

Fuzzy Sets & Expert Systems in Computer Eng. (5):

Linguistic Variables

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Outline

- **Linguistic variable**
- **Language**
- **Grammars**
- **Notation and Operations for Sampled Membership Functions**
- **Notation and Operations for Parametrized Membership Functions**
 - Possible Problems with Convexity and Normality
- **Bug Report**
- **Defuzzification Methods**

Linguistic Variables

A numerical variables takes numerical values:

Age = 65

A linguistic variables takes linguistic values:

Age is old

A linguistic values is a fuzzy set.

It has a LABEL and a MEANING

LABEL: Symbol, Sentence in a Language

**MEANING: Fuzzy Subset of a Universe of
Discourse**

Linguistic Variables

All linguistic values form a term set:

**T(age) = {young, not young, very young, ...
middle aged, not middle aged, ...
old, not old, very old, more or less old, ...
not very young and not very old, ...}**

Linguistic Variable (Formal Definition)

A Linguistic Variable is characterized by the five-tuple:

$(\sigma, T(\sigma), U, G, A)$

where:

σ is the name of the variable

$T(\sigma)$ is the Term Set of σ

U is the Universe of Discourse associated with the base variable “u”

G is the Syntactic Rule (grammar) for generating the labels “t” in the termset

A is the Semantic Rule for associating to each label t, element of $T(\sigma)$, its meaning $A(t)$ - algorithmic procedure

Language (Zadeh 71)

A language can be seen as a fuzzy relation from a set of labels T to a Universe of Discourse U , which assigns a grade of membership $\mu_L(t, u)$ to each pair $(t, u) \in T \times U$

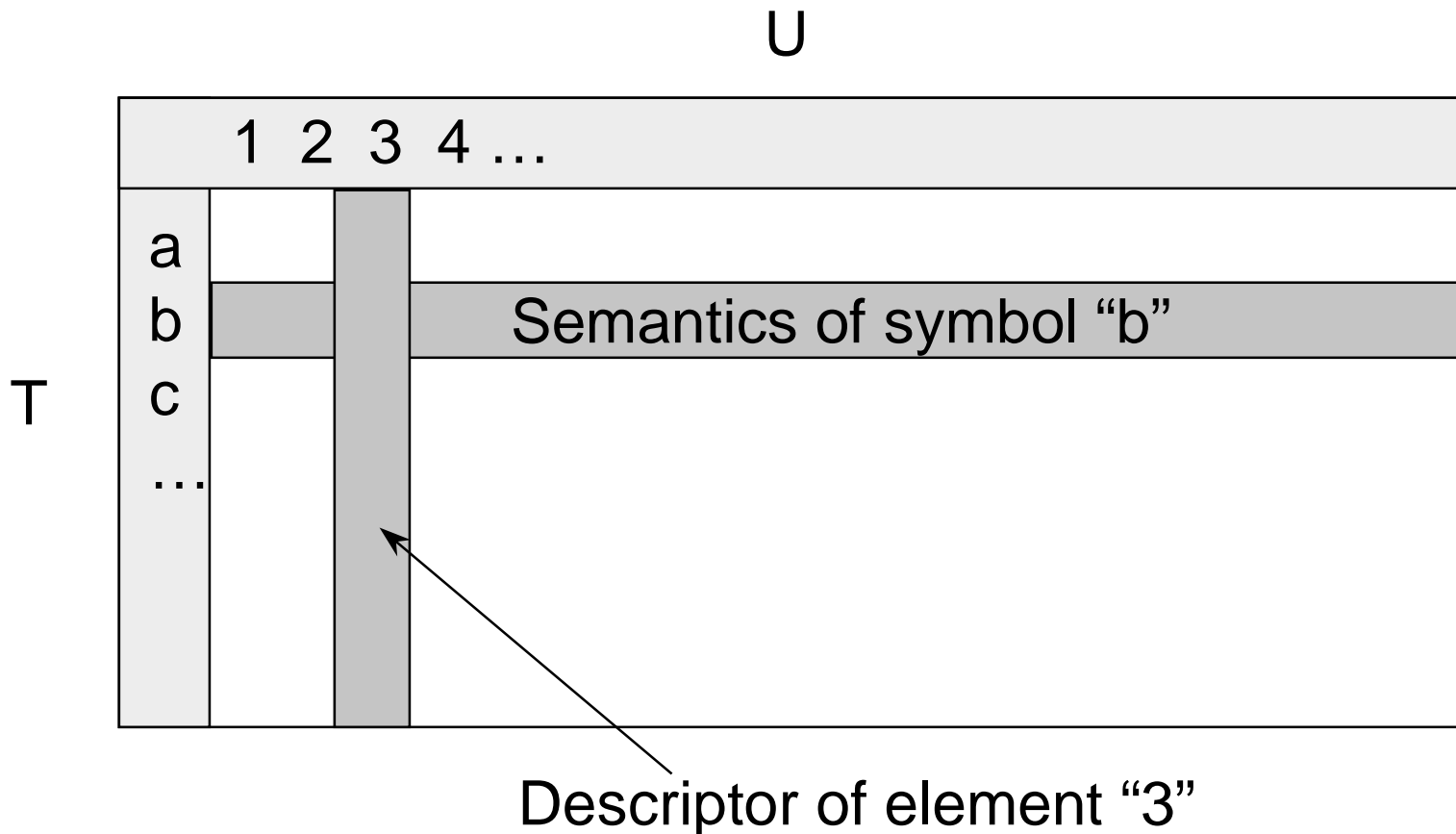
If we fix the term “ t ” say t_i , then $\mu_L(t_i, u)$ determines the fuzzy subset $A(t_i)$ whose membership function is $\mu_{A(t_i)}(u) = \mu_L(t_i, u)$

