

Fuzzy Logic and Systems

Why Fuzzy

- * Based on Intuition and judgment
- * No need for a mathematical model
- * Provides a smooth transition between members and nonmembers
- * Relatively simple, fast and adaptive
- * Less sensitive to system fluctuations
- * Can implement design objectives, difficult to express mathematically, in linguistic or descriptive rules.

Applications Domain

- * Fuzzy Logic
- * Fuzzy Control
 - Neuro-Fuzzy System
 - Intelligent Control
 - Hybrid Control
- * Fuzzy Pattern Recognition
- * Fuzzy Modeling

Some Interesting Applications

- Ride smoothness control
- Camcorder auto-focus and jiggle control
- Braking systems
- Copier quality control
- Rice cooker temperature control
- High performance drives
- Air-conditioning systems

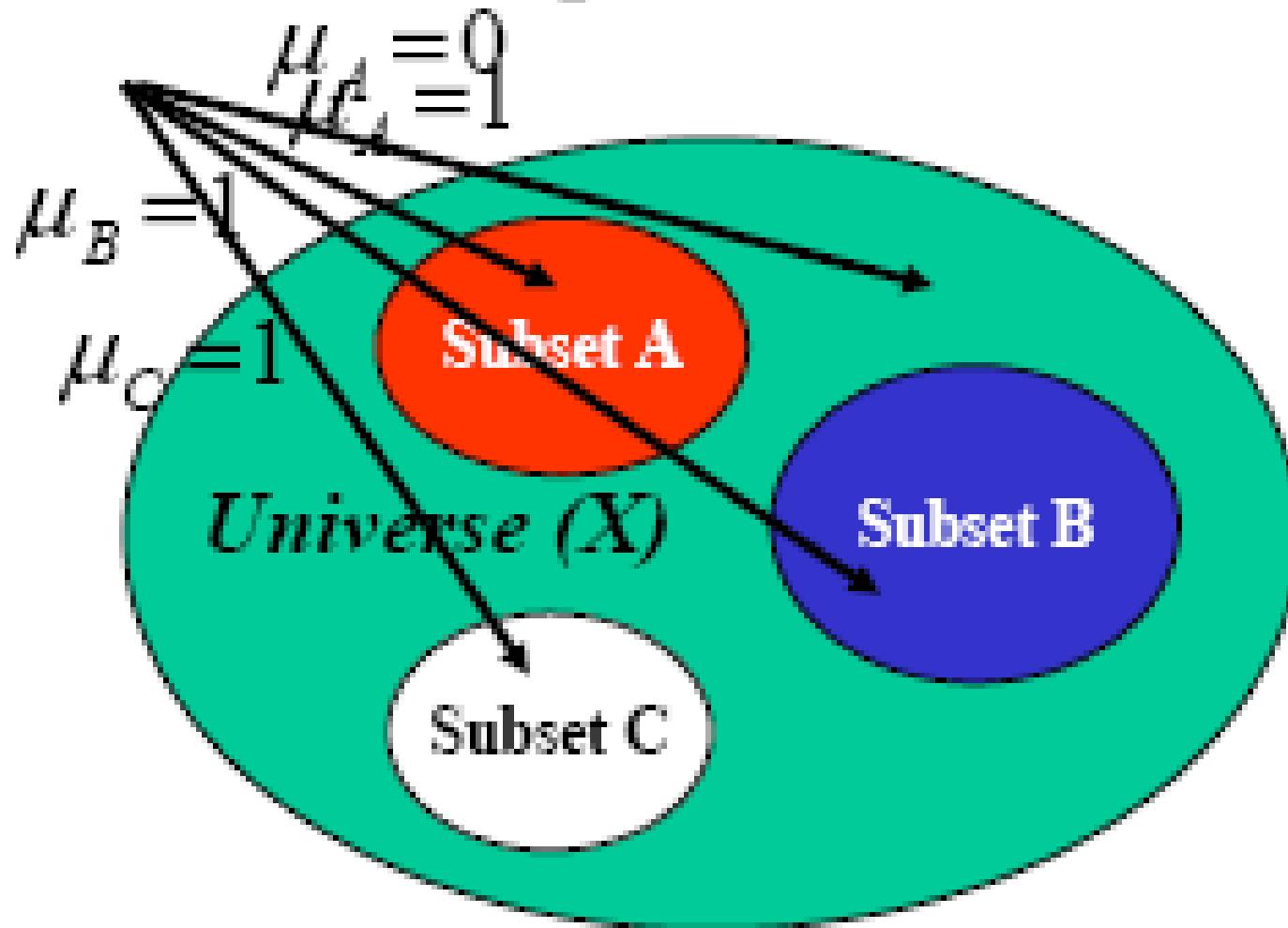
Crisp Set

Conventional or *crisp* sets are binary. An element either belongs to the set or doesn't.

{True, false}

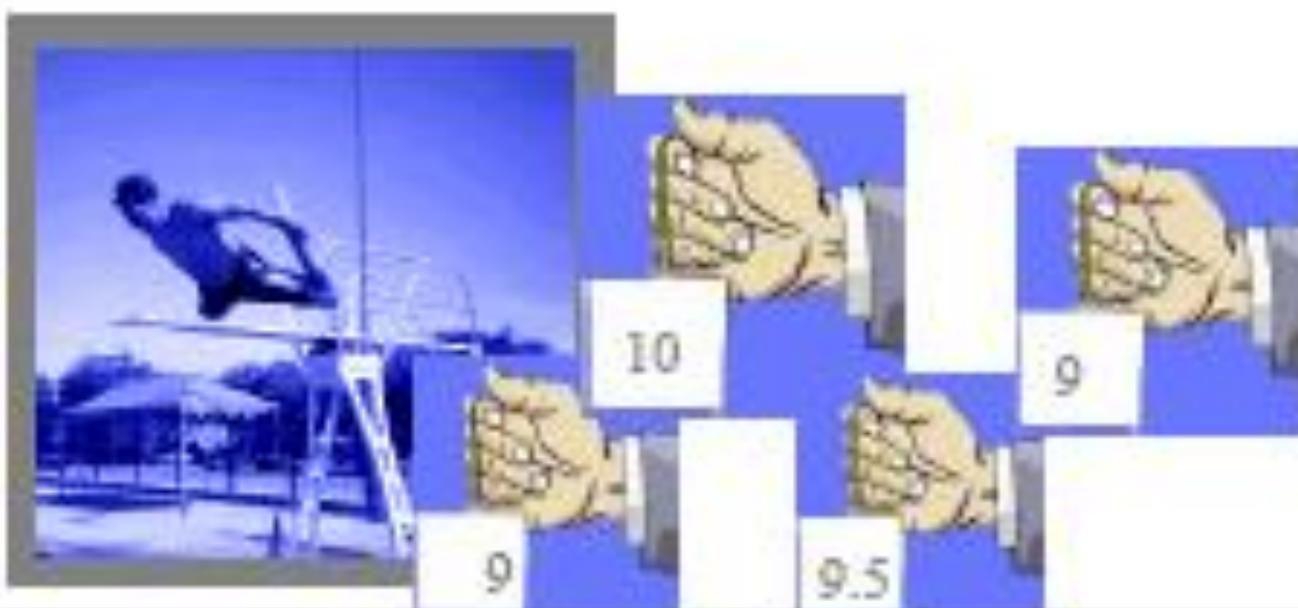
{1, 0}

Crisp Set/Subset



Fuzzy Linguistic Variables

- Examples of fuzzy measures include *close*, *heavy*, *light*, *big*, *small*, *smart*, *fast*, *slow*, *hot*, *cold*, *tall* and *short*.



Fuzzy Indicators

- Can you distinguish between American and French person?
- Some Rules:
 - *If* speaks English *then* American
 - *If* speaks French *then* French
 - *If* loves perfume *then* French
 - *If* loves outdoors *then* American
 - *If* good cook *then* French
 - *If* plays baseball *then* American

Fuzzy Indicators

- Rules may give contradictory indicators
{good cook, loves outdoors, speaks French}
- The right answer is a question of a degree of association
- Fuzzy logic resolves these conflicting indicators
 - Membership of the person in the French set is 0.9
 - Membership of the person in the American set is 0.1

Fuzzy Versus Probability

- Fuzzy \neq Probability
- Probability deals with uncertainty and likelihood
- Fuzzy logic deals with ambiguity and vagueness

Fuzzy Versus Probability

- Fuzzy \neq Probability
- Example #1
 - Billy has ten toes. The probability Billy has nine toes is zero. The fuzzy membership of Billy in the set of people with nine toes, however, is nonzero.

(c) Professor Maxx)



Fuzzy Versus Probability

Example #2

- A bottle of liquid has a probability of $\frac{1}{2}$ of being rat poison and $\frac{1}{2}$ of being pure water.
- A second bottle's contents, in the fuzzy set of liquids containing lots of rat poison, is $\frac{1}{2}$.
- The meaning of $\frac{1}{2}$ for the two bottles clearly differs significantly and would impact your choice should you be dying of thirst.
- 50% probability means 50% chance that the water is clean.
- 50% fuzzy membership means that the water has poison.

