

# **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

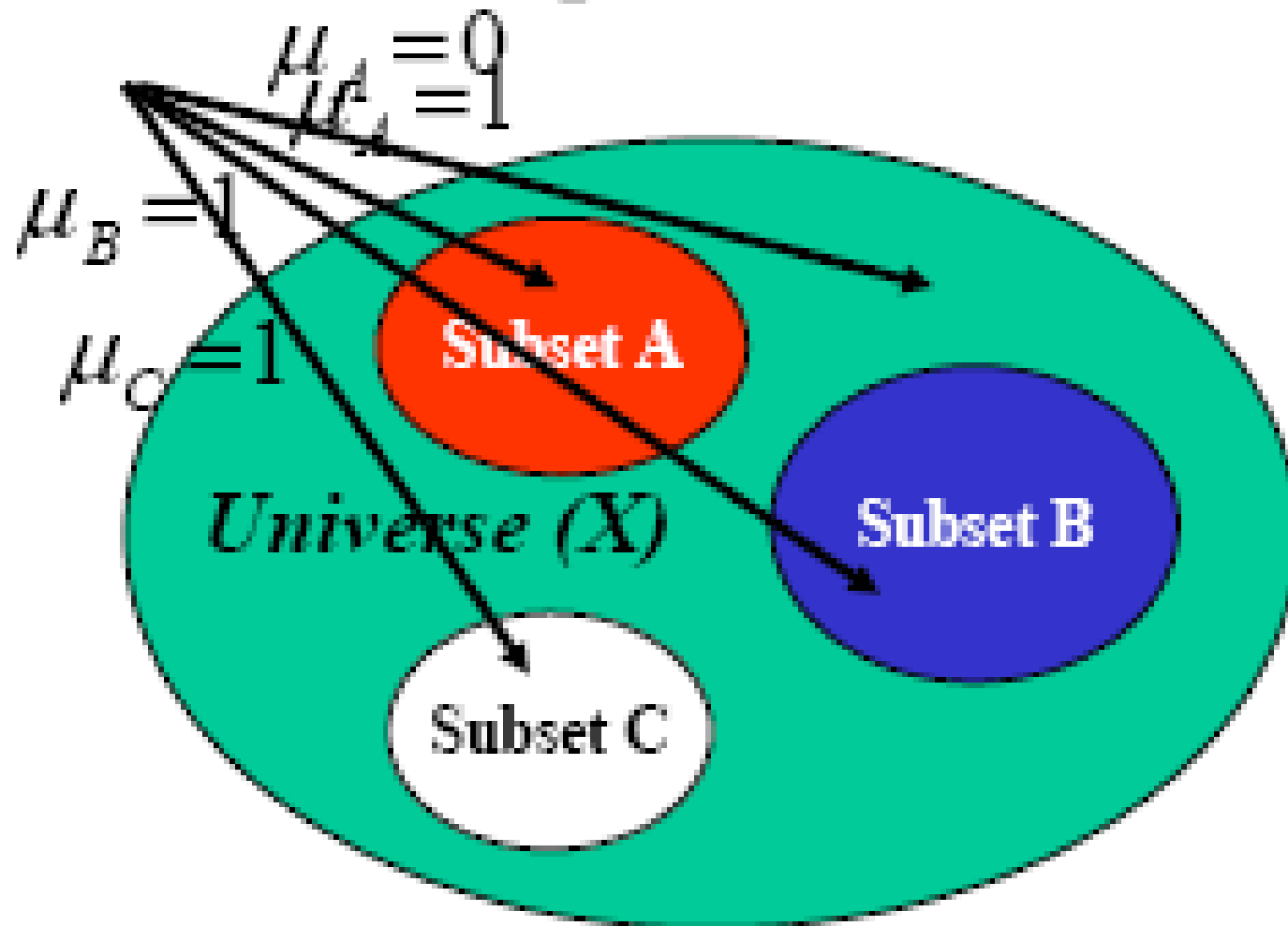
# Crisp Set

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

$\{\text{True, false}\}$

$\{1, 0\}$

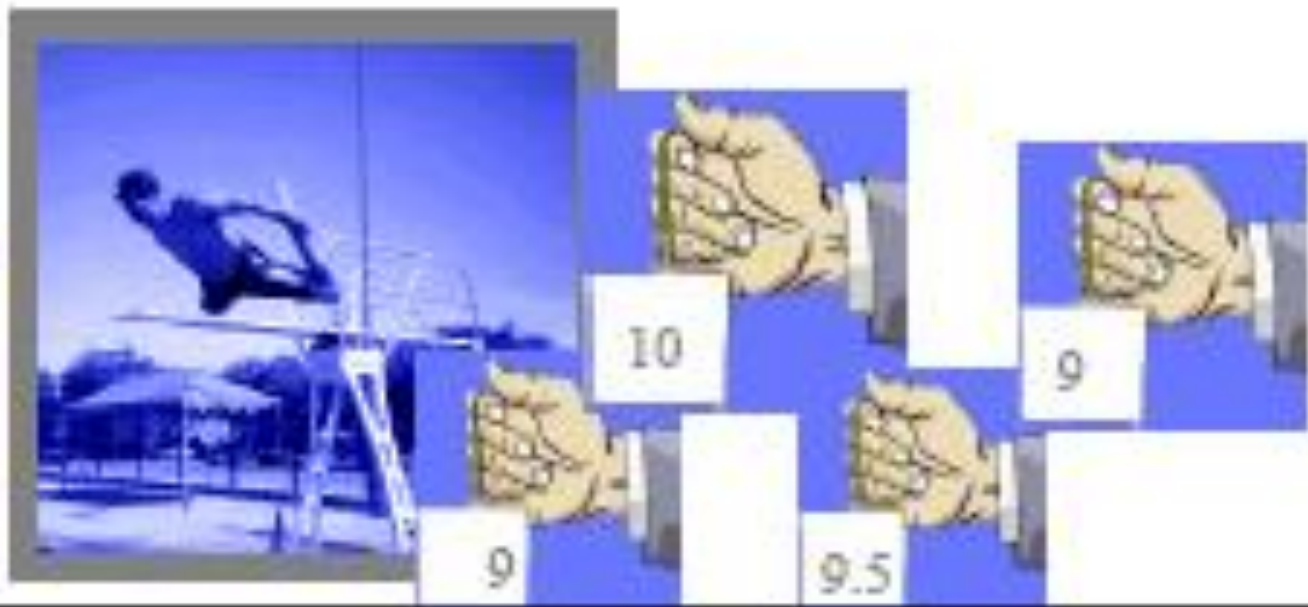
## *Crisp Set/Subset*



# Fuzzy Linguistic Variables

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

e.g. On a scale  
of one to 10,  
how good was  
the dive?



# *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.

(cite: Professor MacIsaac)



# 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.

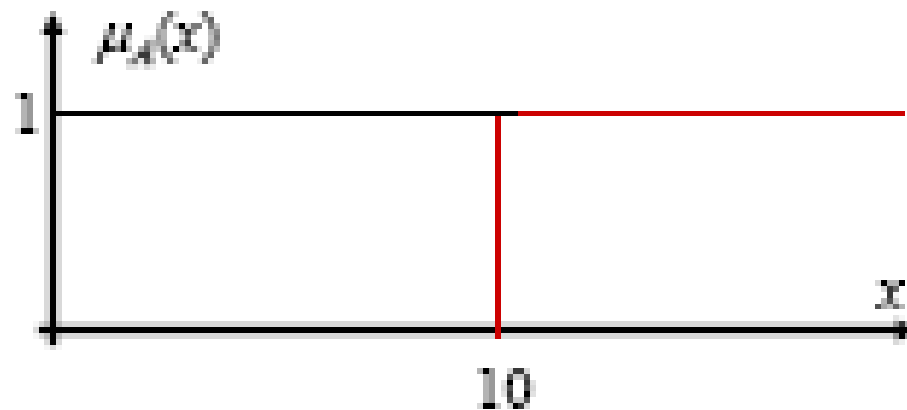


(John Heintzel)

# Crisp Membership Functions

- Crisp membership functions ( $\mu$ ) are either one or zero.
- e.g. Numbers greater than 10.