$\mu_{A(t_i)}(u)$ is the **MEANING** of the linguistic value whose **LABEL** is t_i

If we fix the term u , say u_j , then $\mu_L(t, u_j)$ is the **DESCRIPTOR** of u_j

Language (Zadeh 71)

A language is a fuzzy mapping $U \times T \Rightarrow [0,1]$



Linguistic Terms (Termset)

Label	Symbol	Universe of Discourse								
		1	2	3	4	5	6	7	8	9
Very Low	Ver_L	1	0.64	0.09	0.01	0	0	0	0	0
Low	L	1	0.80	0.30	0.10	0	0	0	0	0
More or less Low	Mor_L	1	0.89	0.55	0.32	0	0	0	0	0
Above Low	>L	0	0.20	0.70	0.90	1	1	1	1	1
Below Medium	<M	1	1.00	0.90	0.25	0	0	0	0	0
Above Low and Below Medium		0	0.20	0.70	0.25	0	0	0	0	0
Between Medium and High	L-M	0	0.29	1	0.36	0	0	0	0	0
Medium	M	0	0	0.10	0.75	1	0.75	0.10	0	0
Above Medium	>M	0	0	0	0	0	0.25	0.90	1	1
Below High	<H	1	1	1	1	1	0.90	0.70	0.20	0
Above Medium and Below High		0	0	0	0	0	0.25	0.70	0.20	0
Between Medium and High	M-H	0	0	0	0	0	0.36	1	0.29	0
More or less High	Mor_H	0	0	0	0	0	0.32	0.55	0.89	1
High	H	0	0	0	0	0	0.10	0.30	0.80	1
Very High	Ver_H	0	0	0	0	0	0.01	0.09	0.64	1

Generative Grammars

A grammar is a set of rules and procedures which produces a structurally correct language.