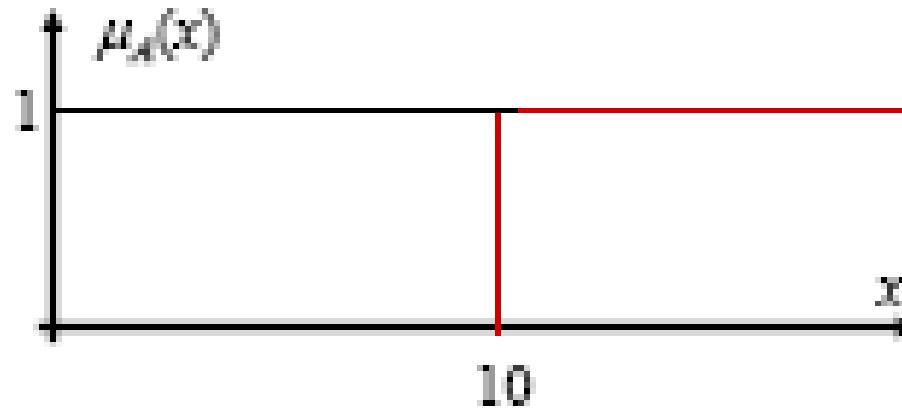


(See Results)

Crisp Membership Functions

- * Crisp membership functions (μ) are either one or zero.
- * e.g, Numbers greater than 10.

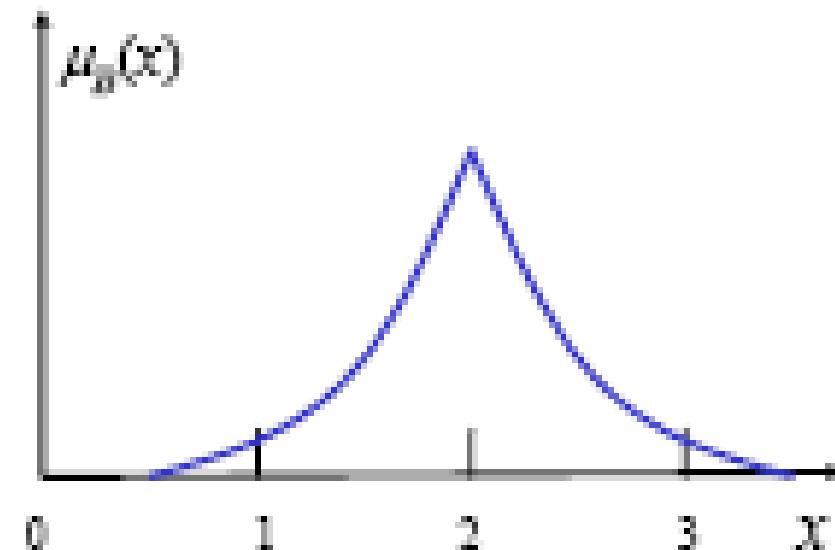
$$A = \{x \mid x > 10\}$$



Fuzzy Membership Functions

- The set, B , of numbers near to 2 can be represented by a membership function

$$\mu_B(x) = e^{-|x-2|}$$



Fuzzy Subsets

- * A fuzzy set, A , is said to be a subset of B if

$$\mu_A(x) \leq \mu_B(x)$$

- * e.g. $B = \text{far}$ and $A = \text{very far}$.
- * For example...

$$\mu_A(x) = \mu_B^2(x)$$

Fuzzy Sets



Tall

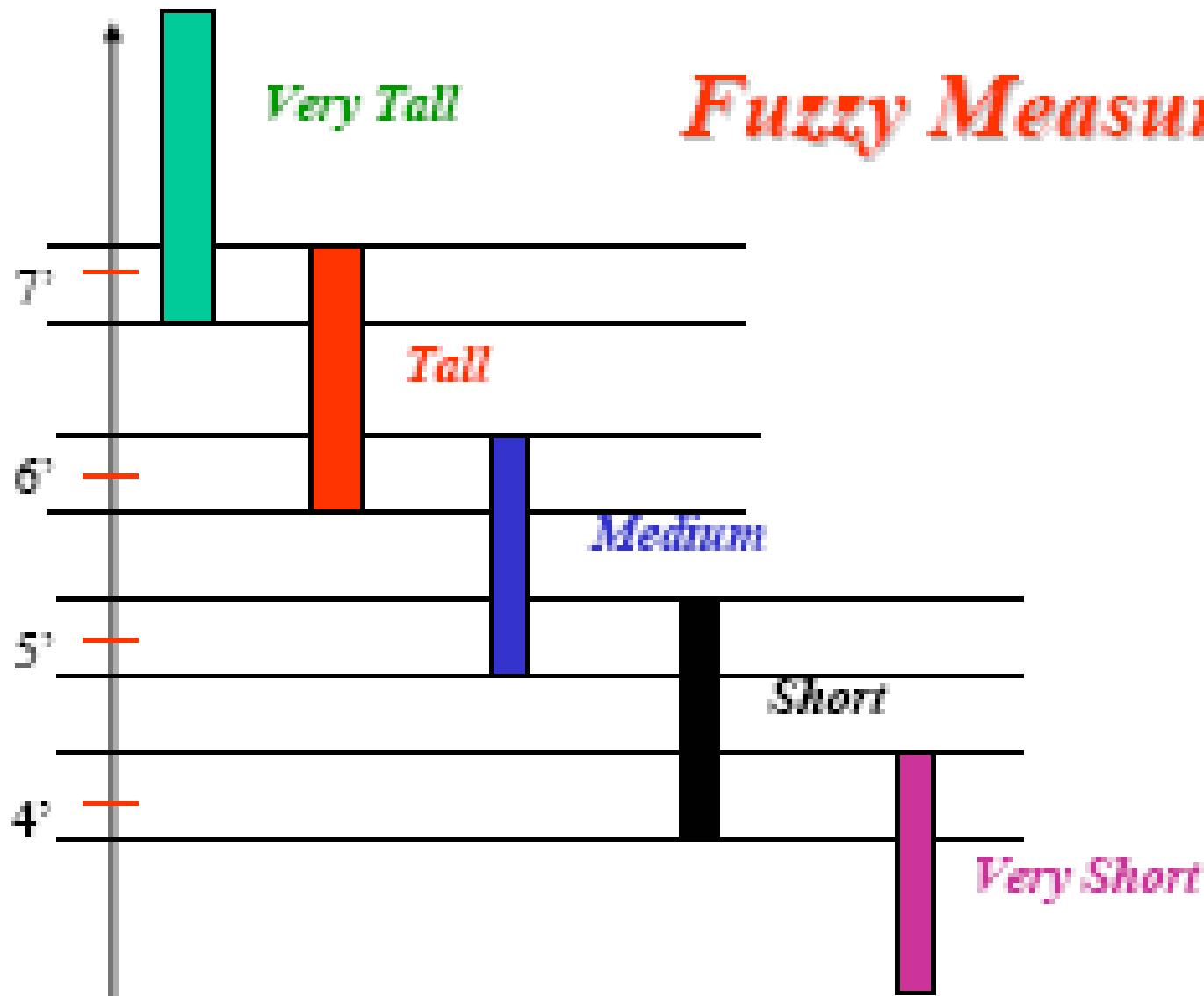


Tall or Short?

