$, and
 $A = \{0.5/a, 0.3/b, 0.2/c, 1/d\}$
 then
 $aA = \{0.25/a, 0.15/b, 0.1/c, 0.5/d\}$
- $A^a = \{\mu_A(x)^a, \forall x \in X\}$
 Let $a = 2$, and
 $A = \{0.5/a, 0.3/b, 0.2/c, 1/d\}$
 then
 $A^a = \{0.25/a, 0.09/b, 0.04/c, 1/d\}$
- ...

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Fuzzy Sets Examples

- Consider two fuzzy subsets of the set X ,
 $X = \{a, b, c, d, e\}$

referred to as A and B

$$A = \{1/a, 0.3/b, 0.2/c, 0.8/d, 0/e\}$$

and

$$B = \{0.6/a, 0.9/b, 0.1/c, 0.3/d, 0.2/e\}$$

**Find support, core, cardinality of A and B .
 Also Find complement A , $A \cup B$ and $A \cap B$**

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Fuzzy Sets Examples

- Support:
 $supp(A) = \{a, b, c, d\}$
 $supp(B) = \{a, b, c, d, e\}$
- Core:
 $core(A) = \{a\}$
 $core(B) = \{\}$
- Cardinality:
 $card(A) = 1 + 0.3 + 0.2 + 0.8 + 0 = 2.3$
 $card(B) = 0.6 + 0.9 + 0.1 + 0.3 + 0.2 = 2.1$

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Fuzzy Sets Examples

- Complement:
 $A = \{1/a, 0.3/b, 0.2/c, 0.8/d, 0/e\}$
 $\neg A = \{0/a, 0.7/b, 0.8/c, 0.2/d, 1/e\}$
- Union:
 $A \cup B = \{1/a, 0.9/b, 0.2/c, 0.8/d, 0.2/e\}$
- Intersection:
 $A \cap B = \{0.6/a, 0.3/b, 0.1/c, 0.3/d, 0/e\}$

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Fuzzy Sets Examples

- \underline{aA} :
for $a=0.5$
 $aA = \{0.5/a, 0.15/b, 0.1/c, 0.4/d, 0/e\}$
- $\underline{A^a}$:
for $a=2$
 $A^a = \{1/a, 0.09/b, 0.04/c, 0.64/d, 0/e\}$
- $\underline{\alpha\text{-cut}}$:
 $A_{0.2} = \{a, b, c, d\}$
 $A_{0.3} = \{a, b, d\}$
 $A_{0.8} = \{a, d\}$
 $A_1 = \{a\}$

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1. Consider two fuzzy sets \underline{A} and \underline{B} find Complement, Union, Intersection, Difference,

$$\underline{A} = \left\{ \frac{1}{2} + \frac{0.5}{3} + \frac{0.6}{4} + \frac{0.2}{5} + \frac{0.6}{6} \right\},$$

$$\underline{B} = \left\{ \frac{0.5}{2} + \frac{0.8}{3} + \frac{0.4}{4} + \frac{0.7}{5} + \frac{0.3}{6} \right\}.$$

2. The membership functions for the two sensors in standard discrete form are

$$S_1 = \left\{ \frac{0.5}{20} + \frac{0.65}{40} + \frac{0.85}{60} + \frac{1}{80} + \frac{1}{100} \right\}$$

$$S_2 = \left\{ \frac{0.35}{20} + \frac{0.5}{40} + \frac{0.75}{60} + \frac{0.90}{80} + \frac{1}{100} \right\}$$

De Morgan's law.

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Given

$$P_{\sim} = \left\{ \frac{0.1}{x_1} + \frac{0.2}{x_2} + \frac{0.7}{x_3} + \frac{0.5}{x_4} + \frac{0.4}{x_5} \right\},$$

$$Q_{\sim} = \left\{ \frac{0.9}{x_1} + \frac{0.6}{x_2} + \frac{0.3}{x_3} + \frac{0.2}{x_4} + \frac{0.8}{x_5} \right\}.$$

Find the following λ cut sets

(a) $\left(\overline{P} \right)_{0.2}$ (b) $(Q)_{0.3}$ (c) $\left(P \cup Q \right)_{0.5}$ (d) $\left(P \cap Q \right)_{0.4}$
 (e) $\left(Q \cup \overline{P} \right)_{0.8}$ (f) $\left(P \cup \overline{P} \right)_{0.2}$.

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Fuzzy Relations

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▪ Cartesian Product of two Crisp Sets

Let **A** and **B** be two crisp sets in the universe of discourse **X** and **Y**.