A generative grammar G is defined by the four-tuple:

$$G = \langle V_H, V_E, P, S \rangle$$

where:

V_H is a finite, non-empty set of Help or *non-terminal* vocabulary of G

V_E is a finite, non-empty set of End or *terminal* vocabulary of G

P is a finite, non-empty set of ordered pairs (l,r) , where P is in $V^* \times V^*$. These pairs are called production rules ($l \rightarrow r$)

S is the starting symbol (an element of V_H)

Generative Grammars and Languages

A Language $L(G)$ is the set of strings made only with terminal symbols, that can be generated from the starting symbol S

$$L(G) = \{ w \mid S \xrightarrow{*} w, w \in V_E^* \}$$

Generative Grammars and Languages (Ex 1)

$$G = \langle \{S\}, \{a, b\}, \{S \rightarrow aS, S \rightarrow b\}, S \rangle$$

$$L(G) = b, ab, aab, aaab, \dots$$

$$= \{ a^n b \mid n \geq 0 \}$$

Types of Generative Grammars

Chomsky defines four types of grammars, depending on the restrictions imposed on P:

0 **unrestricted**

1 **Context-Sensitive**

$$n_1 A n_2 \rightarrow n_1 w n_2$$

where $n_1, n_2 \in (V_H \cup V_E)^*$, $A \in V_H$

and $w \in (V_H \cup V_E)^+$ except $S \rightarrow \varepsilon$

2 **Context-Free**

$A \rightarrow w$ where $A \in V_H$ and $w \in (V_H \cup V_E)^*$

3 **Regular - each production is of the type**

$A \rightarrow a$ or $A \rightarrow aB$ where $A, B \in V_H$ and $a \in V_E$

Generative Grammars (Ex 2)

$S \rightarrow NP (AUX) VP \mid AUX NP VP$

$NP \rightarrow (ART) (ADJ^*) N (PP^*)$

$VP \rightarrow V (NP) (PP^*)$

$PP \rightarrow PREP NP$

$ART \rightarrow the \mid a \mid an$

$N \rightarrow John \mid Mary \mid book \mid \dots$

...

See Augmented Finite State Transition Network

Notation for Sampled Membership Functions

Let \mathbf{X} be the N-dimensional vector $[X(u_1), \dots, X(u_n)]$

All the vector operations used in the definitions are to be computed element-wise, so that the notation $f(\mathbf{X})$ is equivalent to applying $f(\cdot)$ to each element $X(u_i)$, e.g.:

$$\mathbf{X}^2 = [X(u_1)^2, \dots, X(u_n)^2]$$

$$1 - \mathbf{X} = [1 - X(u_1), \dots, 1 - X(u_n)]$$

$$\text{Max} \{ \mathbf{X}, \mathbf{Y} \} = [\max\{X(u_1), Y(u_1)\}, \dots, \max\{X(u_n), Y(u_n)\}]$$

Semantics for Hedges and Connectors

Defined Sampled Membership Functions

<i>X and Y</i>	⇒ $\min \{X, Y\}$
<i>X or Y</i>	⇒ $\max \{X, Y\}$
<i>not X</i>	⇒ $1 - X$
<i>very X</i>	⇒ $\text{CONTR}(X) = X^2$
<i>extremely X</i>	⇒ $\text{CONTR}(\text{CONTR}(X)) = X^4$
<i>more or less X</i>	⇒ $\text{DIL}(X) = X^{0.5}$
<i>indeed X</i>	⇒ $\text{INT}(X)$
<i>hardly X</i>	⇒ $\text{FUZ}(X)$
<i>younger than X</i>	⇒ $\text{SM}(X)$
<i>older than X</i>	⇒ $\text{GR}(X)$
<i>in between (X)(Y)</i>	⇒ $\text{NORM}\{\text{GR}(X) \text{ and } \text{SM}(Y)\}$
	⇒ $\text{NORM}\{\text{not } \text{SMEQ}(X) \text{ and } \text{not } \text{GREQ}(Y)\}$
<i>from to (X)(Y)</i>	⇒ $\text{GREQ}(X) \text{ and } \text{SMEQ}(Y)$

Semantics for Hedges and Connectors

Defined Sampled Membership Functions

where:

NORM(X)	= $X / \text{Max}_i X(u_i)$	
INT(X)	= $2X^2$	when $X < 0.5$
	= $1 - 2(1 - X)^2$	when $X > 0.5$
FUZ(X)	= $0.5 - 2(0.5 - X)^2$	when $X < 0.5$
	= $0.5 + 2(X - 0.5)^2$	when $X > 0.5$
SM(X)	= <u>not</u> GREQ(X)	
GR(X)	= <u>not</u> SMEQ(X)	
GREQ(X)	= X for $u_i < u^*$	
	= 1 for $u_i > u^*$	
SMEQ(X)	= 1 for $u_i < u^*$	
	= X for $u_i > u^*$	