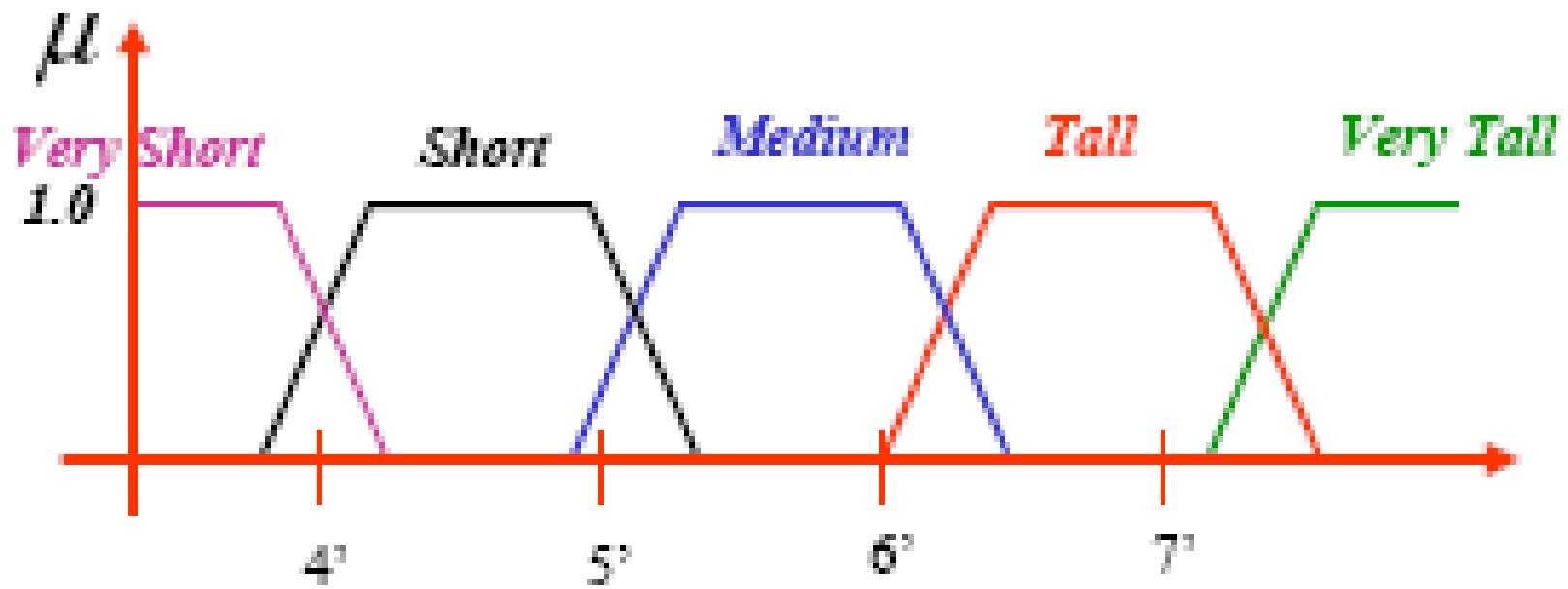


Short

Fuzzy Measures

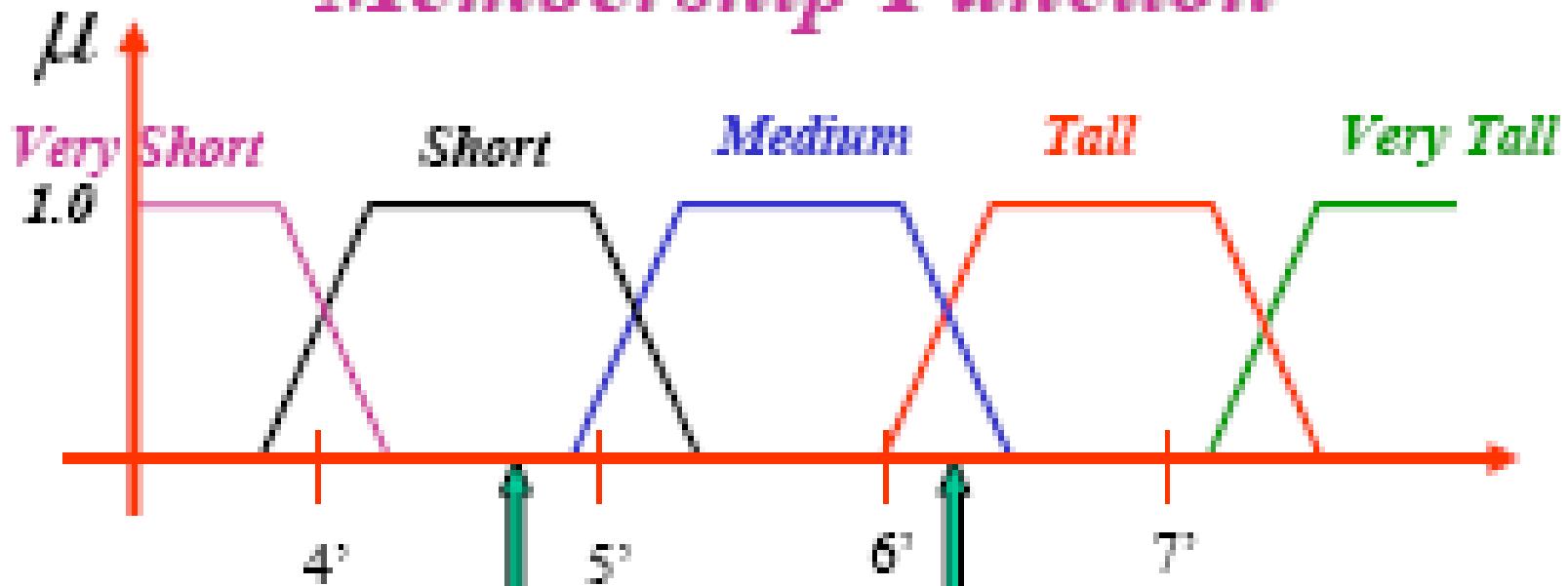


Membership Function



$$\mu = [\mu_{vs}, \mu_s, \mu_m, \mu_t, \mu_{vt}]$$

Membership Function



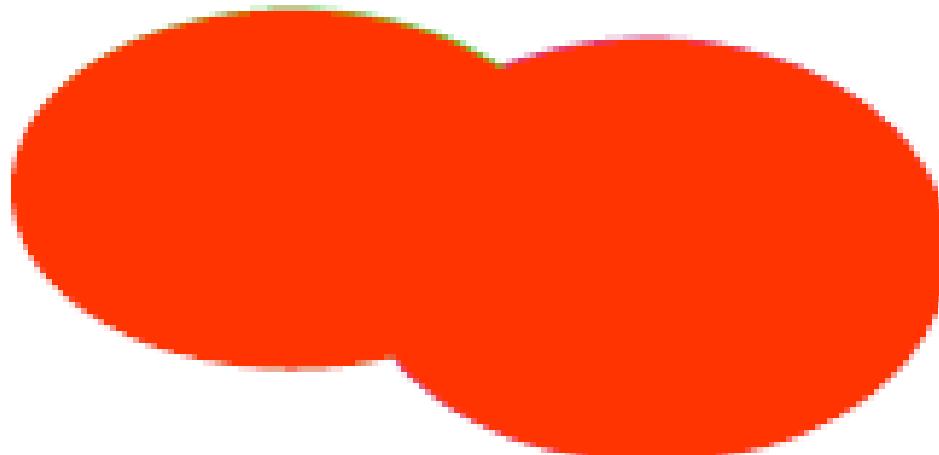
Short Medium Tall

$$\mu = [0, 1, 0, 0, 0]$$

$$\mu = [0, 0, 0.5, 0.5, 0]$$

Fuzzy Logic Operations

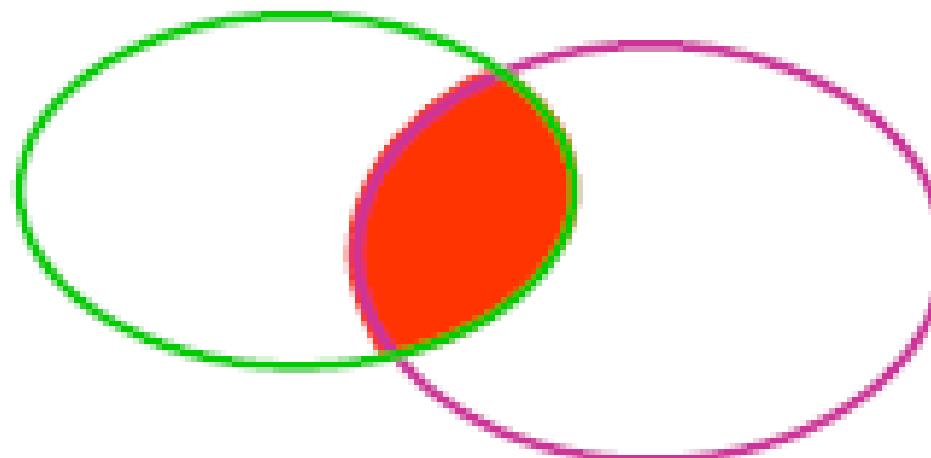
Fuzzy union operation or fuzzy OR



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

Fuzzy Logic Operations

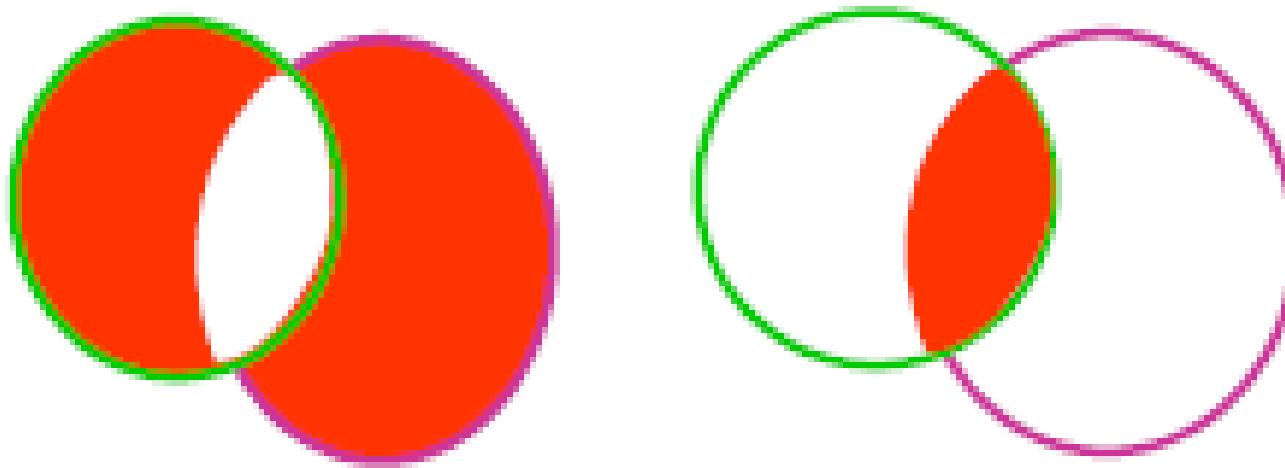
Fuzzy intersection operation or fuzzy AND



