## SCS5107 Computational Intelligence Unit V

## **Fuzzy Systems**

#### CLASSICAL PROPOSITION LOGIC

As in our ordinary informal language,  ${}^{a}$ **sentence** ${}^{o}$  is used in the logic. Especially, a sentence having only  ${}^{a}$ true (1) ${}^{o}$  or  ${}^{a}$ false (0) ${}^{o}$  as its truth value is called  ${}^{a}$ **proposition** ${}^{o}$ .

$$2 + 4 = 7$$
 (false)

For every x, if  $f(x) = \sin x$ , then  $f \boxplus (x) = \cos x$ . (true) .It rains now. (true or false depending whether it rains or not)

Connectives - combine prepositional variables .

Negation, Conjunction, Disjunction, Implication

What does it mean if p implies q? Example: p = sky is overcast, q = sun not visible,  $p \rightarrow q$ 

p	$oldsymbol{q}$	eg p	$p \lor q$	$p \wedge q$	p  o q
T	T	$oldsymbol{F}$	T	T	T
$oldsymbol{T}$	$oldsymbol{F}$	$oldsymbol{F}$	T	$oldsymbol{F}$	$oldsymbol{F}$
$oldsymbol{F}$	$\boldsymbol{T}$	T	T	$oldsymbol{F}$	$oldsymbol{T}$
$oldsymbol{F}$	$oldsymbol{F}$	T	$egin{array}{cccccccccccccccccccccccccccccccccccc$	${m F}$	$oldsymbol{T}$

p	$\boldsymbol{q}$	p  o q
T	$\boldsymbol{T}$	$oldsymbol{T}$
T	$oldsymbol{F}$	$oldsymbol{F}$
$oldsymbol{F}$	$\boldsymbol{T}$	$oldsymbol{T}$
$oldsymbol{F}$	$oldsymbol{F}$	T

#### LOGIC FUNCTIONS

Logic function : a combination of propositional variables by using connectives Logic formula : Truth values 0 and 1 are logic formulas

- If v is a logic variable, v and v' are a logic formulas
- If a and b represent a logic formulas,

#### TAUTOLOGY AND INFERENCE RULE

Tautology:  $A^{\underline{a}}$  tautology<sup> $\underline{o}$ </sup> is a logic formula whose value is always true regardless of its logic variables.  $A^{\underline{a}}$  contradiction<sup> $\underline{o}$ </sup> is one which is always false.

a	b	$a \rightarrow b$	$\overline{(a \rightarrow b)}$	$ar{b}$	$\overline{(a \to b)} \to \bar{b}$
1	1	1	0	0	1
1	0	0	1	1	1
0	1	1	0	0	1
0	0	1	0	1	1

#### TAUTOLOGY AND INFERENCE RULE

a	b	(a → b)	$(a \wedge (a \rightarrow b))$	$(a \wedge (a \rightarrow b)) \rightarrow b$
1	1	1	1	1
1	0	0	0	1
0	1	1	0	1
0	0	1	0	1

#### Predicate logic:

"Predicate logic" is a logic which represents a proposition with the predicate and an individual (object)

Objects in predicate logic can represented by variables. Then a predicate proposition can be evaluated for truth if an element of a universal set is instantiated to the variable.

"x is a man"

x="Tom", the proposition becomes "Tom is a man"

<sup>&</sup>quot;Socrates is a man"

<sup>&</sup>quot;Socrates" – object

<sup>&</sup>quot;is a man" – predicate

<sup>&</sup>quot;Two is less than four"

<sup>&</sup>quot;Two", "four" – objects

<sup>&</sup>quot;is less than" - predicate

"x satisfies P" can be denoted P(x) is a man(Tom)

## FUZZY SETS

Definition: let X be a non-empty set and be called the universe of discourse. A fuzzy set ACX is characterized by the membership function

where  $\mu_A(x)$  is a grade (degree) of membership of x in set A.

$$\mu_A: X \rightarrow [0,1]$$

## FUZZY SETS

Definition of fuzzy sets:

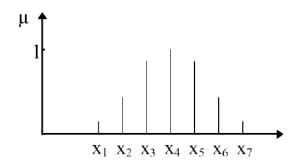
Fuzzy set A can be represented as a set of ordered pairs

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

## FUZZY SETS

Discrete example:

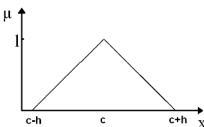
$$\mu_{\!\scriptscriptstyle A} = 0.1/x1 + 0.4/x2 + 0.8/x3 + 1.0/x4 + 0.8/x5 + \\ 0.4/x6 + 0.1/x7$$