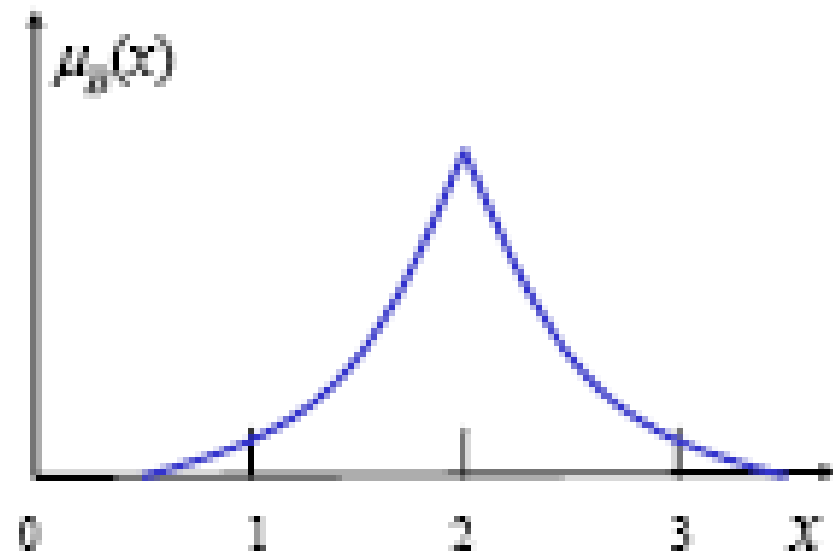
$$A = \{x \mid x > 10\}$$



# 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



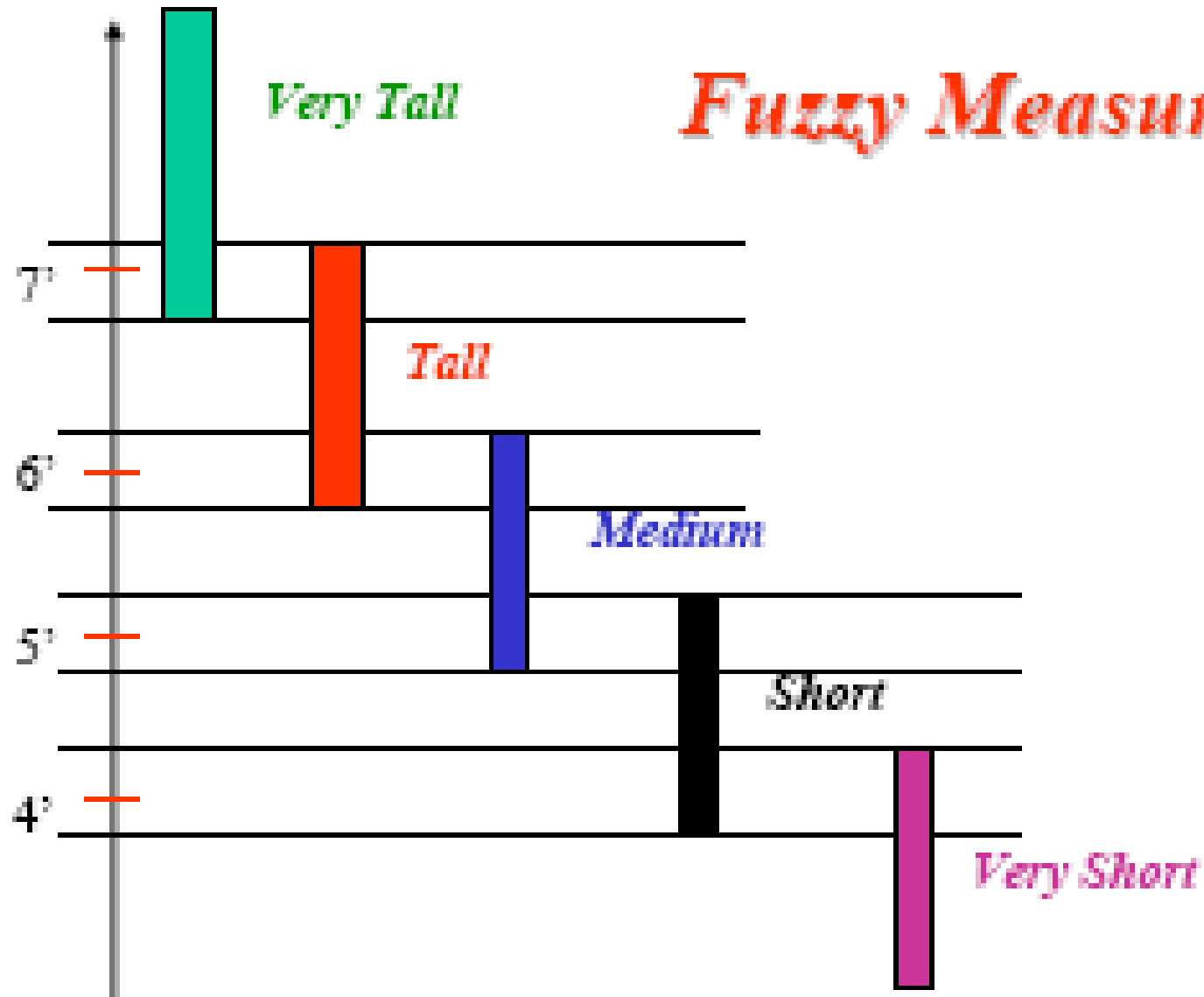
*Tall or Short?*



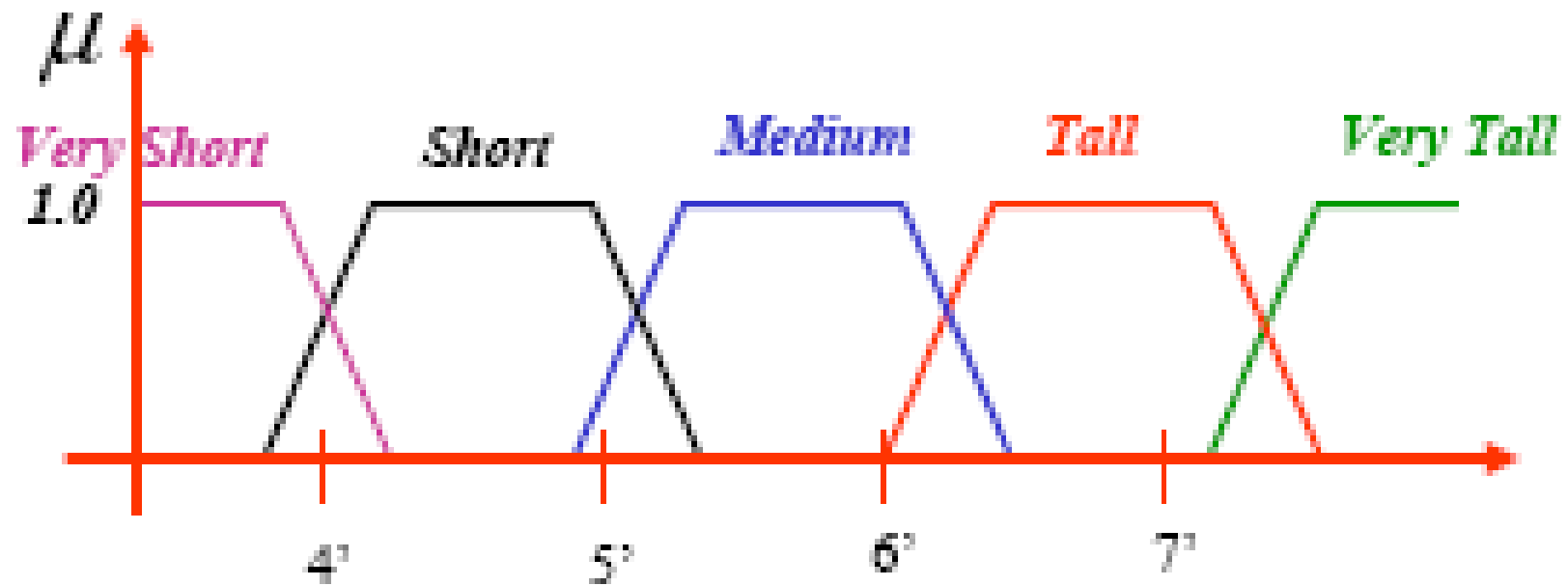
Short



# *Fuzzy Measures*

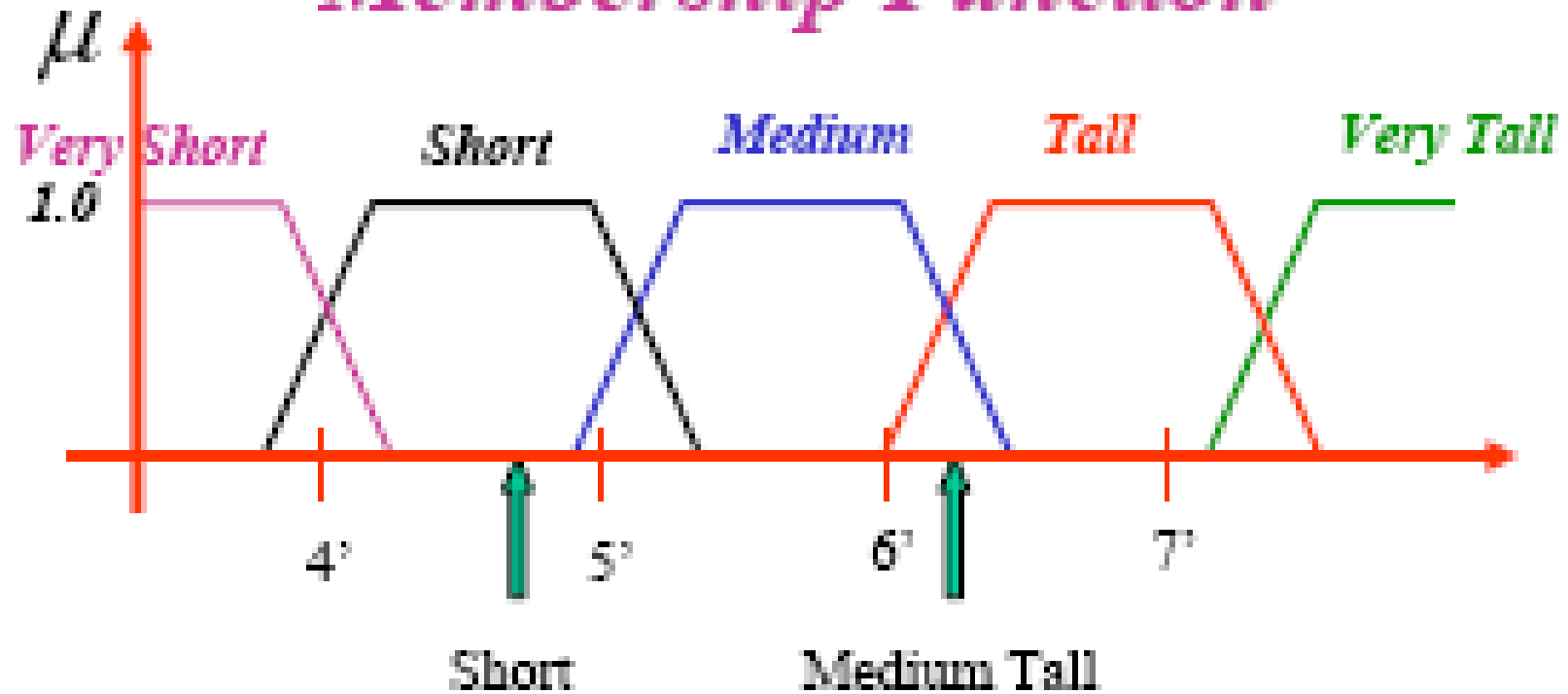


# Membership Function



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