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

Fuzzy Logic Operations

Complement operation



$$\mu_{\bar{A}}(x) = 1 - \mu_A(x)$$

Fuzzy Logic Operations

Fuzzy union operation or fuzzy OR

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

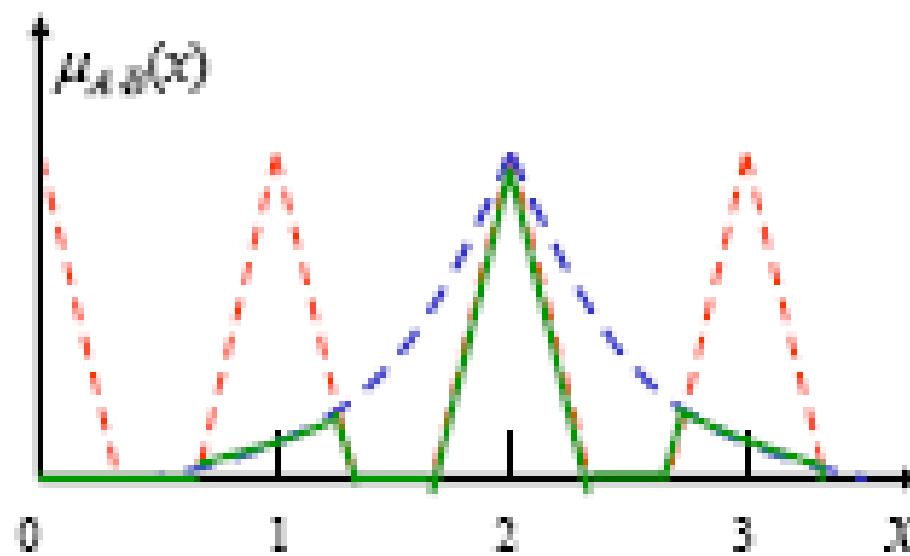
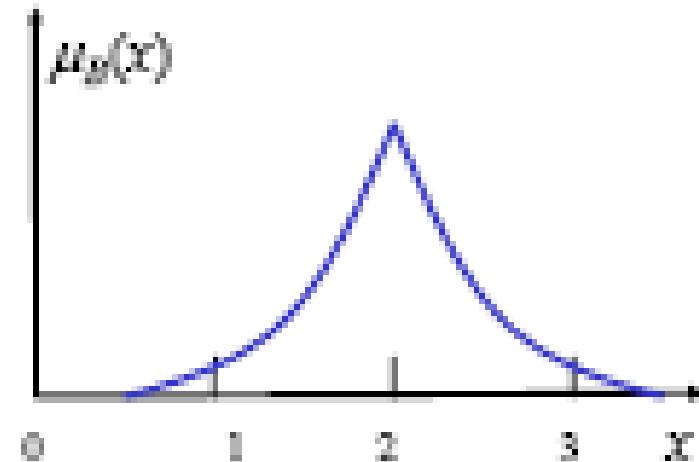
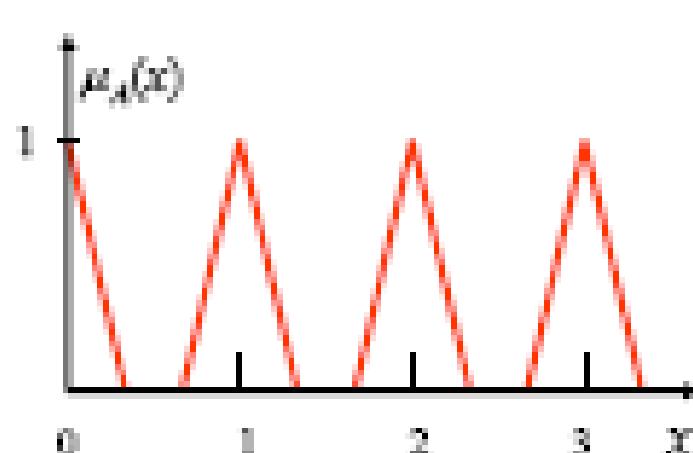
Fuzzy intersection operation or fuzzy AND

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

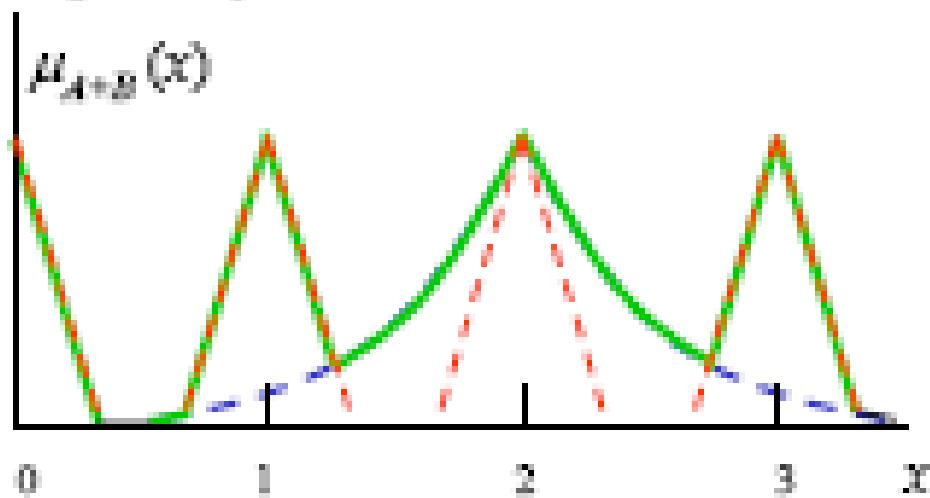
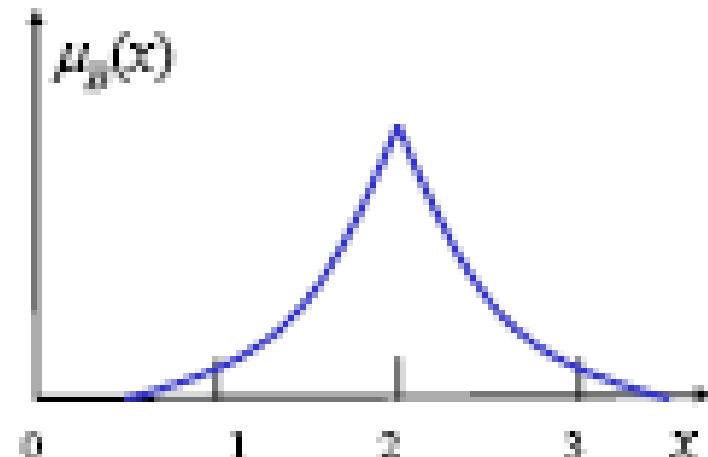
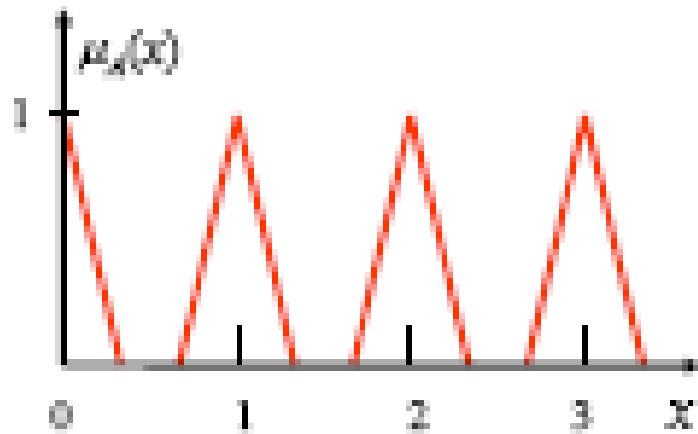
Complement operation