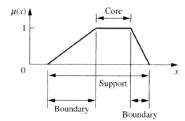
## FUZZY SETS

Continuous example:

$$\mu_{A}(x) = \begin{cases} 1 + \frac{x - c}{h}, & x \in [c - h, c] \\ 1 - \frac{x - c}{h}, & x \in [c, c + h] \\ 0, & \text{otherwise} \end{cases}$$



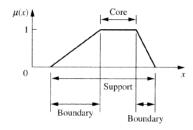
## PROPERTIES OF FUZZY SETS



Support: support of a fuzzy set A is a crisp set that contains all elements of A with non-zero membership grade:

$$\operatorname{supp}(A) = \left\{ x \in X \middle| \mu_A(x) > 0 \right\}$$

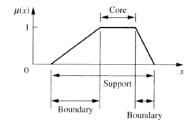
## PROPERTIES OF FUZZY SETS



Core: comprises those elements x of the universe such that  $\mu_{A}(x) = 1$ .

$$core(A) = \left\{ x \in X \middle| \mu_A(x) = 1 \right\}$$

## PROPERTIES OF FUZZY SETS



Boundary: boundaries comprise those elements x of the universe such that  $0 < \mu_A(x) < 1$ 

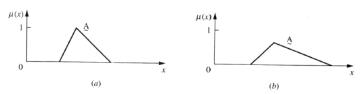
bnd(A) = 
$$\{x \in X | 0 < \mu_A(x) < 1\}$$

## PROPERTIES OF FUZZY SETS

Height: the height of a fuzzy set A id defined

$$hgt(A) = \sup_{x \in X} \mu_A(x)$$

Set A is called normal if hgt(A)=1 and subnormal if hgt(A)<1



Fuzzy sets that are normal (a) and subnormal (b).

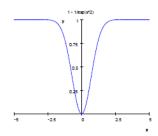
## PROPERTIES OF FUZZY SETS

Question?

Is the fuzzy set defined as

$$\mu_A(x) = 1 - 1/e^{-x^2}$$

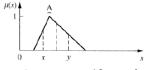
Normal or subnormal?



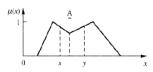
## PROPERTIES OF FUZZY SETS

Convex Fuzzy set: a fuzzy set A is convex iff

$$\forall x, y \in X \text{ and } \forall \lambda \in [0,1]$$
  
 $\mu_A(\lambda x + (1-\lambda)y) \ge \min(\mu_A(x), \mu_A(y))$ 



Convex, normal fuzzy set



Non convex, normal fuzzy set

## OPERATIONS ON FUZZY SETS

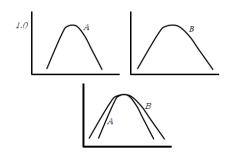
- $\frac{\text{Empty set}}{\mu_{\varnothing}} \equiv 0$
- 2. Basic set (universe)  $\mu_X \equiv 1$
- 3. <u>Identity</u>

$$A = B \iff \mu_A(x) = \mu_B(x) \quad \forall x \in X$$

#### OPERATIONS ON FUZZY SETS

4. Subset

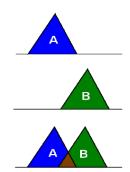
$$A \subset B \iff \mu_A(x) \le \mu_B(x) \quad \forall x \in X$$



#### OPERATIONS ON FUZZY SETS

5. <u>Union</u>

$$U[\mu_A, \mu_B] = \max[\mu_A, \mu_B],$$
  
$$\forall x \in X : \mu_{A \cup B}(x) = \max(\mu_A(x), \mu_B(x))$$



### OPERATIONS ON FUZZY SETS

#### 6. Intersection

Axioms for intersection function  $I:[0,1]\times[0,1]\to[0,1]$   $\mu_{A\cap B}(x)=I[\mu_A(x),\ \mu_B(x)]$ 

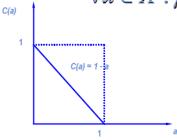
- I(1, 1) = 1, I(1, 0) = 0, I(0, 1) = 0, I(0, 0) = 0
- I(a, b) = I(b, a), Commutativity.
- If  $a \le a'$  and  $b \le b'$ ,  $I(a, b) \le I(a', b')$ , monotonicity.
- I(I(a, b), c) = I(a, I(b, c)), Associativity.
- *I* is a continuous function.
- I(a, a) = a, idempotency.