# Membership Function



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

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

# *Fuzzy Logic Operations*

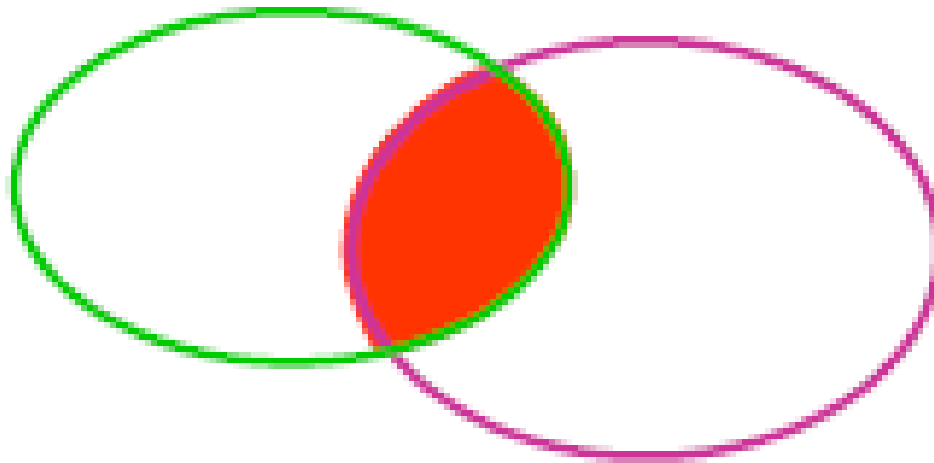
## **Fuzzy union operation or fuzzy *OR***



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

# *Fuzzy Logic Operations*

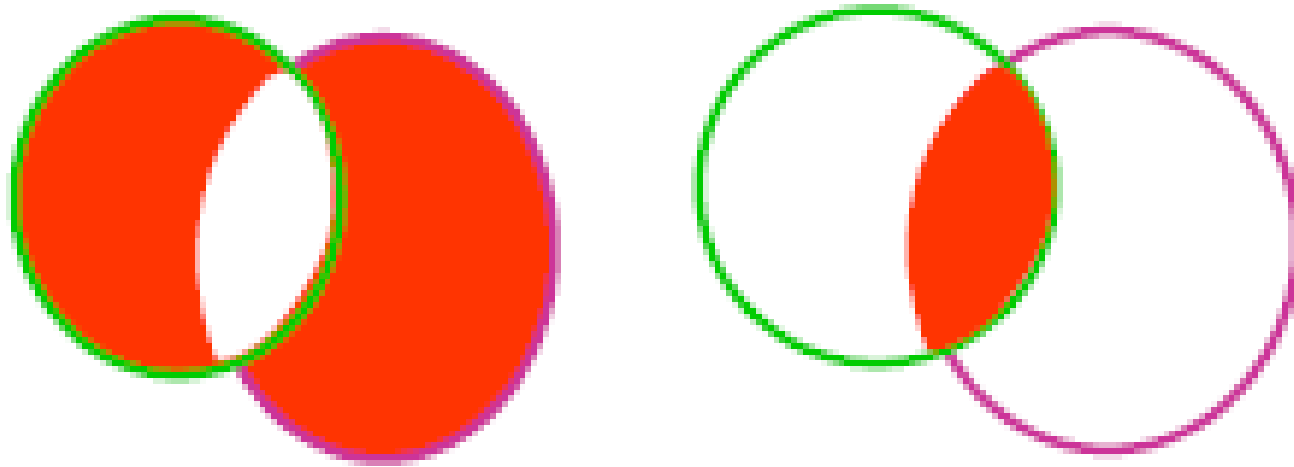
**Fuzzy intersection operation or fuzzy *AND***



