



**DRONACHARYA**  
College of Engineering

INTELLIGENT SYSTEMS (CSE-303-F)

Section C

**Fuzzy Reasoning**

# Contents

- **Bivalent and Multivalent Logics**
- **Linguistic Variables**
- **Fuzzy Sets**
- **Membership Functions**
- **Fuzzy Set Operators**
- **Hedges**

# Contents

- **Fuzzy Logic**
- **Fuzzy Rules**
- **Fuzzy Inference**
- **Fuzzy Expert Systems**
- **Neuro-Fuzzy Systems**

# Bivalent and Multivalent Logics

- **Bivalent (Aristotelian) logic uses two logical values – true and false.**
- **Multivalent logics use many logical values – often in a range of real numbers from 0 to 1.**
- **Important to note the difference between multivalent logic and probability –  $P(A) = 0.5$  means that A may be true or may be false – a logical value of 0.5 means both true and false at the same time.**

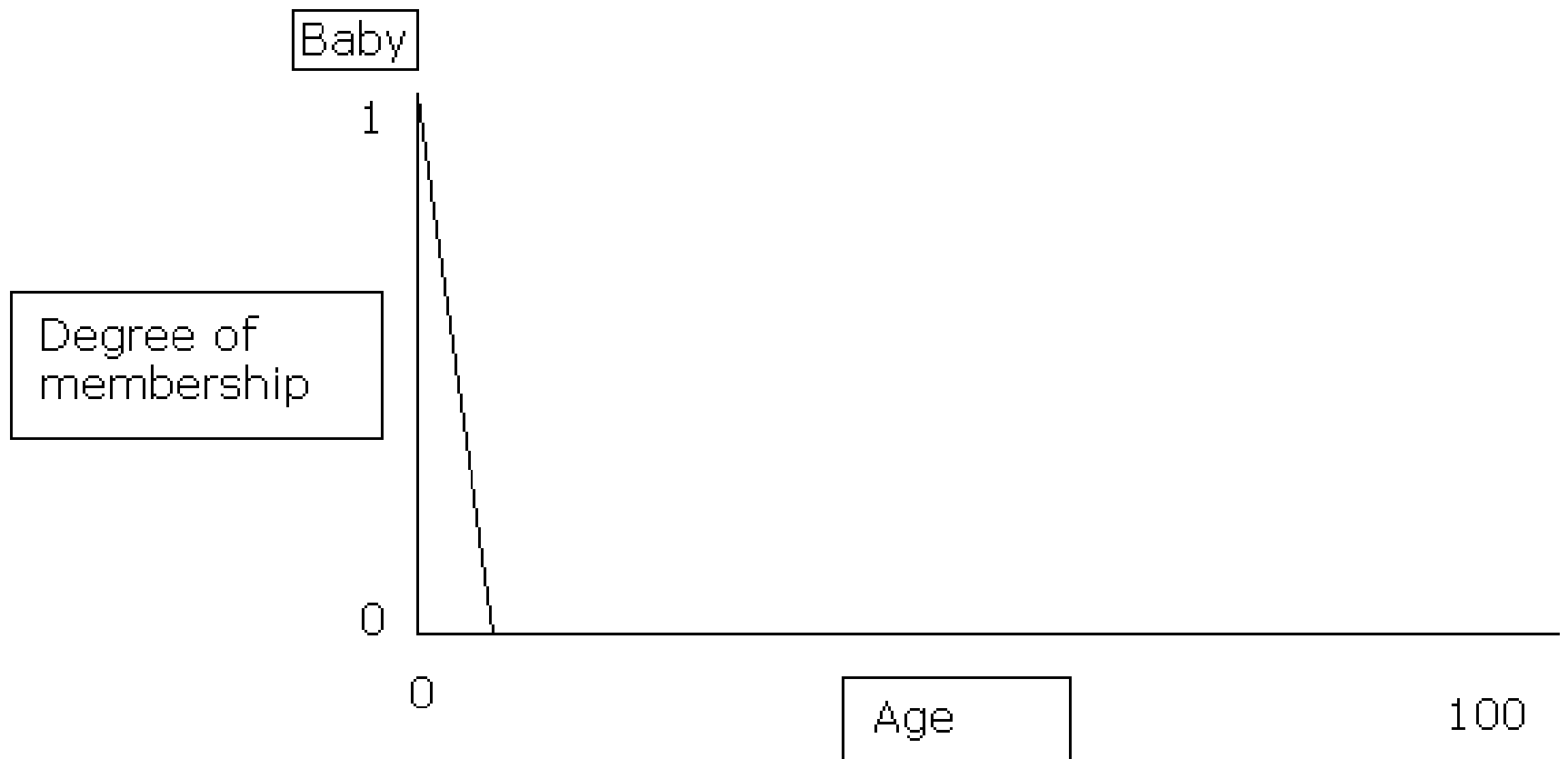
# Linguistic Variables

- **Variables used in fuzzy systems to express qualities such as height, which can take values such as “tall”, “short” or “very tall”.**
- **These values define subsets of the universe of discourse.**