Notes:

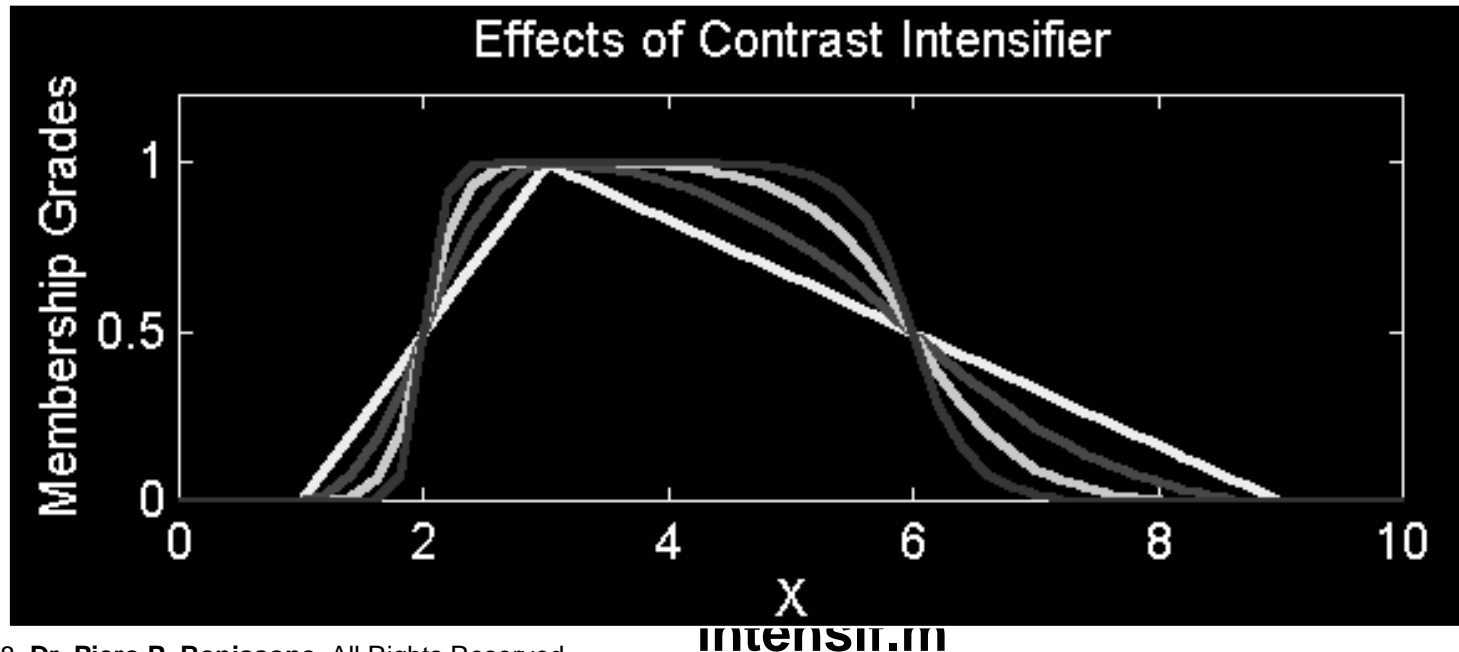
- $u^* = \text{Min} \{u_i \mid X(u_i) = 1\}$ (leftmost value of U with membership = 1)
- If needed, normalize $X(u)$ before using SMEQ(X) or GREQ(X)