The Cartesian product of **A** and **B** is denoted by **A x B**

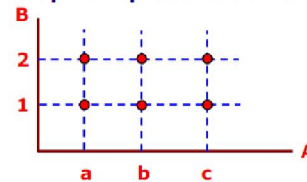
Defined as **A x B = { (a, b) | a ∈ A, b ∈ B }**

Example :

Let **A = {a, b, c}** and **B = {1, 2}**

then **A x B = { (a, 1), (a, 2),
(b, 1), (b, 2),
(c, 1), (c, 2) }**

Graphic representation of A x B



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▪ Cartesian product of two Fuzzy Sets

Let **A** and **B** be two fuzzy sets in the universe of discourse **X** and **Y**.

The Cartesian product of **A** and **B** is denoted by **A x B**

Defined by their membership function **μ_A(x)** and **μ_B(y)** as

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

or
$$\mu_{A \times B}(x, y) = \mu_A(x) \mu_B(y)$$

for all **x ∈ X** and **y ∈ Y**

Thus the Cartesian product **A x B** is a fuzzy set of ordered pair **(x, y)** for all **x ∈ X** and **y ∈ Y**, with grade membership of **(x, y)** in **X x Y** given by the above equations .

In a sense Cartesian product of two Fuzzy sets is a **Fuzzy Relation**.

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Fuzzy relation definition

Consider a Cartesian product

$$A \times B = \{ (x, y) \mid x \in A, y \in B \}$$

where A and B are subsets of universal sets U_1 and U_2 .

Fuzzy relation on $A \times B$ is denoted by R or $R(x, y)$ is defined as the set

$$R = \{ ((x, y), \mu_R(x, y)) \mid (x, y) \in A \times B, \mu_R(x, y) \in [0, 1] \}$$

where $\mu_R(x, y)$ is a function in two variables called membership function.

- It gives the degree of membership of the ordered pair (x, y) in R associating with each pair (x, y) in $A \times B$ a real number in the interval $[0, 1]$.
- The degree of membership indicates the degree to which x is in relation to y .

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Example of Fuzzy Relation

$$R = \{ ((x_1, y_1), 0), ((x_1, y_2), 0.1), ((x_1, y_3), 0.2), ((x_2, y_1), 0.7), ((x_2, y_2), 0.2), ((x_2, y_3), 0.3), ((x_3, y_1), 1), ((x_3, y_2), 0.6), ((x_3, y_3), 0.2) \}$$

The relation can be written in matrix form as

	y	y ₁	y ₂	y ₃
x				
x ₁		0	0.1	0.2
x ₂		0.7	0.2	0.3
x ₃		1	0.6	0.2

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Composition of Fussy sets

The operation composition combines the fuzzy relations in different variables, say (x, y) and (y, z) ; $x \in A, y \in B, z \in C$.

Consider the relations :

$$R_1(x, y) = \{ ((x, y), \mu_{R_1}(x, y)) \mid (x, y) \in A \times B \}$$

$$R_2(y, z) = \{ ((y, z), \mu_{R_2}(y, z)) \mid (y, z) \in B \times C \}$$

The domain of R_1 is $A \times B$ and the domain of R_2 is $B \times C$

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Composition

Max-Min Composition

Definition : The Max-Min composition denoted by $R_1 \circ R_2$ with membership function $\mu_{R_1 \circ R_2}$ defined as

$$R_1 \circ R_2 = \{ ((x, z), \max_y(\min(\mu_{R_1}(x, y), \mu_{R_2}(y, z)))) \}, \\ (x, z) \in A \times C, y \in B$$

Thus $R_1 \circ R_2$ is relation in the domain $A \times C$

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Composition

Example : Max-Min Composition

Consider the relations $R_1(x, y)$ and $R_2(y, z)$ as given below.

$R_1 \triangleq$		y	y_1	y_2	y_3
	x				
	x_1		0.1	0.3	0
	x_2		0.8	1	0.3

$R_2 \triangleq$		z	z_1	z_2	z_3
	y				
	y_1		0.8	0.2	0
	y_2		0.2	1	0.6
	y_3		0.5	0	0.4

Note : Number of columns in the first table and second table are equal.

Compute max-min composition denoted by $R_1 \circ R_2$:

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Step -1 Compute min operation

Consider row x_1 and column z_1 , means the pair (x_1, z_1) for all y_j , $j = 1, 2, 3$, and perform min operation

$$\min(\mu_{R_1}(x_1, y_1), \mu_{R_2}(y_1, z_1)) = \min(0.1, 0.8) = 0.1,$$

$$\min(\mu_{R_1}(x_1, y_2), \mu_{R_2}(y_2, z_1)) = \min(0.3, 0.2) = 0.2,$$

$$\min(\mu_{R_1}(x_1, y_3), \mu_{R_2}(y_3, z_1)) = \min(0, 0.5) = 0,$$

Step -2 Compute max operation (definition in previous slide).

