

# Fuzzy Sets

→ The sets that we have studied so far are classical sets termed as crisp sets.

→ In a crisp set  $A$ , an element of a given universal set  $X$  is either a member of  $A$  or it is not a member of  $A$  which can be defined as

$$\mu_A : X \rightarrow \{0, 1\}$$

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

Ex:- Let us consider a set of brilliant students of a class.

Criteria: A score greater than or equal to 75 in the previous examination.

∴ The set will include all those students who have scored 75% or more, whereas a student who has scored 74.9% will be excluded from the set and will be considered as not brilliant.

→ Terms such as brilliant, slow, fast, low, high, cold, hot, etc. are vague terms and crisp sets are not suitable for defining vague terms or concepts.

→ We can overcome this problem in fuzzy sets by associating a grade of membership with every  $x$  of set  $A$ .

• Def:- A fuzzy set  $A$  for a given universal set  $X$  is a set of ordered pairs

$$A = \{ (x, \mu_A(x)) : x \in X, \mu_A(x) : X \rightarrow [0, 1] \}$$

where  $\mu_A(x) \rightarrow$  membership function.

assigns each element of  $X$  a real no. in  $[0, 1]$

called the degree of membership.

Note: The membership function can be defined as per the suitability of the concept.

Ex:- In a certain class, based on the percentage of marks in the final exam, we can define a fuzzy set of brilliant students (B) as follows:

$$\mu_B(x) = \begin{cases} 1 & \text{if } x \geq 75 \\ \frac{x}{75} & \text{if } x < 75 \end{cases}$$

Therefore, the membership degree can be computed as:

<u>Student</u>	<u>Marks</u>	<u>Membership Degree</u>
Stud 1	79	1
Stud 2	74	0.99
Stud 3	50	0.67

→ The grades of membership show the belongingness of the element to the set.

## (I) Operations on Fuzzy Sets

Let  $A$  and  $B$  be two fuzzy sets with respect to a universal set  $X$ . Then, the standard union and intersection of  $A$  and  $B$  are defined as follows:

$$(a) \quad A \cup B = \{ (x, \mu_{A \cup B}(x)) : x \in X,$$

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

$$(b) \quad A \cap B = \{ (x, \mu_{A \cap B}(x)) : x \in X,$$

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

$$(c) \quad \bar{A} = \{ (x, \mu_{\bar{A}}(x)) : x \in X, \mu_{\bar{A}}(x) = 1 - \mu_A(x) \}$$

Ex:- Let  $A$  and  $B$  be two fuzzy sets defined on a set  $X = \{a, b, c, d\}$  and

$$A = \{ (a, 0.2), (b, 0.4), (c, 0.3), (d, 0.7) \}$$

$$B = \{ (a, 0.5), (b, 0.3), (c, 0.4), (d, 0.5) \}$$

$$(a) \bar{A} = \{(a, 0.8), (b, 0.6), (c, 0.7), (d, 0.3)\}$$

$$(b) A \cup B = \{(a, 0.5), (b, 0.4), (c, 0.4), (d, 0.7)\}$$

$$(c) A \cap B = \{(a, 0.2), (b, 0.3), (c, 0.3), (d, 0.5)\}$$

## (II) $\alpha$ -cut and strong $\alpha$ -cut

Def: Given a fuzzy set  $A$ , defined on  $X$  and any number  $\alpha \in [0, 1]$ ,

$\alpha$ -cut and strong  $\alpha$ -cut are the crisp sets defined as follows:

$$\alpha_A \text{ or } A_\alpha = \{x : \mu_A(x) \geq \alpha\}$$

$$\alpha^+_A \text{ or } A^+_\alpha = \{x : \mu_A(x) > \alpha\}$$

Ex:- Let  $A$  be a fuzzy set on a set  $X$

$X = \{10, 20, 30, 40, 50\}$  whose membership function is defined as  $\mu_A(x) = \frac{x}{x+10}$

Find  $A_\alpha$  for  $\alpha = 0.6$ .