$$\mu_{\bar{A}}(x) = 1 - \mu_A(x)$$

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

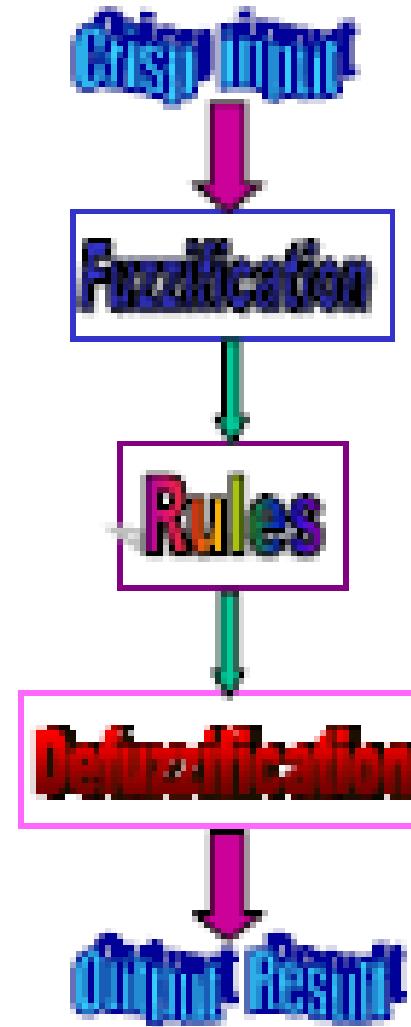


$$\mu_{A+B}(x) = \max [\mu_A(x), \mu_B(x)]$$



Fuzzy System

Basic Components



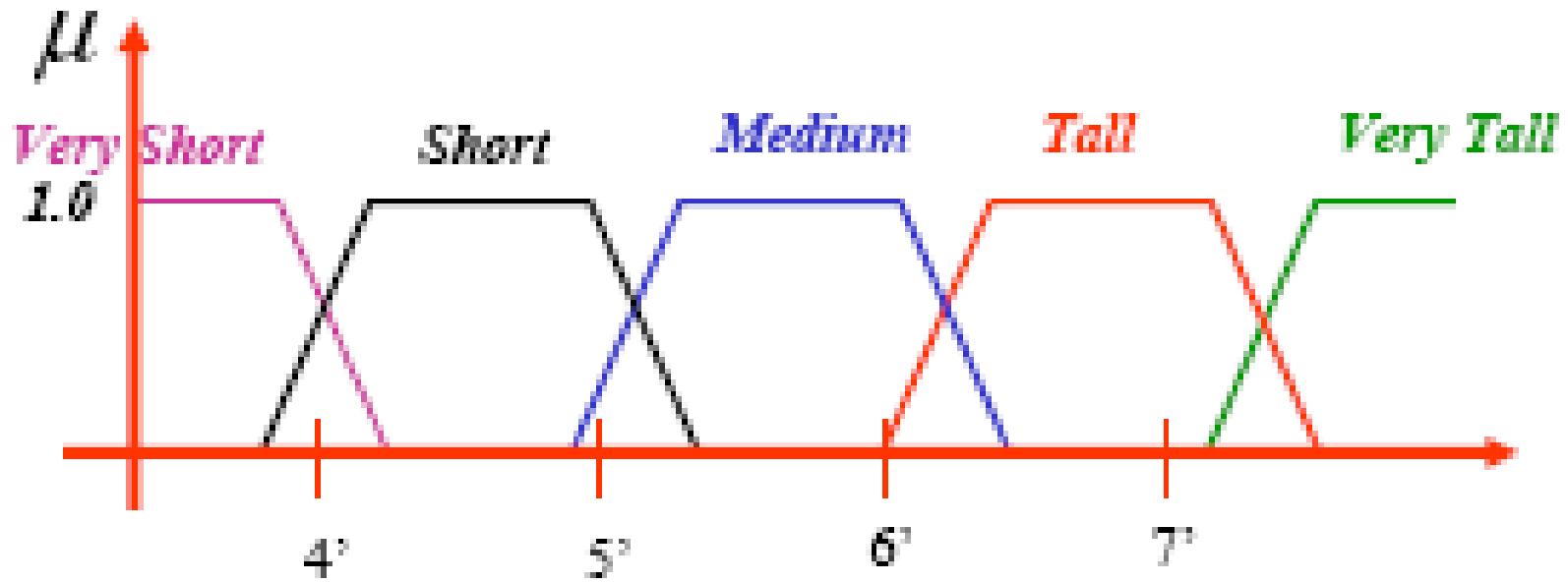
Step 1: Fuzzification

- Fuzzifier converts a crisp input into a fuzzy variable.
- Definition of the membership functions must
 - reflects the designer's knowledge
 - provides smooth transition between member and nonmembers of a fuzzy set
 - simple to calculate
- Typical shapes of the membership function are Gaussian, trapezoidal and triangular.

Example 1

- Assume we want to evaluate the health of a person based on his height and weight.
- The input variables are the crisp numbers of the person's height and weight.
- Fuzzification is a process by which the numbers are changes into linguistic words

Fuzzification of Height



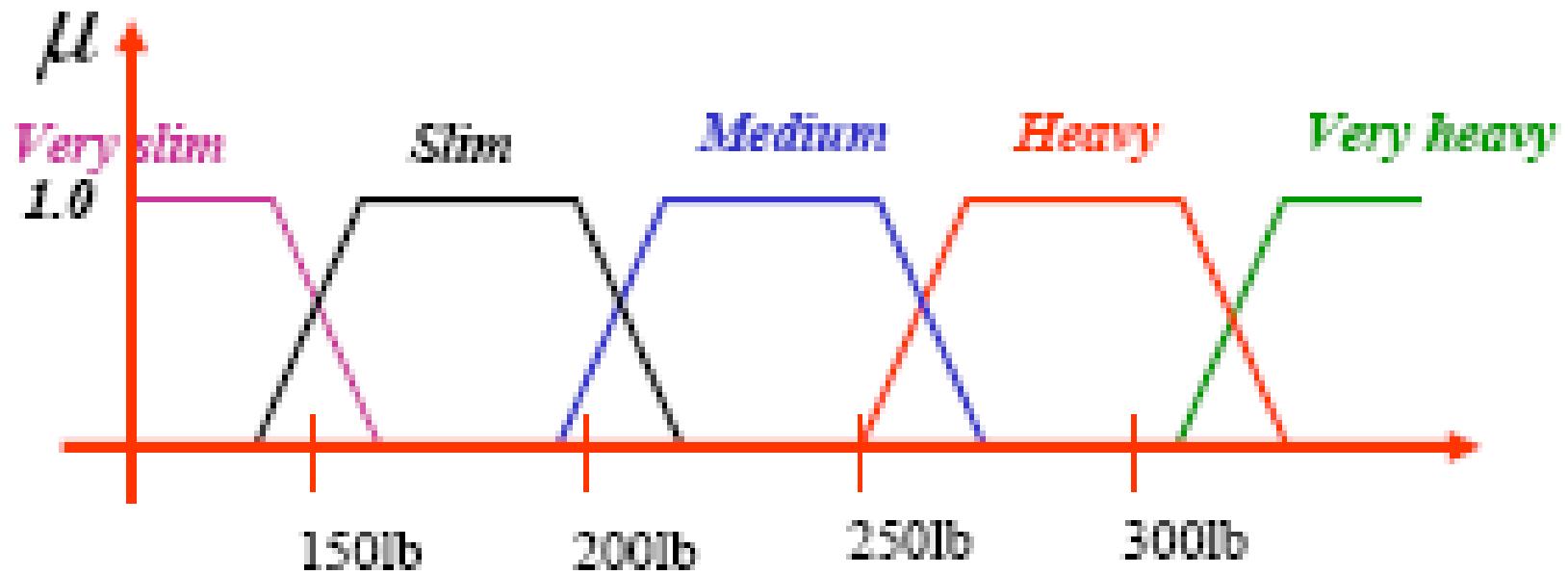
VS = very short

S = Short

M = Medium

etc.

Fuzzification of Weight



VS = very slim

S = Slim

M = Medium

etc.