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

# *Fuzzy Logic Operations*

## Complement operation



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

# *Fuzzy Logic Operations*

Fuzzy union operation or fuzzy *OR*

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

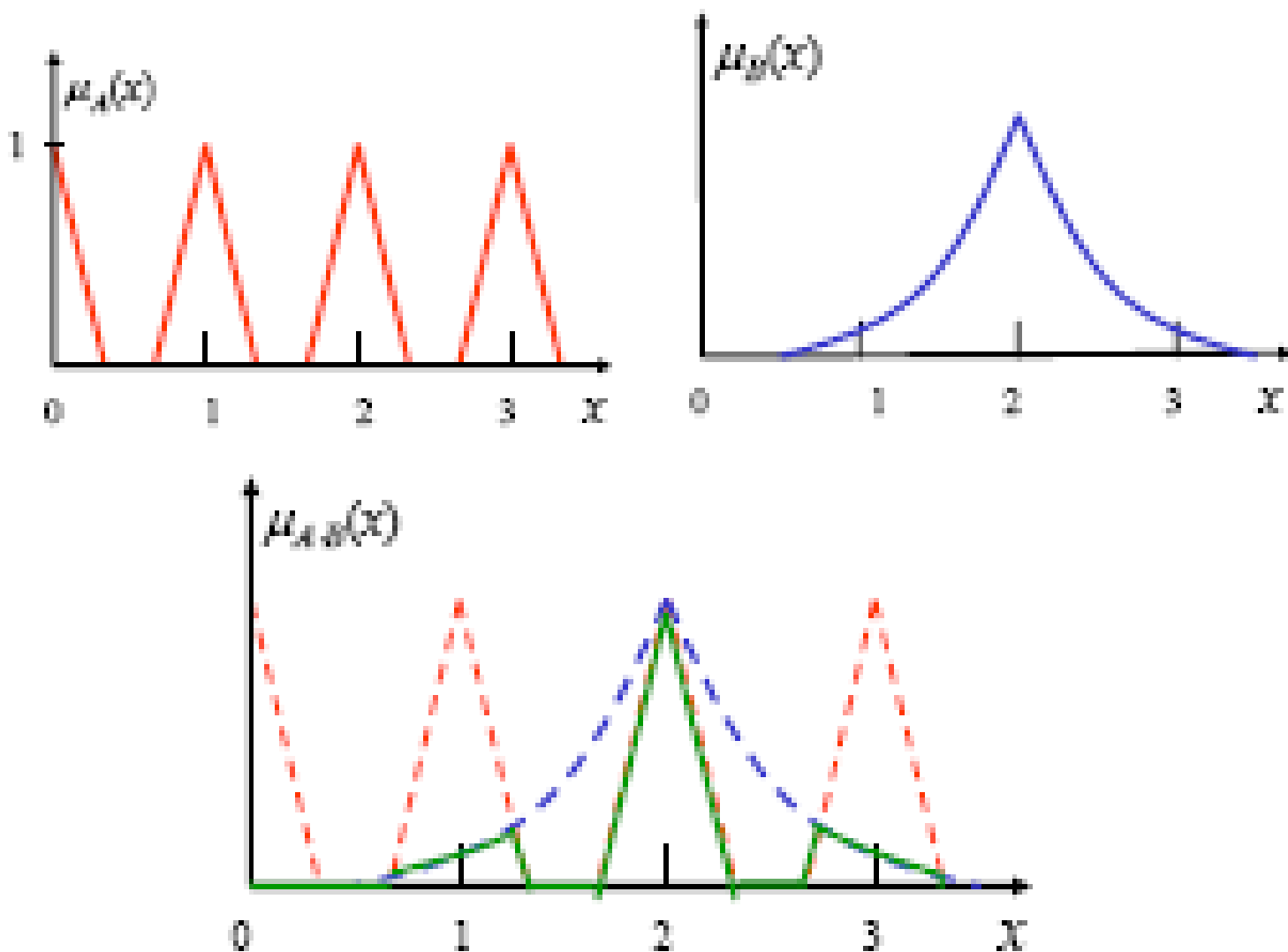
Fuzzy intersection operation or fuzzy *AND*