# Fuzzy Sets

- **A crisp set is a set for which each value either is or is not contained in the set.**
- **For a fuzzy set, every value has a membership value, and so is a member to some extent.**
- **The membership value defines the extent to which a variable is a member of a fuzzy set.**
- **The membership value is from 0 (not at all a member of the set) to 1.**

# Membership Functions



# Crisp Set Operators

- Not  $A$  – the complement of  $A$ , which contains the elements which are not contained in  $A$ .
- $A \cap B$  – the intersection of  $A$  and  $B$ , which contains those elements which are contained in both  $A$  and  $B$ .
- $A \cup B$  – the union of  $A$  and  $B$  which contains all the elements of  $A$  and all the elements of  $B$ .
- Fuzzy sets use the same operators, but the operators have different meanings.



# Fuzzy Set Operators

- **Fuzzy set operators can be defined by their membership functions**
  - $M_{\neg A}(x) = 1 - M_A(x)$
  - $M_{A \cap B}(x) = \text{MIN}(M_A(x), M_B(x))$
  - $M_{A \cup B}(x) = \text{MAX}(M_A(x), M_B(x))$
- **We can also define containment (subset operator):**
  - $B \subset A \text{ iff } \forall x (M_B(x) \leq M_A(x))$

# Hedges

- A hedge is a qualifier such as “very”, “quite”, “somewhat” or “extremely”.
- When a hedge is applied to a fuzzy set it creates a new fuzzy set.
- Mathematic functions are usually used to define the effect of a hedge.
- For example, “Very” might be defined as:
  - $M_{VA}(x) = (M_A(x))^2$

# Fuzzy Logic (1)

- A nonmonotonic logical system that applies to fuzzy variables.
- We use connectives defined as:
  - $A \vee B \equiv \text{MAX} (A, B)$
  - $A \wedge B \equiv \text{MIN} (A, B)$
  - $\neg A \equiv 1 - A$
- We can also define truth tables:

A	B	$A \vee B$
0	0	0
0	0.5	0.5
0	1	1
0.5	0	0.5
0.5	0.5	0.5
0.5	1	1
1	0	1
1	0.5	1
1	1	1

# Fuzzy Inference (1)

- Inference is harder to manage.

- Since:

$$A \rightarrow B \equiv \neg A \vee B$$

- Hence, we might define fuzzy inference as:

$$A \rightarrow B \equiv \text{MAX} ((1 - A), B)$$

- This gives the unintuitive truth table shown on the right.

- This gives us  $0.5 \rightarrow 0 = 0.5$ , where we would expect  $0.5 \rightarrow 0 = 0$

A	B	$A \rightarrow B$
0	0	1
0	0.5	1
0	1	1
0.5	0	0.5
0.5	0.5	0.5
0.5	1	1
1	0	0
1	0.5	0.5
1	1	1

## Fuzzy Inference (2)

- An alternative is Gödel implication, which is defined as:

$$A \rightarrow B \equiv (A \leq B) \vee B$$

- This gives a more intuitive truth table.

A	B	A → B
0	0	1
0	0.5	1
0	1	1
0.5	0	0
0.5	0.5	1
0.5	1	1
1	0	0
1	0.5	0.5
1	1	1

# Fuzzy Inference (3)

- Mamdani inference derives a single crisp value by applying fuzzy rules to a set of crisp input values.

Step 1: Fuzzify the inputs.

Step 2: Apply the inputs to the antecedents of the fuzzy rules to obtain a set of fuzzy outputs.

Step 3: Convert the fuzzy outputs to a single crisp value using defuzzification.

# Fuzzy Rules

- A fuzzy rule takes the following form:  
IF A op x then B = y
- op is an operator such as  $>$ ,  $<$  or  $=$ .
- For example:  
IF temperature  $>$  50 then fan speed = fast  
IF height = tall then trouser length = long  
IF study time = short then grades = poor

# Fuzzy Expert Systems

- A fuzzy expert system is built by creating a set of fuzzy rules, and applying fuzzy inference.
- In many ways this is more appropriate than standard expert systems since expert knowledge is not usually black and white but has elements of grey.
- The first stage in building a fuzzy expert system is choosing suitable linguistic variables.
- Rules are then generated based on the expert's knowledge, using the linguistic variables.



# Neuro-Fuzzy Systems

- A fuzzy neural network is usually a feed-forward network with five layers:
  1. Input layer – receives crisp inputs
  2. Fuzzy input membership functions
  3. Fuzzy rules
  4. Fuzzy output membership functions
  5. Output layer – outputs crisp values