Operations on Linguistic Values

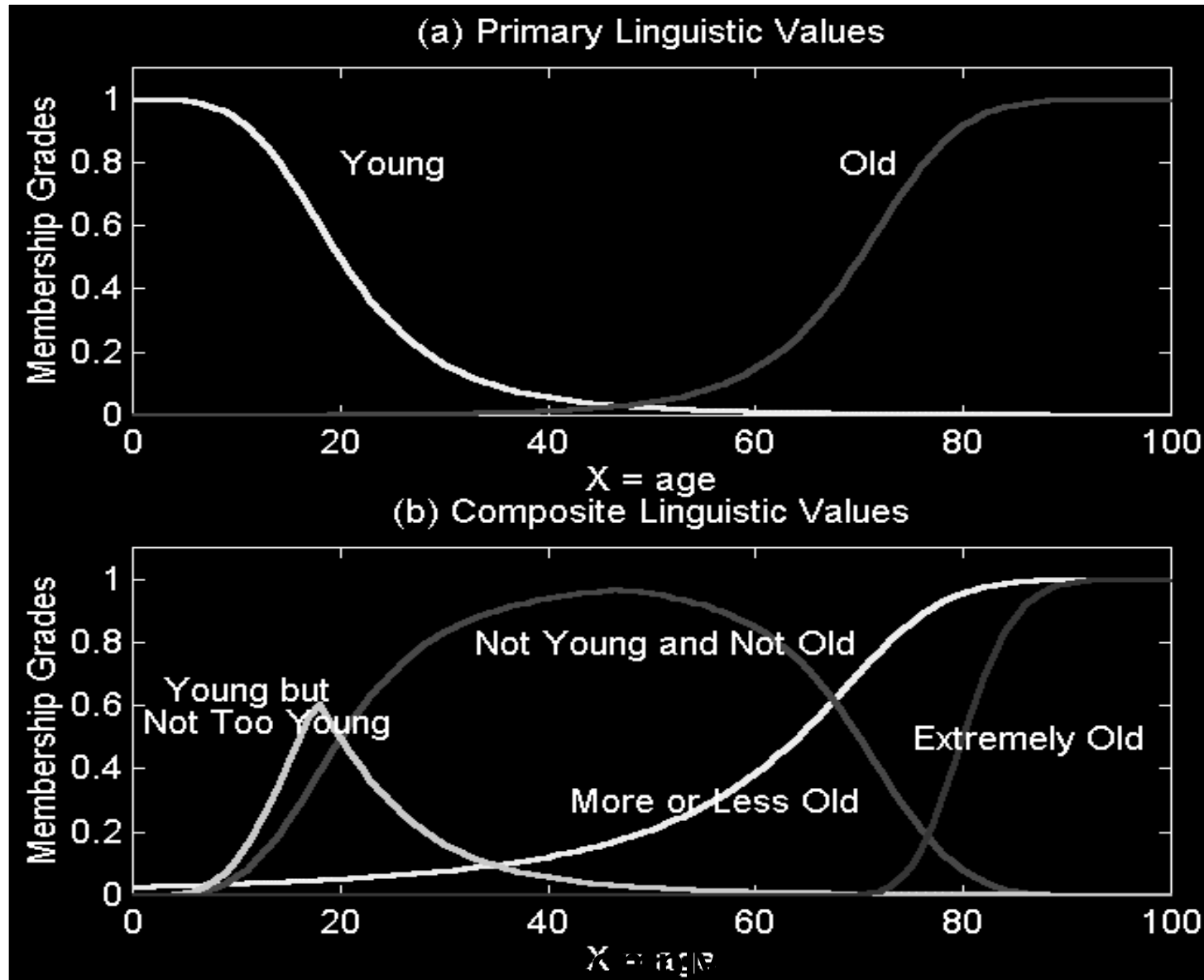
Concentration: $\Rightarrow CON(X) = X^2$

Dilation: $\Rightarrow DIL(X) = X^{0.5}$

Contrast intensification: $\Rightarrow INT(X) = \begin{cases} 2X^2, & 0 \leq \mu_X(x) \leq 0.5 \\ 1 - 2(1 - X)^2, & 0.5 \leq \mu_X(x) \leq 1 \end{cases}$



Linguistic Values (Terms)



Notation for Parametrized Membership Functions (for Optional steps)

- Each *primary term* is now represented by a 4-parameter list.
- Let $X = (a, b, \alpha, \beta)$ and $Y = (c, d, \gamma, \delta)$.
- Let the universe of discourse be the interval $[w, z]$.