Step 2: Rules

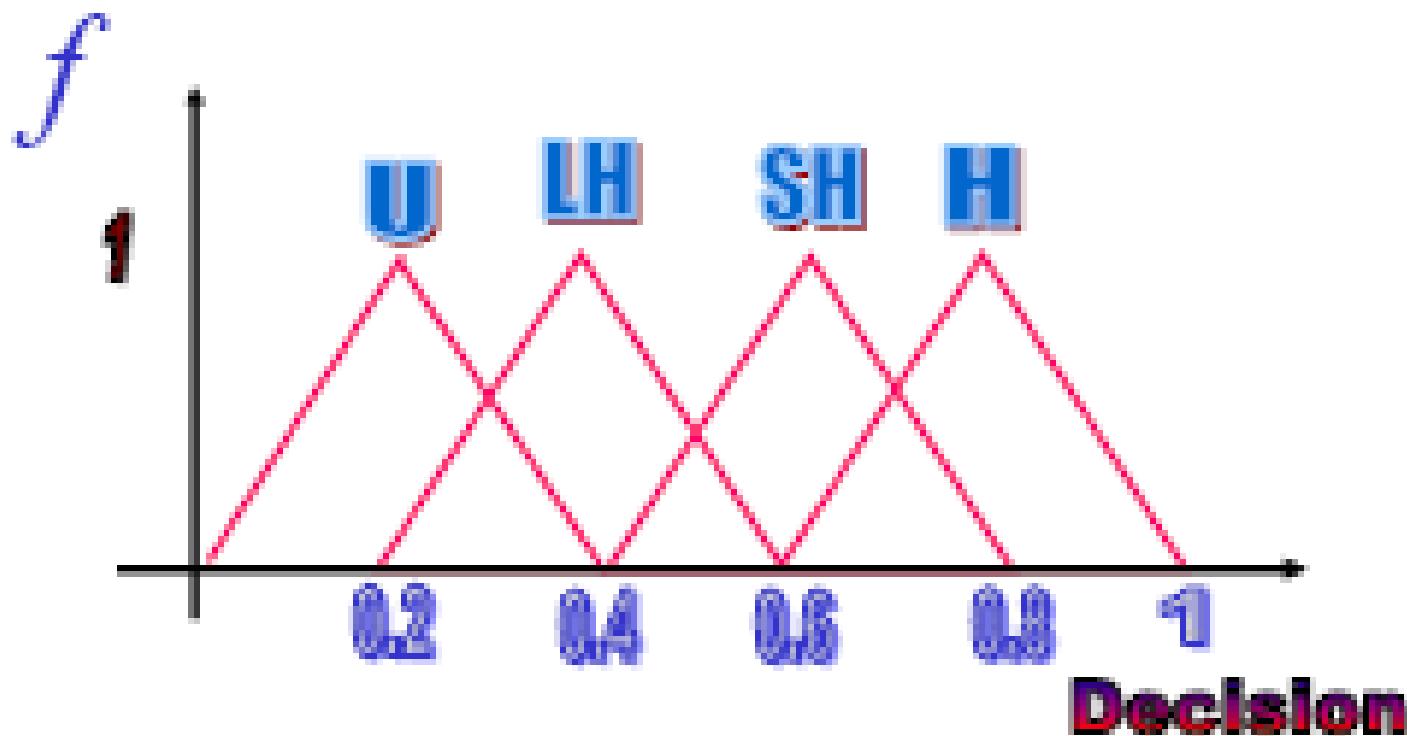
- * Rules reflect experts decisions.
- * Rules are tabulated as fuzzy words
- * Rules can be grouped in subsets
- * Rules can be redundant
- * Rules can be adjusted to match desired results

Rules Function

- Rules are tabulated as fuzzy words
 - Healthy (H)
 - Somewhat healthy (SH)
 - Less Healthy (LH)
 - Unhealthy (U)
- Rule function f

$$f = \{U, LH, SH, H\}$$

Fuzzified Decision



$$f = \{U, LH, SH, H\}$$

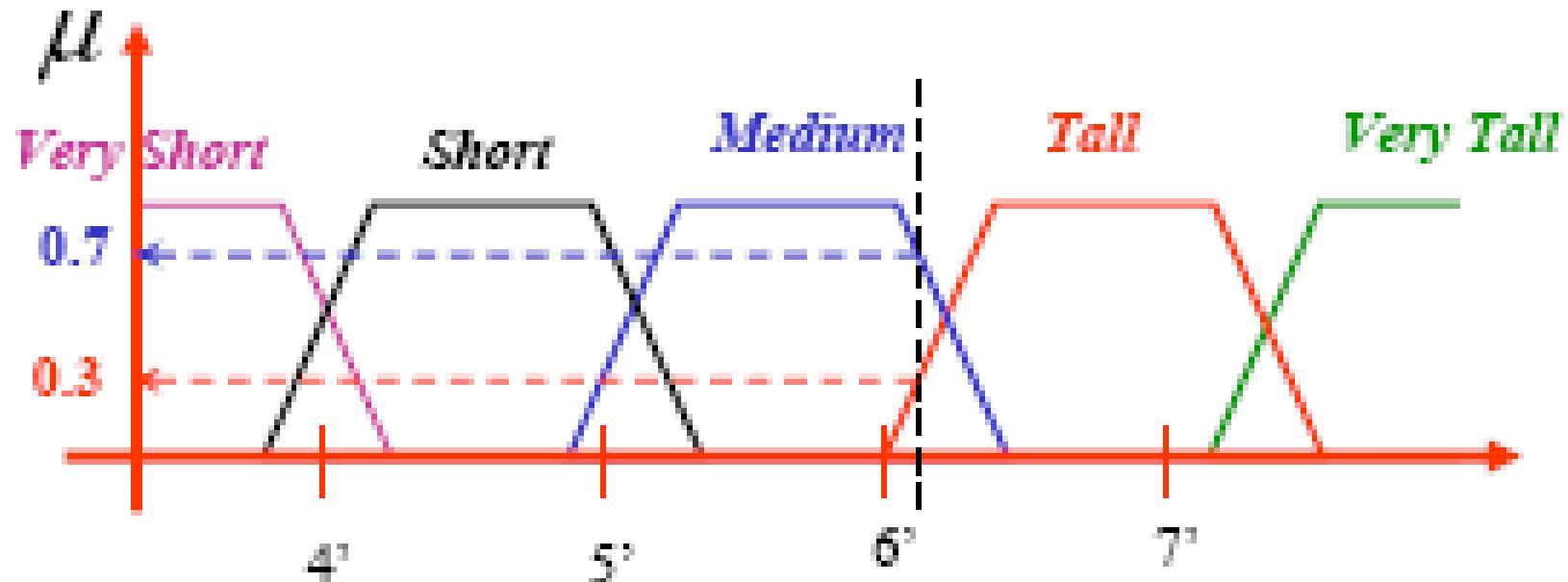
Fuzzy Rules Table

| | | Weight | | | | |
|--------|------------|-----------|------|--------|-------|------------|
| | | Very Slim | Slim | Medium | Heavy | Very Heavy |
| Height | Very Short | H | SH | LH | U | U |
| | Short | SH | H | SH | LH | U |
| | Medium | LH | H | H | LH | U |
| | Tall | U | SH | H | SH | U |
| | Very Tall | U | LH | H | SH | LH |

Step 3: Calculate

- For a given person, compute the membership of his/her weight and height
- Example:
 - Assume that a person height is 6' 1"
 - Assume that the person's weight is 140 lb

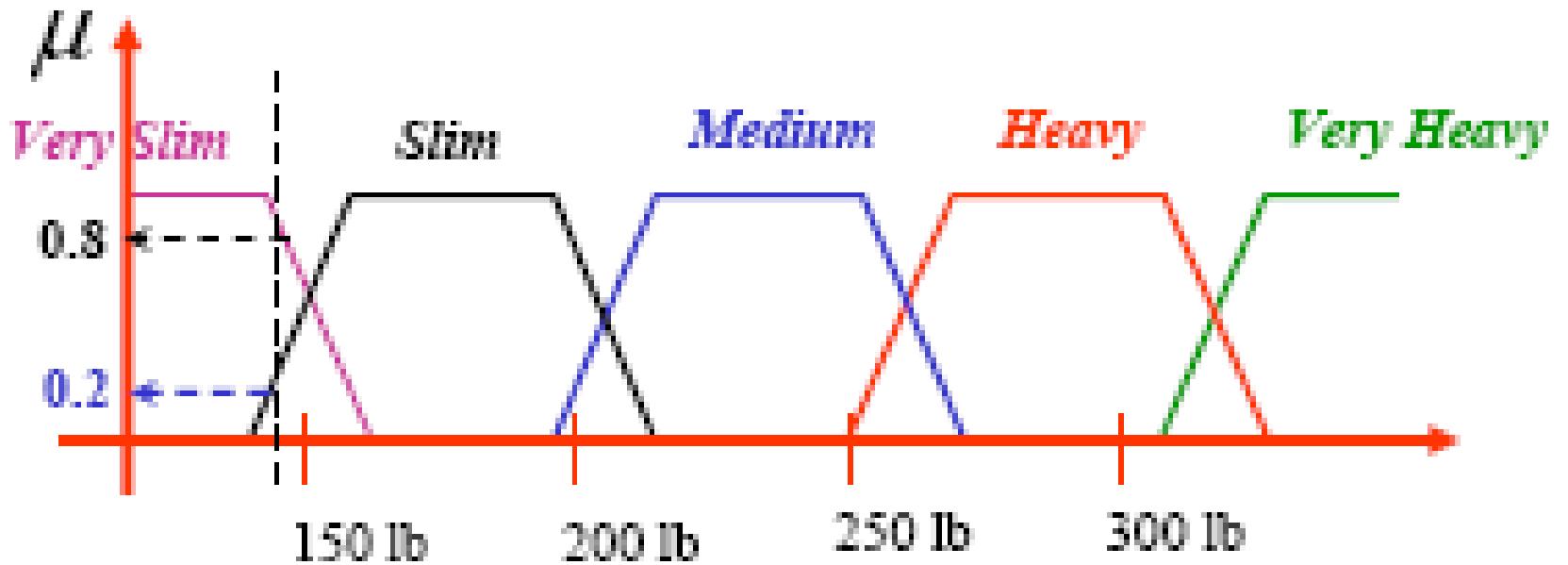
Membership of Height



$$\mu_{\text{height}} = \{\mu_{VS} \ \mu_S \ \mu_M \ \mu_T \ \mu_{VT}\}$$

$$\mu_{\text{height}} = \{0 \quad 0 \quad 0.7 \quad 0.3 \quad 0\}$$

Membership of Weight



$$\mu_{\text{Weight}} = \{\mu_{VS} \ \mu_S \ \mu_M \ \mu_H \ \mu_{VH}\}$$

$$\mu_{\text{Weight}} = \{0.8 \ 0.2 \ 0 \ 0 \ 0\}$$

Step 4: Activate Rules

| | | Weight | | | | |
|--------|------------|-----------|------|--------|-------|------------|
| | | Very Slim | Slim | Medium | Heavy | Very Heavy |
| Height | Very Short | H | SH | LH | U | U |
| | Short | SH | H | SH | LH | U |
| | Medium | | | H | LH | U |
| | Tall | | | H | SH | U |
| | Very Tall | U | LH | H | SH | LH |