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

Complement operation

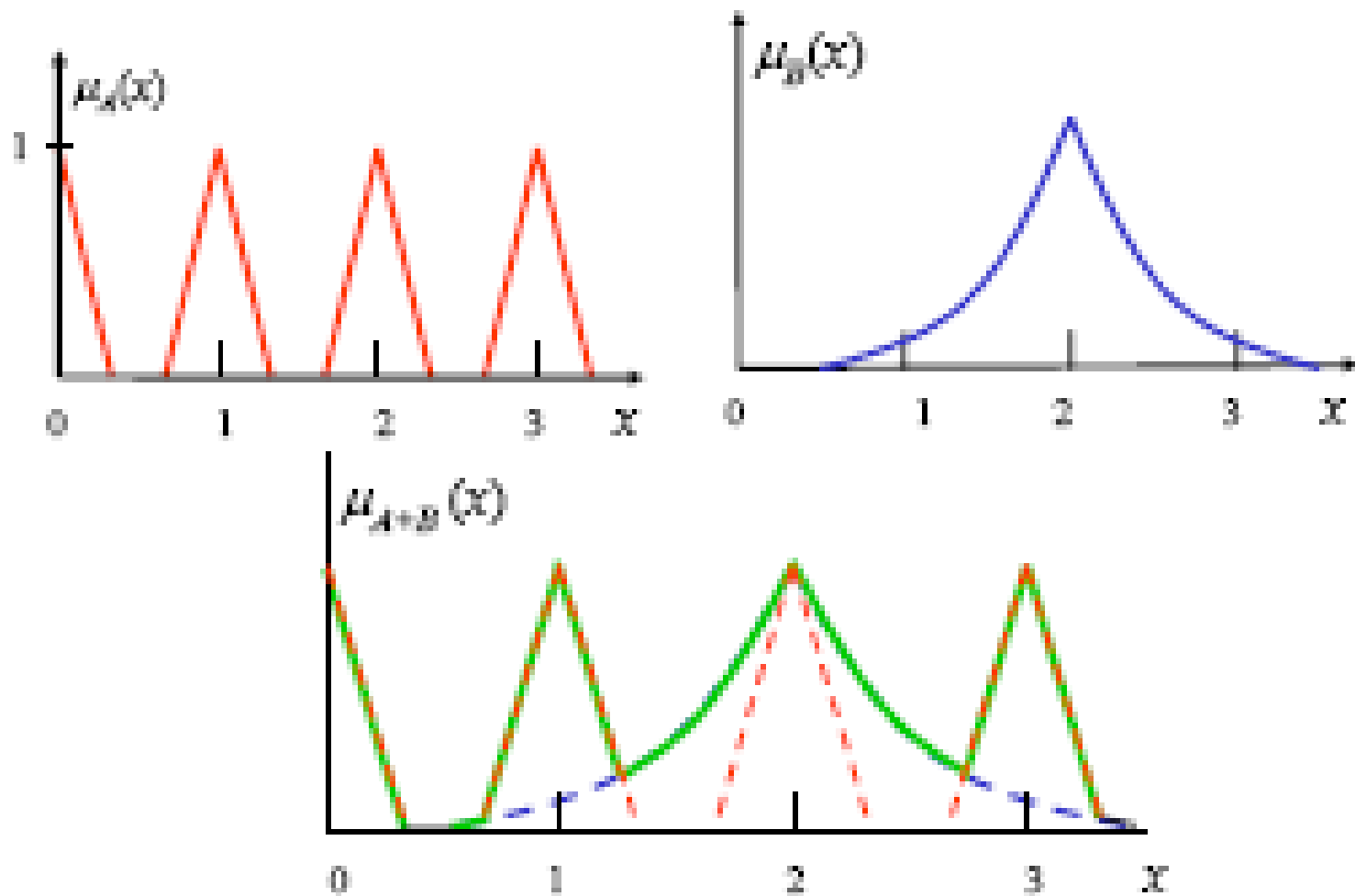
$$\mu_{\bar{A}}(x) = 1 - \mu_A(x)$$

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

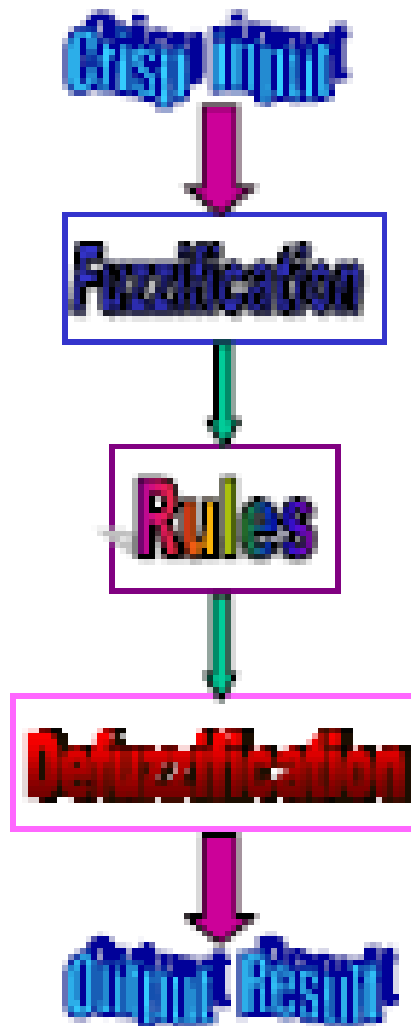




$$\mu_{A+B}(x) = \max[\mu_A(x), \mu_B(x)]$$



# Fuzzy System Basic Components



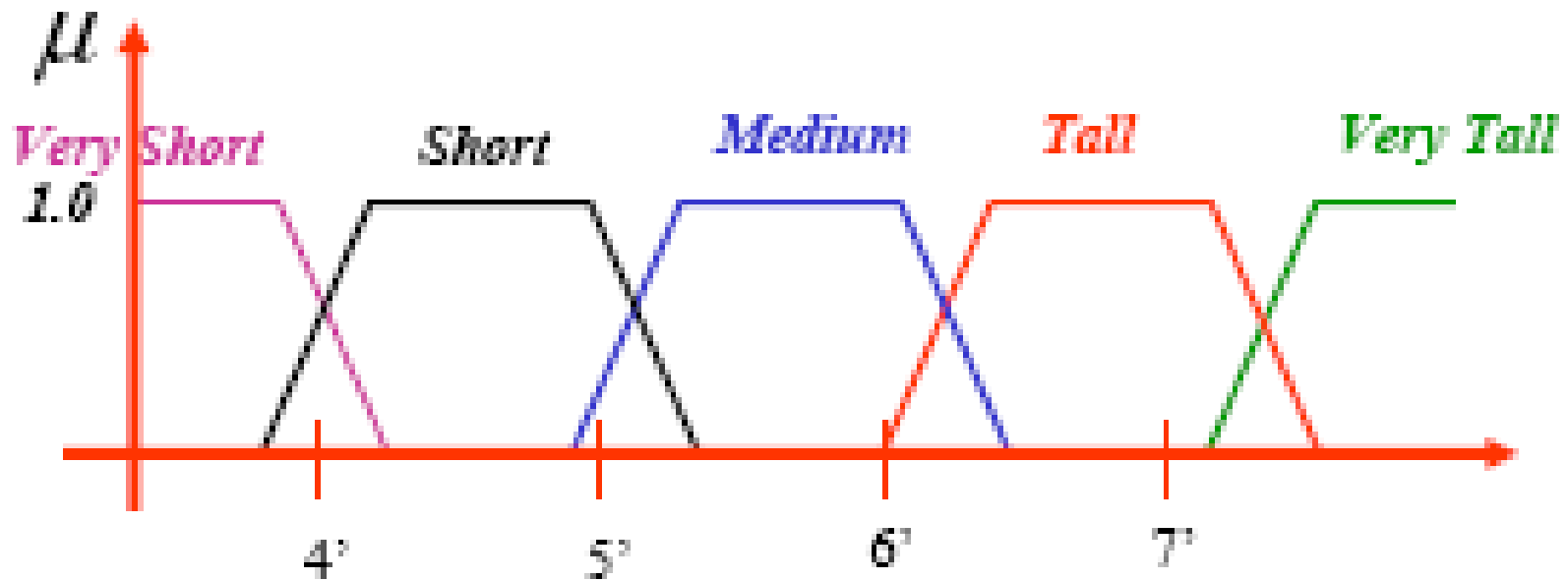
# Step 1: Fuzzification

- 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*