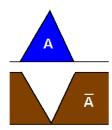
## 7. Complement

Standard complement function:

$$C(a) = 1 - a$$

$$\forall x \in X : \mu_{\overline{A}}(x) = 1 - \mu_A(x)$$





## Properties of fuzzy operations:

(1) Involution  $\overline{\overline{A}} = A$ 

(2) Commutativity  $A \cup B = B \cup A$ 

 $A \cap B = B \cap A$ 

(3) Associativity  $(A \cup B) \cup C = A \cup (B \cup C)$ 

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

(4) Distributivity  $A \cap (B \cup C) = (A \cap B) \cup (A \cap C)$ 

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

(5) Idempotency  $A \cup A = A$ 

 $A \cap A = A$ 

(6) Absorption  $A \cup (A \cap B) = A$ 

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

(7) Absorption by X and  $\emptyset$   $A \cup X = X$ 

 $A \cap \emptyset = \emptyset$ 

(8) Identity  $A \cup \emptyset = A$ 

 $A \cap X = A$ 

(9) De Morgan's law  $\overline{A \cap B} = \overline{A} \cup \overline{B}$ 

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

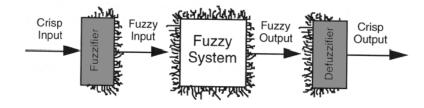
(10) Equivalence formula

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

(11) Symmetrical difference formula

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

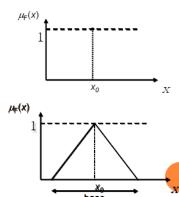
## **Fuzzy System**

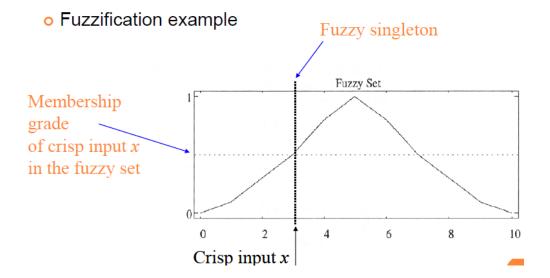


- \* Fuzzify crisp inputs to get the fuzzy inputs
- \* Defuzzify the fuzzy outputs to get crisp outputs

## **FUZZIFICATION**

- Process of making a crisp quantity fuzzy
- If it is assumed that input data do not contain noise of vagueness, a fuzzy singleton can be used
- If the data are vague or perturbed by noise, they should be converted into a fuzzy number





#### BUILD A FUZZY CONTROLLER

#### 3 Steps

- 1. Pick the linguistic variables
  - Example: Let temperature (X) be input and motor speed
     (Y) be output
- 2. Pick the fuzzy sets
  - · Define fuzzy subsets of the X and Y
- 3. Pick the fuzzy rules
  - · Associate output to the input

# EXAMPLE: BUILD A FUZZY CONTROLLER

## Goal: Design a motor speed controller for air conditioner

## Step 1: assign input and output variables

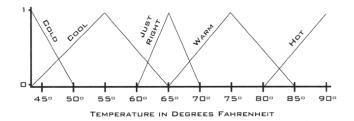
- Let X be the temperature in Fahrenheit
- o Let Y be the motor speed of the air conditioner

#### Step 2: Pick fuzzy sets

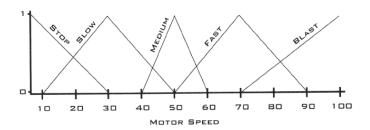
Define linguistic terms of the linguistic variables temperature (X) and motor speed (Y) and associate them with fuzzy sets

- For example, 5 linguistic terms / fuzzy sets on X
  - oCold, Cool, Just Right, Warm, and Hot
- Say 5 linguistic terms / fuzzy sets on Y
  - Stop, Slow, Medium, Fast, and Blast

## Input Fuzzy sets

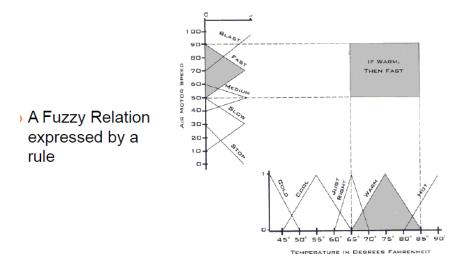


## Output Fuzzy sets

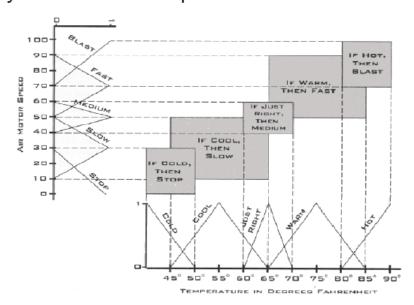


<u>Step 3:</u> Assign a motor speed set to each temperature set

- If temperature is *cold* then motor speed is *stop*
- If temperature is **cool** then motor speed is **slow**
- If temperature is <u>just right</u> then motor speed is <u>mediur</u>
- If temperature is warm then motor speed is fast
- If temperature is **hot** then motor speed is **blast**



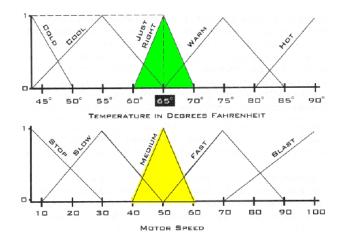
## A Fuzzy controller with 5 patches



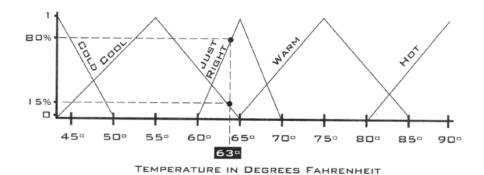
In Mamdani and Larsen model, which rule "fires" or activates at which time?