Note:

- This parametric representation is valid only for *convex, normal* Fuzzy Numbers.
- For this optional step, disregard the terms derivable from non-terminal U, because neither *no age* nor *unknown age* are normal fuzzy numbers.
- Therefore, they cannot be represented by the four parameter representation.

Operations (and) for Parametrized Membership Functions

$$X \text{ and } Y \ \& = (p, q, L_a, R_a)$$

$$p = \max(a, c)$$

$$q = \min(b, d)$$

$$L_a = (c + \alpha) - \min(a, c)$$

$$= (a + \gamma) - \min(a, c)$$

$$= \alpha$$

$$= \gamma$$

$$= \alpha = \gamma$$

$$\text{if } (a - \alpha) > (c - \gamma)$$

$$\text{if } (a - \alpha) < (c - \gamma)$$

$$\text{if } (a - \alpha) = (c - \gamma), a > c$$

$$\text{if } (a - \alpha) = (c - \gamma), a < c$$

$$\text{if } (a - \alpha) = (c - \gamma), a = c$$

$$R_a = (d + \delta) - \min(b, d)$$

$$= (b + \beta) - \min(b, d)$$

$$= \delta$$

$$= \beta$$

$$= \beta = \delta$$

$$\text{if } (b + \beta) > (d - \delta)$$

$$\text{if } (b + \beta) < (d - \delta)$$

$$\text{if } (b + \beta) = (d - \delta), b > d$$

$$\text{if } (b + \beta) = (d - \delta), b < d$$

$$\text{if } (b + \beta) = (d - \delta), b = d$$

Possible Problems with the and Operator and Normality

If $\max(a, c) \leq \min(b, d)$, the result is still a *normal* set.

Otherwise, normalize the result by finding the intersection of the two slopes, and scaling the membership value of that point b (making it equal to one):

$$X \text{ and } Y = \text{NORM}(p, q, L_a, R_a) = (x, x, L_n, R_n)$$

where

$$x = (q R_a + p L_a) / (L_a + R_a)$$

$$L_n = x - (p - L_a)$$

$$R_n = (q + R_a) - x$$

Operations (or) for Parametrized Membership Functions

$$X \text{ or } Y = (r, s, L_o, R_o)$$

$$r = \min(a, c)$$

$$s = \max(b, d)$$

$$L_o = (a + \gamma) - \max(a, c)$$

$$= (c + \alpha) - \max(a, c)$$

$$= \gamma$$

$$= \alpha$$

$$= \alpha = \gamma$$

$$\text{if } (a - \alpha) > (c - \gamma)$$

$$\text{if } (a - \alpha) < (c - \gamma)$$

$$\text{if } (a - \alpha) = (c - \gamma), a > c$$

$$\text{if } (a - \alpha) = (c - \gamma), a < c$$

$$\text{if } (a - \alpha) = (c - \gamma), a = c$$

$$R_o = (b + \beta) - \max(b, d)$$

$$= (d + \delta) - \max(b, d)$$

$$= \delta$$

$$= \beta$$

$$= \beta = \delta$$

$$\text{if } (b + \beta) > (d - \delta)$$

$$\text{if } (b + \beta) < (d - \delta)$$

$$\text{if } (b + \beta) = (d - \delta), b > d$$

$$\text{if } (b + \beta) = (d - \delta), b < d$$

$$\text{if } (b + \beta) = (d - \delta), b = d$$

Possible Problems with the or Operator and Convexity

If $\max(a, c) \leq \min(b, d)$ the result is still a *convex* set.

Otherwise, maintain both fuzzy sets as elements of a data structure called UNION-LIST, i.e.,

$X \text{ or } Y = \text{UNION-LIST}(X, Y)$.

Any subsequent operation on the result of $X \text{ or } Y$ can be distributed to each element in UNION-LIST.