For $x = x_1$, $z = z_1$, $y = y_j$, $j = 1, 2, 3$,

Calculate the grade membership of the pair (x_1, z_1) as

$$\{(x_1, z_1), \max(\min(0.1, 0.8), \min(0.3, 0.2), \min(0, 0.5))\}$$

i.e. $\{(x_1, z_1), \max(0.1, 0.2, 0)\}$

i.e. $\{(x_1, z_1), 0.2\}$

Hence the grade membership of the pair (x_1, z_1) is 0.2.

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Similarly, find all the grade membership of the pairs

$$(x_1, z_2), (x_1, z_3), (x_2, z_1), (x_2, z_2), (x_2, z_3)$$

The final result is

$$R_1 \circ R_2 =$$

x	z	z_1	z_2	z_3
x_1		0.1	0.3	0
x_2		0.8	1	0.3

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Consider fuzzy relations:

$$R = \begin{matrix} & y_1 & y_2 \\ \begin{matrix} x_1 \\ x_2 \end{matrix} & \begin{bmatrix} 0.7 & 0.6 \\ 0.8 & 0.3 \end{bmatrix} \end{matrix}, \quad S = \begin{matrix} & z_1 & z_2 & z_3 \\ \begin{matrix} y_1 \\ y_2 \end{matrix} & \begin{bmatrix} 0.8 & 0.5 & 0.4 \\ 0.1 & 0.6 & 0.7 \end{bmatrix} \end{matrix}.$$

Find the relation $T = R \circ S$ using max-min and max-product composition.

$$\begin{aligned} T &= R \circ S \\ \mu_T(x_1, z_1) &= \max[\min(0.7, 0.8), \min(0.6, 0.1)] \\ &= \max[0.7, 0.1] \\ &= 0.7, \\ \mu_T(x_1, z_2) &= \max[\min(0.7, 0.5), \min(0.6, 0.6)] \\ &= \max[0.5, 0.6] \\ &= 0.6, \end{aligned}$$

$$T = \begin{matrix} & z_1 & z_2 & z_3 \\ \begin{matrix} x_1 \\ x_2 \end{matrix} & \begin{bmatrix} 0.7 & 0.6 & 0.7 \\ 0.8 & 0.5 & 0.4 \end{bmatrix} \end{matrix}.$$

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Max-Product Composition

$$\begin{aligned}\mu_T(x_1, z_1) &= \max[\min(0.7 \times 0.8), \min(0.6 \times 0.1)] \\ &= \max[0.56, 0.06] \\ &= 0.56,\end{aligned}$$

$$\begin{aligned}\mu_T(x_1, z_2) &= \max[\min(0.7 \times 0.5), \min(0.6 \times 0.6)] \\ &= \max[0.35, 0.36] \\ &= 0.36,\end{aligned}$$

$$\begin{aligned}\mu_T(x_1, z_3) &= \max[\min(0.7 \times 0.4), \min(0.5 \times 0.7)] \\ &= \max[0.28, 0.35] \\ &= 0.35,\end{aligned}$$

$$T_{\sim} = \begin{bmatrix} 0.56 & 0.36 & 0.35 \\ 0.64 & 0.40 & 0.32 \end{bmatrix}.$$

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In the field of computer networking there is an imprecise relationship between the level of use of a network communication bandwidth and the latency experienced in peer-to-peer communication. Let \tilde{X} be a fuzzy set of use levels (in terms of the percentage of full bandwidth used) and \tilde{Y} be a fuzzy set of latencies (in milliseconds) with the following membership function:

$$\tilde{X} = \left\{ \frac{0.2}{10} + \frac{0.5}{20} + \frac{0.8}{40} + \frac{1.0}{60} + \frac{0.6}{80} + \frac{0.1}{100} \right\},$$

$$\tilde{Y} = \left\{ \frac{0.3}{0.5} + \frac{0.6}{1} + \frac{0.9}{1.5} + \frac{1.0}{4} + \frac{0.6}{8} + \frac{0.3}{20} \right\}.$$

(a) Find the Cartesian product represented by the relation $\tilde{R} = \tilde{X} \times \tilde{Y}$.

Now, suppose we have second fuzzy set of bandwidth usage given by

$$\tilde{X} = \left\{ \frac{0.3}{10} + \frac{0.6}{20} + \frac{0.7}{40} + \frac{0.9}{60} + \frac{1}{80} + \frac{0.5}{100} \right\}.$$

(b) Find $\tilde{S} = \tilde{Z} \underset{\sim 1 \times 6}{\circ} \underset{\sim 6 \times 6}{R}$ using (1) Max-min composition and (2) Using max-product composition.

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