- They all fire all the time
- They fire in parallel
  - o All rules fire to some degree
  - o Most fire to zero degree
- The result is a union of fuzzy results from each rule

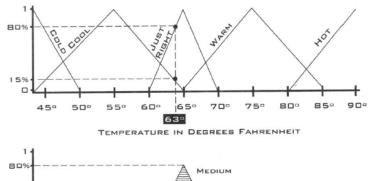
## IF TEMPERATURE IS $\underline{JUST\ RIGHT}$ THEN MOTOR SPEED IS $\underline{MEDIUM}$

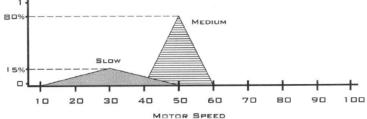


If temperature is <u>cool</u> then motor speed is <u>slow</u>
If temperature is <u>just right</u> then motor speed is <u>medium</u>

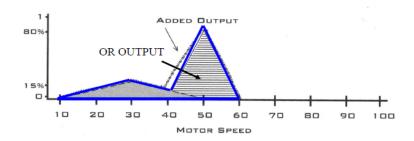


## EXAMPLE: T = 63 DEGREE F.

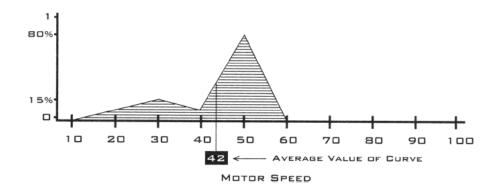




# Summed (MAXed) of the partially fired then-part fuzzy sets



## Defuzzify to find the output motor speed



Question: how to convert a fuzzy set into a crisp value?

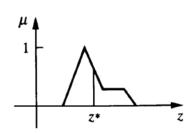
## **DEFUZZIFICATION**



- o Converts a fuzzy set into a crisp output.
- Defuzzification is a process to get a non-fuzzy value th best represents the possibility distribution of an inferred fuzzy control action.
- There is no systematic procedure for choosing a good defuzzification strategy.
- Selection of defuzzification procedure depends on the properties of the application.

## **DEFUZZIFICATION**

- Centroid of the Area: the most prevalent and physically appealing of all the defuzzification methods [Sugeno, 1985; Lee, 1990]
- A disadvantage: computationally intensive

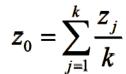


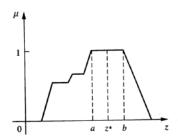
$$z^* = \frac{\int \mu_{\rm C}(z) \cdot z \, dz}{\int \mu_{\rm C}(z) \, dz}$$

where \int denotes an algebraic integration.

• Mean of maximum (MOM)

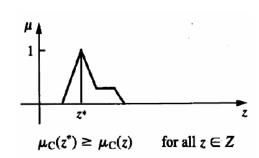
The defuzzified result represents the mean value of all actions, whose membership functions reach the maximum





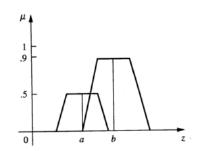
 $z_j$ : control action whose membership functions reach the max k: number of such control actions.

Max-membership principal, also known as height method



## Weighted average method

- Valid for symmetrical output membership functions
- Produces results very close to the COA method
- Less computationally intensive

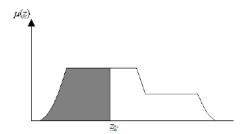


$$z^* = \frac{\sum \mu_{\rm C}(\overline{z}) \cdot \overline{z}}{\sum \mu_{\rm C}(\overline{z})} \qquad z^* = \frac{a(.5) + b(.9)}{.5 + .9}$$

Formed by weighting each functions in the output by its respective maximum membership value

#### Bisector of the Area

 The BOA generates the action (z<sub>0</sub>) which partitions the area into two regions with the same area



$$\int_{\alpha}^{z_0} \mu_C(z) dz = \int_{z_0}^{\beta} \mu_C(z) dz$$

$$\alpha = \min\{z \mid z \in W\}$$

$$\beta = \max\{z \mid z \in W\}$$

## First (or last) of maxima

 Determine the smallest value of the domain with maximized membership degree