Substitute Membership Values

| | | Weight | | | | |
|--------|------------|--------|-----|--------|-------|------------|
| | | 0.8 | 0.2 | Medium | Heavy | Very Heavy |
| Height | Very Short | H | SH | LH | U | U |
| | Short | SH | H | SH | LH | U |
| | 0.7 | LH | H | H | LH | U |
| | 0.3 | U | SH | H | SH | U |
| | Very Tall | U | LH | H | SH | LH |

Perform min operation

| | | Weight | | | | |
|--------|-----------------|--------|-----|---------------|--------------|----------------|
| | | 0.8 | 0.2 | Medium (0) | Heavy (0) | V.Heavy (0) |
| Height | V. Short (0) | 0 | 0 | 0 | 0 | 0 |
| | Short (0) | 0 | 0 | 0 | 0 | 0 |
| | 0.7 | 0.7 | 0.2 | 0 | 0 | 0 |
| | 0.3 | 0.3 | 0.2 | 0 | 0 | 0 |
| | V. Tall (0) | 0 | 0 | 0 | 0 | 0 |

Step 5: Compute Decision Function

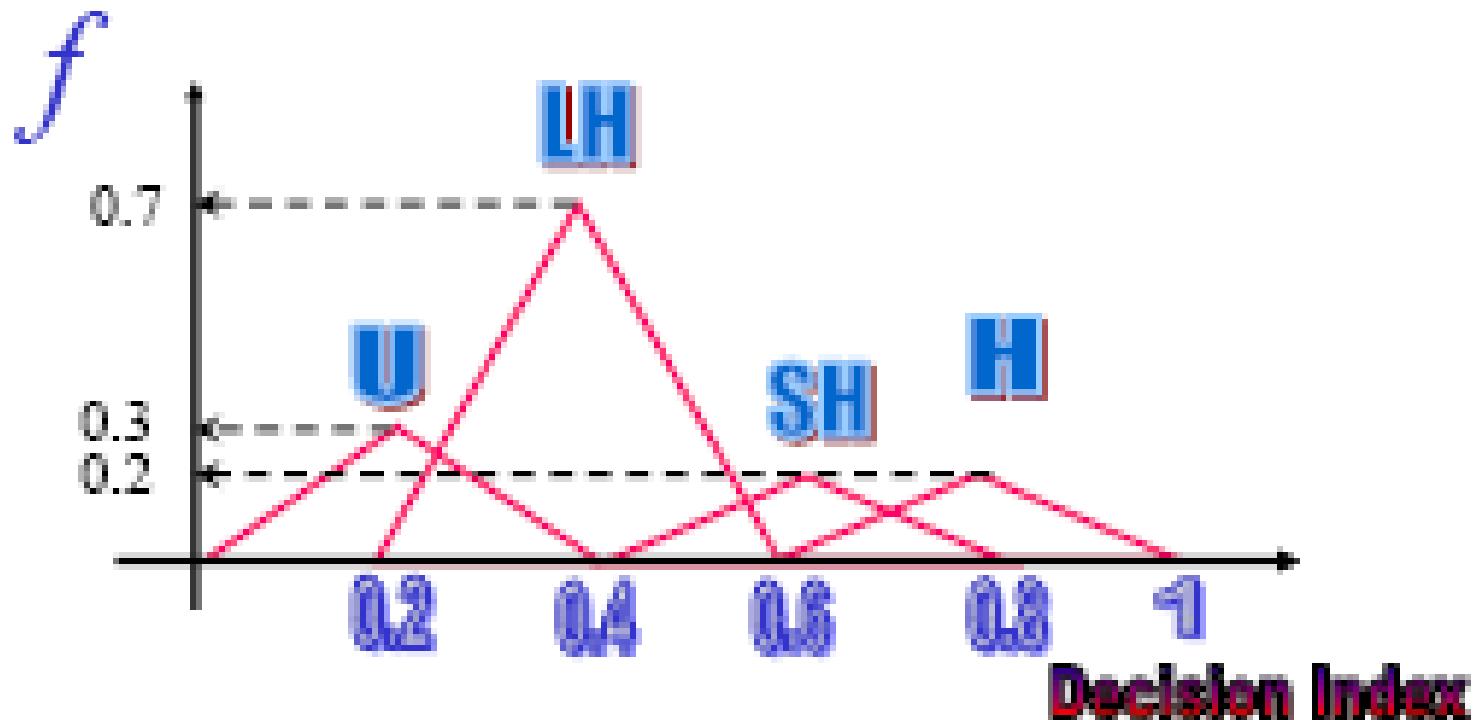
| | | Weights | |
|----------|------------|---------|-----|
| | | 0.8 | 0.2 |
| Tall (H) | Very Short | 0 | 0 |
| | Short | 0 | 0 |
| | 0.7 | 0.7 | 0.2 |
| | 0.3 | 0.3 | 0.2 |
| | Very Tall | 0 | 0 |

| | | Weights | |
|----------|------------|---------|-----|
| | | 0.8 | 0.2 |
| Tall (H) | Very Short | H | SH |
| | Short | SH | H |
| | 0.7 | LH | H |
| | 0.3 | U | SH |
| | Very Tall | U | LH |

$$f = \{U, LH, SH, H\}$$

$$f = \{0.3, 0.7, 0.2, 0.2\}$$

Scaled Fuzzified Decision



$$f = \{U, LH, SH, H\}$$

$$f = \{0.3, 0.7, 0.2, 0.2\}$$

Step 6: Compute Final Decision

- *Use the fuzzified rules to compute the final decision.*
- *Two methods are often used.*
 - *Maximum Method (not often used)*
 - *Centroid*

Max Method

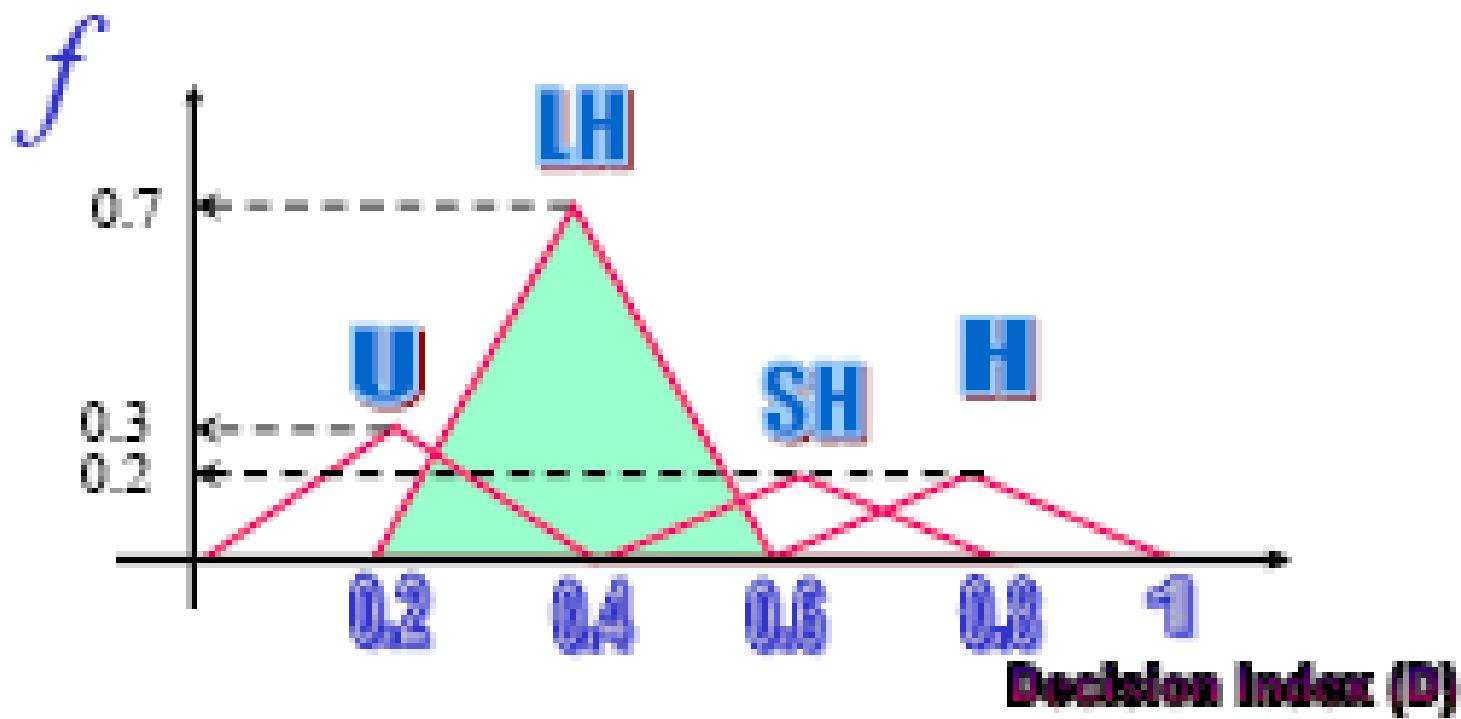
- Fuzzy set with the largest membership value is selected.
- Fuzzy decision:

$$f = \{U, LH, SH, H\}$$

$$f = \{0.3, 0.7, 0.2, 0.2\}$$

- Final Decision (FD) = Less Healthy
- If two decisions have same membership max, use the average of the two.