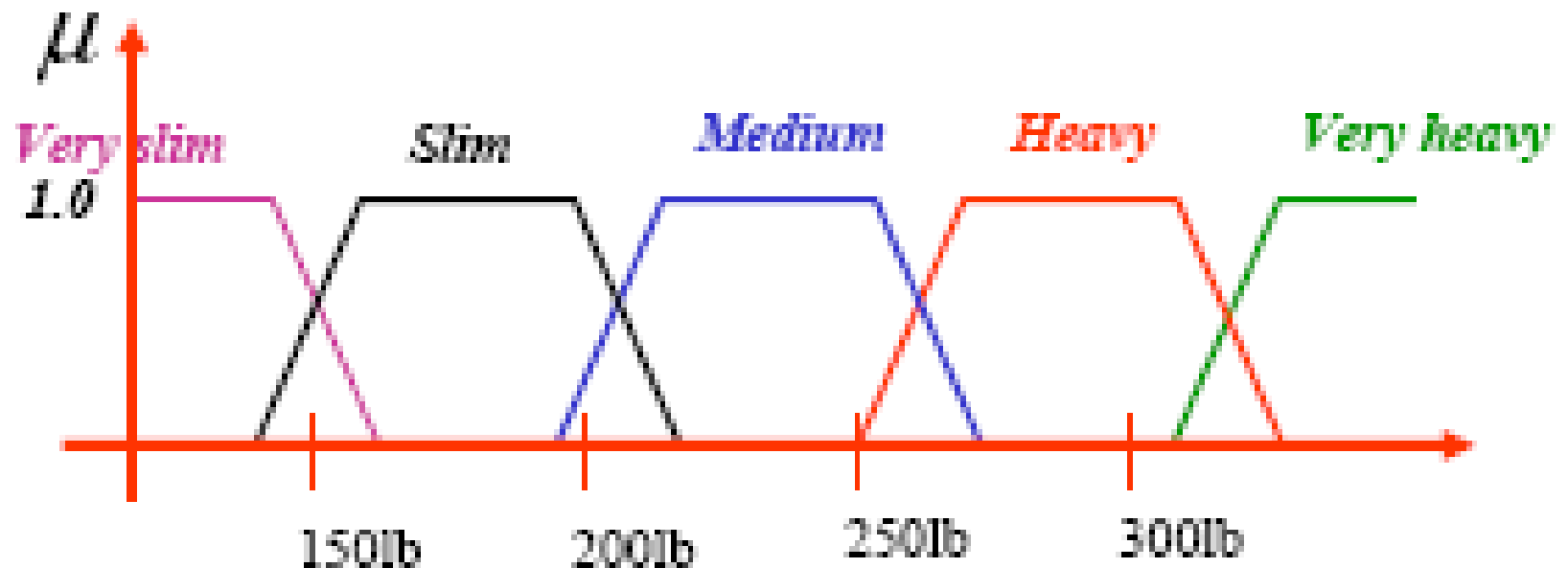
VS = very short

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

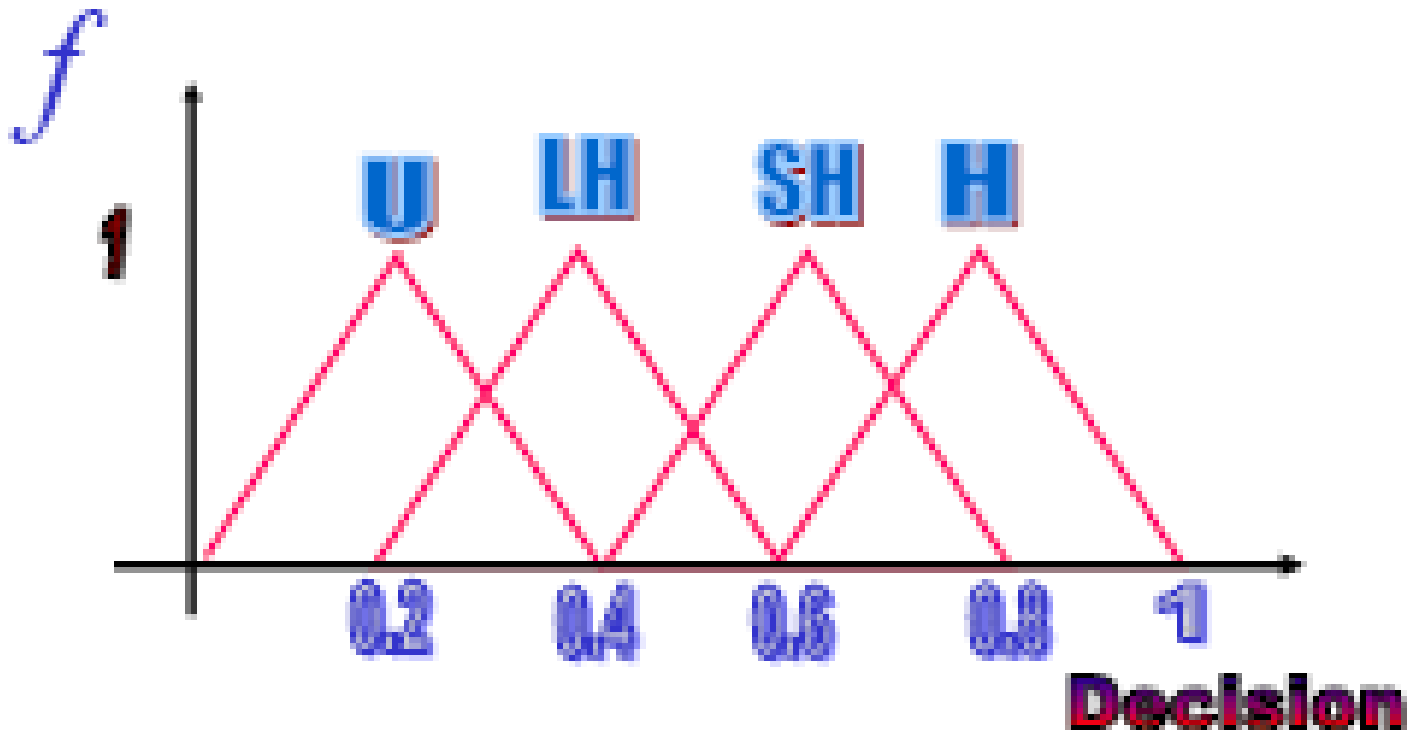
# Rules Function

- 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



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

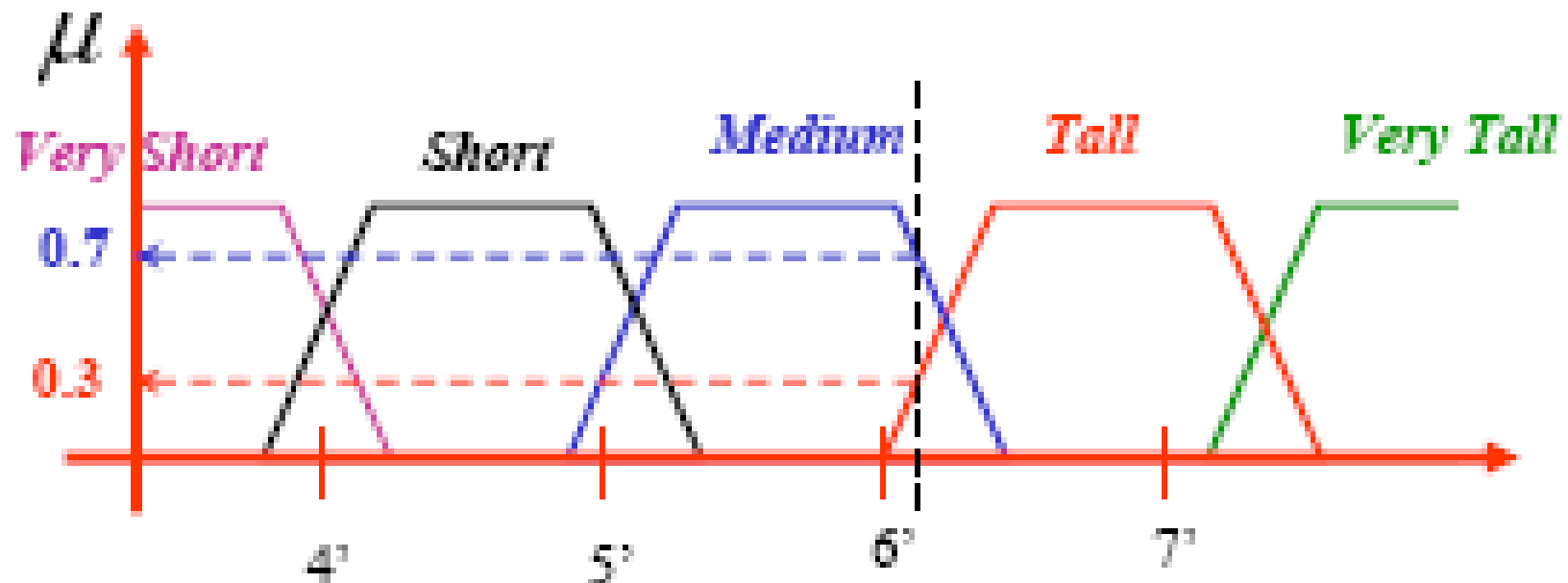
# Fuzzy Rules Table

		Weight				
Height		Very Slim	Slim	Medium	Heavy	Very Heavy
	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