$$\therefore \text{Fuzzy Set } A = \{ (10, 0.5), (20, 0.67), (30, 0.75), (40, 0.80), (50, 0.83) \}$$

$$\text{Now, } A_{0.6} = \{ 20, 30, 40, 50 \}$$

→ Properties of  $\alpha$ -cuts and strong  $\alpha$ -cuts

For a fuzzy set  $A$  and  $\alpha_1, \alpha_2 \in [0, 1]$

$$(i) \quad \alpha_1 < \alpha_2 \Rightarrow A_{\alpha_2} \subseteq A_{\alpha_1}$$

$$(ii) \quad \alpha_1 < \alpha_2 \Rightarrow A_{\alpha_2}^+ \subseteq A_{\alpha_1}^+$$

(III) Some more terms

Given a fuzzy set  $A$ , defined on  $X$

(a) Support of  $A$ : It is the crisp set that contains all the elements of  $X$  that have a non-zero grade of membership in  $A$ .

Denoted as  $\boxed{\text{Supp}(A)}$

Note: It is a strong  $\alpha$ -cut of  $A$  for  $\alpha=0$

$$\boxed{\text{Supp}(A) = A_0^+}$$

The element  $x \in X$  at which  $\mu_A(x) = 0.5$  is called the crossover point.

Ex:- Let  $X = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10\}$

$$\begin{aligned}\text{Fuzzy Set } A &= \{(1, 0), (2, 0), (3, 0.2), (4, 0.5), (5, 0.3), \\ &\quad (6, 0.4), (7, 0), (8, 0), (9, 0), (10, 0)\} \\ &= \{(3, 0.2), (4, 0.5), (5, 0.3), (6, 0.4)\}\end{aligned}$$

where  $\mu_A(x) > 0$

$$\text{Supp}(A) = \{x \in X \mid \mu_A(x) > 0\}$$

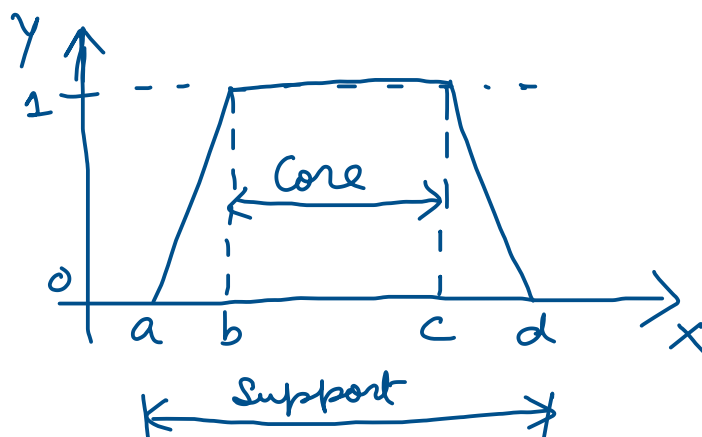
$$\therefore \text{Supp}(A) = \{3, 4, 5, 6\}$$

$$\boxed{x=4} \rightarrow \text{Crossover point}$$

(b) The core of a fuzzy set is the  $\alpha$ -cut of  $A$  for  $\alpha=1$ .

Denoted as  $\boxed{\text{Core}(A) = A_1}$

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



(c) The height of a fuzzy set is the highest grade of membership of any element in the set and it is denoted by  $\boxed{h(A)}$ .

$$\boxed{h(A) = \max_{x \in X} \mu_A(x)}$$

(d) A fuzzy set is called Normal if

$$\boxed{h(A) = 1}$$

and subnormal if

$$\boxed{h(A) < 1}$$



#### (IV) Fuzzy Sets in decision making

Ex:- Rahul is looking for a two-bedroom apartment having low rent (around Rs. 30,000) and is near (within 1km distance) his/her office, then the decision making can be done through fuzzy set theory.

$$\mu_A(x) = \begin{cases} 1 & \text{for } x \leq 30,000 \\ \frac{30,000}{x} & \text{for } x > 30,000 \end{cases}$$

$$\mu_B(x) = \begin{cases} 1 & \text{for } x \leq 1 \\ \frac{1}{x} & \text{for } x > 1 \end{cases}$$

→ Therefore, each apartment in the area will have two grades of membership.

→ We need to compute  $\mu_{A \cap B}(x)$  for every flat  $(x)$ .

→ The highest grade of membership shows the best choice.

→ Applications of fuzzy set includes AI, ML, Knowledge acquisition, decision analysis etc.