Max Method



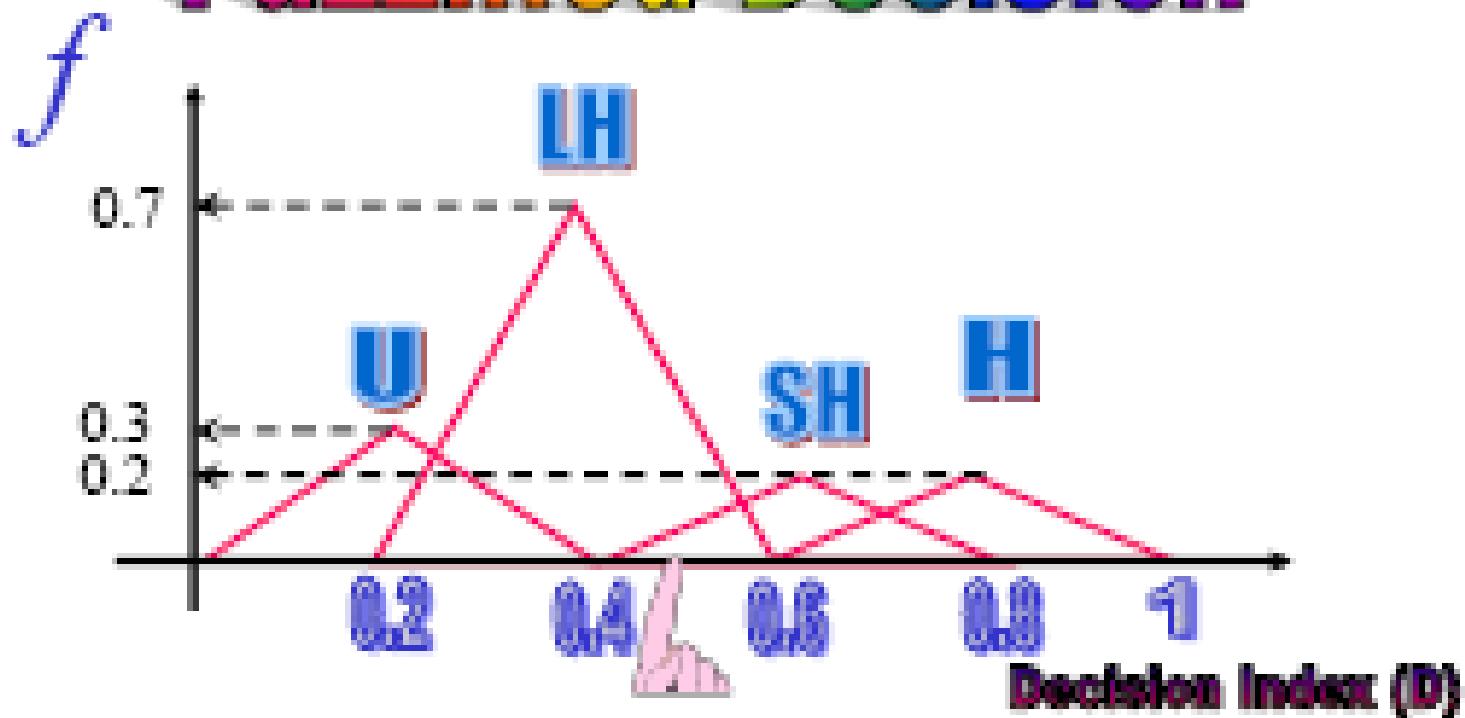
Centroid Method

$$FD = \frac{\sum \mu_i D_i}{\sum \mu_i} = \frac{\mu_U D_U + \mu_{LN} D_{LN} + \dots}{\mu_U + \mu_{LN} + \dots}$$

$$FD = \frac{0.3 \times 0.2 + 0.7 \times 0.4 + 0.2 \times 0.6 + 0.2 \times 0.8}{0.3 + 0.7 + 0.2 + 0.2} = 0.4429$$

Crisp Decision Index (D) = 0.4429

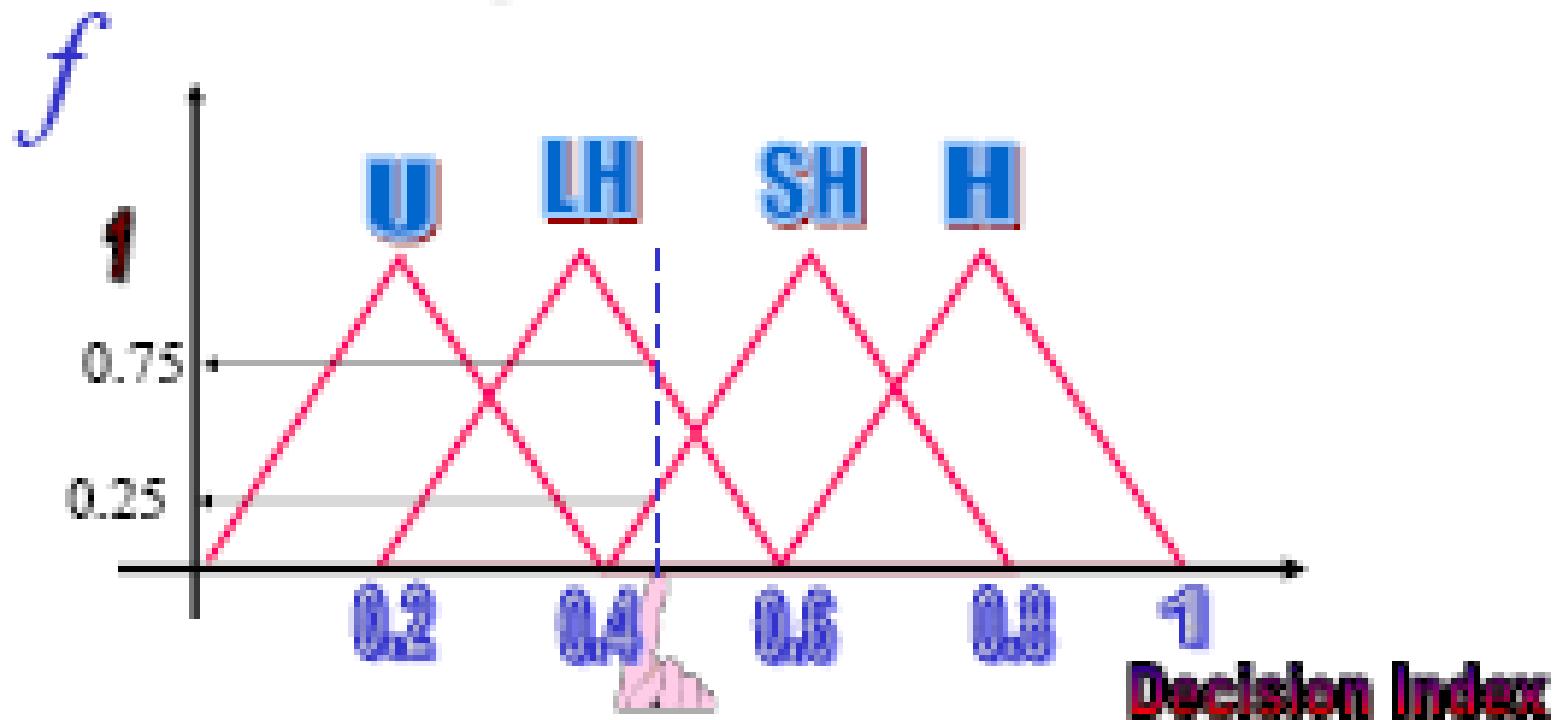
Fuzzified Decision



Crisp Decision Index (D) is the centroid

$$D = 0.4429$$

Fuzzy Decision Index



Fuzzy Decision Index (D)

75% in Less Healthy group

25% in Somewhat Healthy group