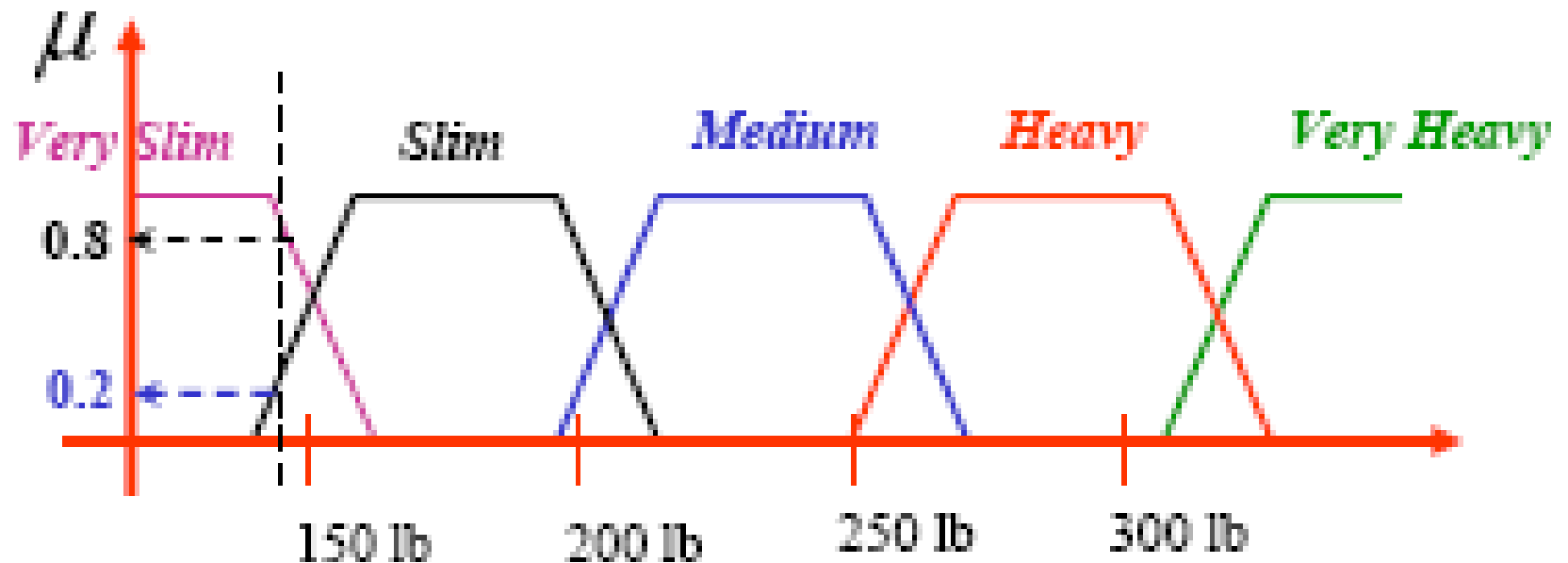
## Membership of Height



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

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

## Membership of Weight



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

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

# Step 4: Activate Rules

		Weight				
Height		Very Slim	Slim	Medium	Heavy	Very Heavy
	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						
Height		0.8	0.2	Medium	Heavy	Very Heavy
	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				
Height		0.8	0.2	Medium (0)	Heavy (0)	V.Heavy (0)
	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

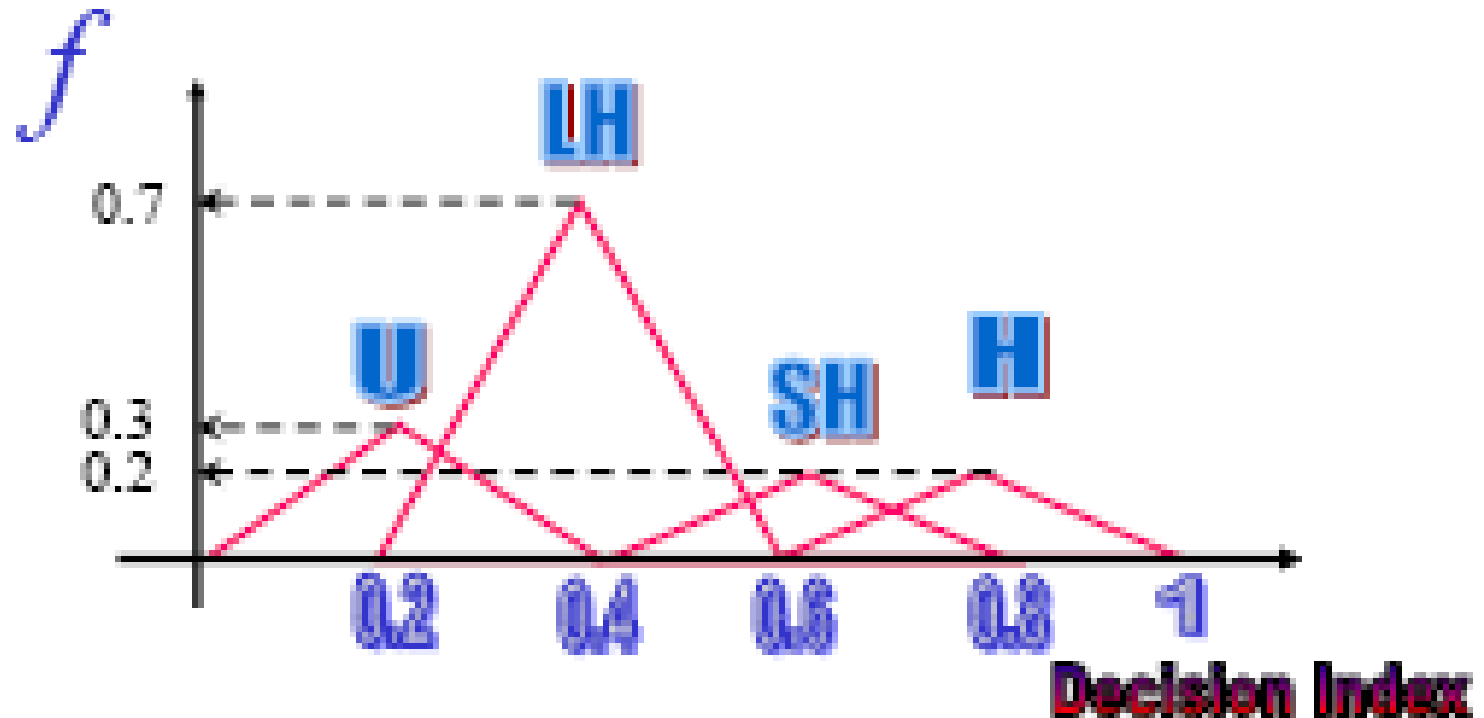
Weight			
		0.8	0.2
	V. Short (S)	0	0
	Short (S)	0	0
	0.7	0.7	0.2
	0.3	0.3	0.2
	V. Tall (T)	0	0

Weight			
		0.8	0.2
	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

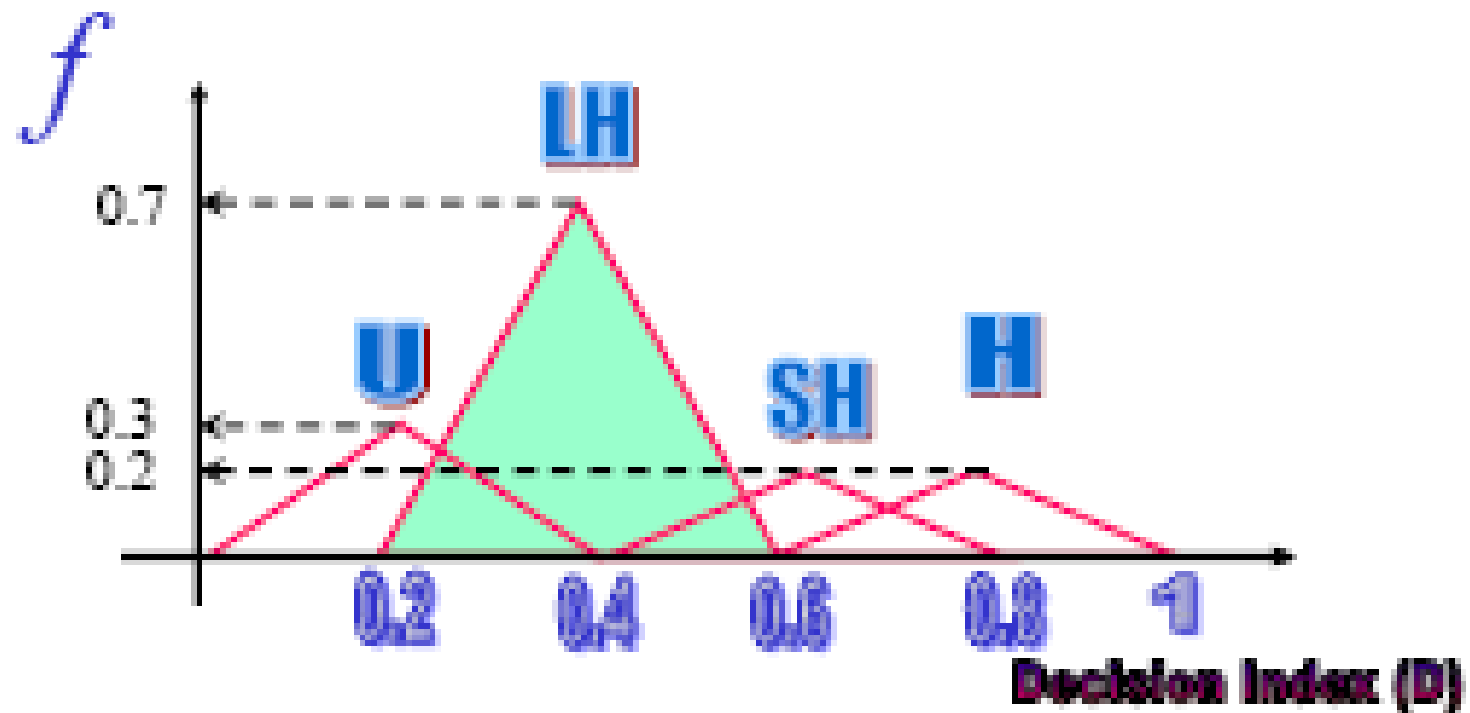
- Fuzzy set with the largest membership value is selected.
- Fuzzy decision:

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

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

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

# Max Method



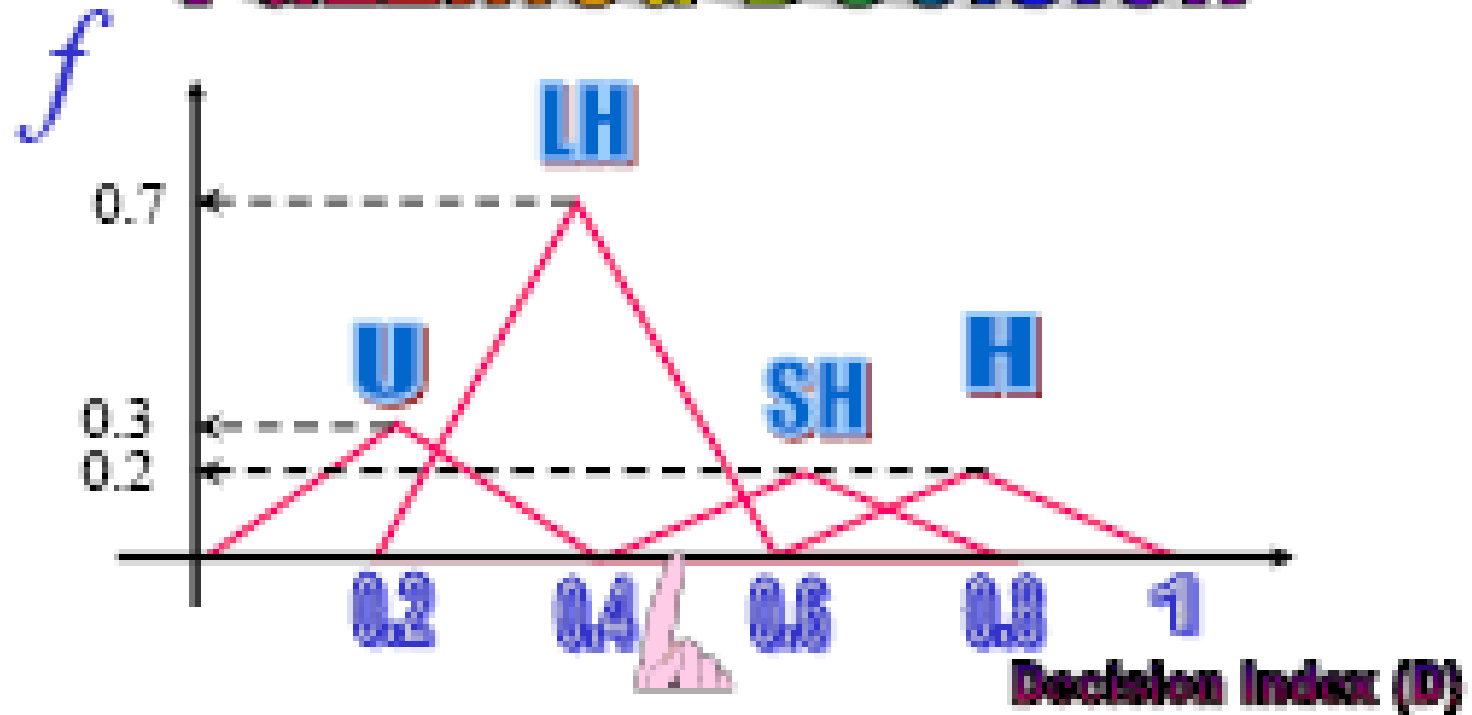
# Centroid Method

$$FD = \frac{\sum \mu D}{\sum \mu} = \frac{\mu_U D_U + \mu_{LN} D_{LN} + \dots}{\mu_U + \mu_{LN} + \dots}$$

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

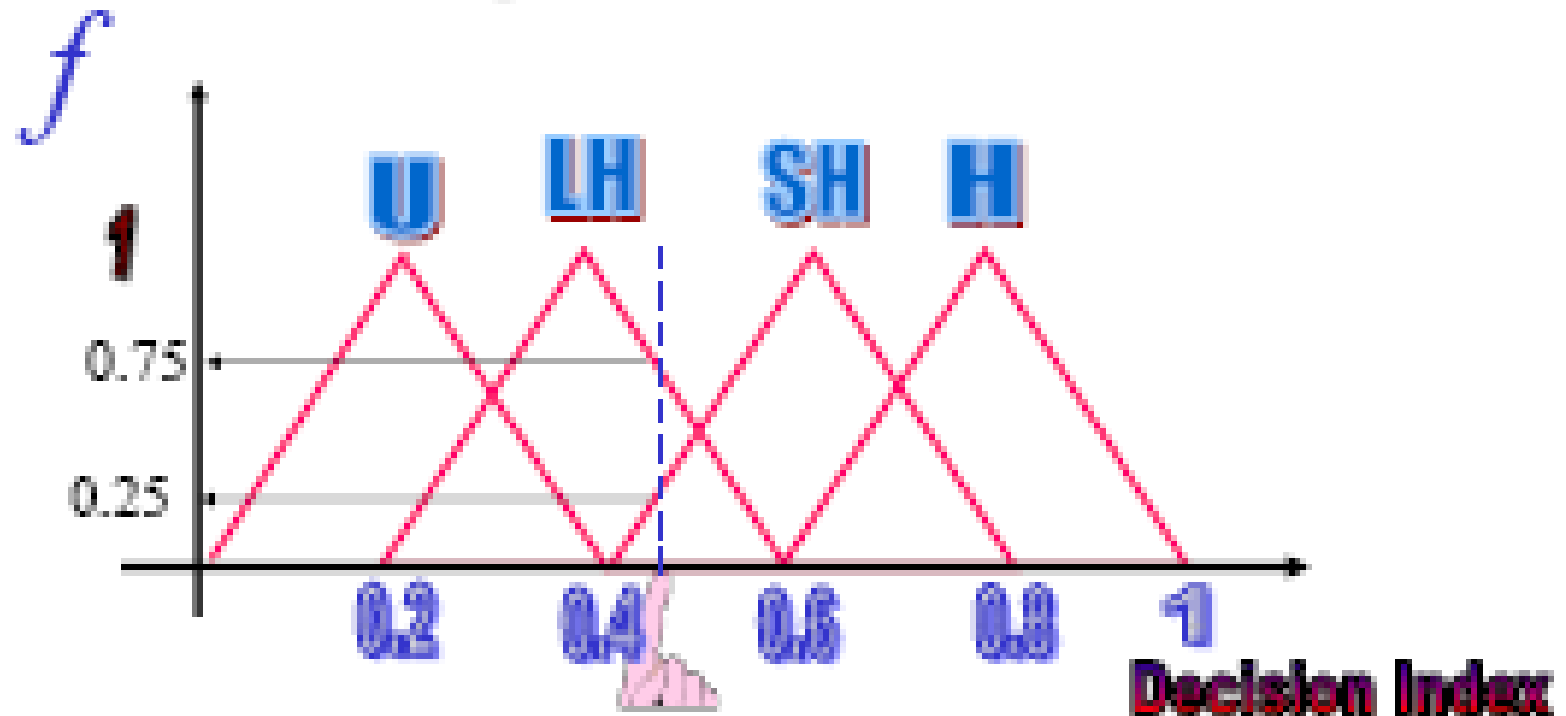
**Crisp Decision Index (D) = 0.4429**

# 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