If $\max(a, c) \leq \min(b, d)$, the result is still a *normal* set.

Remaining Definitions

not X

$$\Rightarrow \text{UNION-LIST}(\text{SM}(X), \text{GR}(X))$$

very X

$$\Rightarrow \text{CONTR}(X) = (a + \alpha/5, b - \beta/5, \alpha, \beta)$$

$$\text{if } (a + \alpha/5) \leq (b - \beta/5)$$

$$\Rightarrow \text{NORM}(\text{CONTR}(X)) \quad \text{otherwise}$$

extremely X

$$= \text{CONTR}(\text{CONTR}(X)) =$$

$$(a + \alpha/2.5, b - \beta/2.5, \alpha, \beta)$$

$$\text{if } (a + \alpha/2.5) \leq (b - \beta/2.5)$$

$$\Rightarrow \text{NORM}(\text{CONTR}(\text{CONTR}(X))) \quad \text{otherwise}$$

more or less X

$$\Rightarrow \text{DIL}(X) = (a - \alpha/5, b + \beta/5, \alpha, \beta)$$

indeed X

$$\Rightarrow \text{INT}(X) = (a - \alpha/4, b + \beta/4, \alpha/2, \beta/2)$$

hardly X

$$\Rightarrow \text{FUZ}(X) = (a + \alpha/4, b - \beta/4, \alpha/2, \beta/2)$$

$$\text{if } (a + \alpha/4) < (b - \beta/4)$$

$$\Rightarrow (a + \alpha(b-a)/(\alpha + \beta), (b - \beta(b-a)/(\alpha + \beta), \alpha/2, \beta/2)$$

$$\text{otherwise}$$

Remaining Definitions (cont.)

younger than X \Rightarrow $SM(X) = (w, a - \alpha, 0, \alpha)$

older than X \Rightarrow $GR(X) = (b + \beta, z, \beta, 0)$

from to (X)(Y) \Rightarrow (a, d, α, δ)

in between (X)(Y) \Rightarrow from to $(GR(X))(SM(Y))$
 \Rightarrow $(b + \beta, c - \gamma, \beta, \gamma)$

if necessary, normalize X before applying SM(X) or GR(X).

Defuzzification Methods

- **COG (Center of Gravity)**
- **MOM (Mean of Maxima)**

Defuzzification Methods (COG)

- Center-of-Gravity Method

The center of gravity (or centroid of the distribution) is obtained by computing the *first moment* of $X(z) \subseteq U$:

$$COG\{X(z)\} = \frac{\int_U zX(z)dz}{\int_U X(z)dz} = \frac{\int_U zX(z)dz}{|X|}$$

For the discrete and finite case, where $U=\{z_1, z_2, \dots, z_n\}$, this expression reduces to:

$$COG\{X(z)\} = \frac{\sum_{i=1}^n z_i X(z_i)}{\sum_{i=1}^n X(z_i)} = \frac{\sum_{i=1}^n z_i X(z_i)}{|X|}$$

Defuzzification Methods (MOM)

- Mean Of Maxima

The MOM method obtains the *mode* of the distribution. i.e the value of z in which the membership distribution $X(z) \subseteq U$ achieves its maximum.

If the maximum is obtained in multiple points, the output is the average of such set of points.

We first define the crisp subset V (for the continuous and discrete cases) as :

$$V(z) = \{z \in U \mid X(z) \geq X(t), \forall t \in U\}$$

$$V(z_i) = \{z_i \in U \mid X(z_i) = \text{SUP}_{j=1, \dots, n} X(z_j)\}$$

Defuzzification Methods (MOM)

Then the MOM for the continuous case is :

$$MOM\{X(z)\} = \frac{\int_V zX(z)dz}{\int_V X(z)dz} = \frac{\int_V zdz}{\int_V dz} = \frac{\int_V zdz}{|V|}$$

And for the discrete and finite case where $U=\{z_1, z_2, \dots, z_n\}$, this expression reduces to the average of the points in V :

$$MOM\{X(z_i)\} = \frac{\sum_{z_i \in V} z_i X(z_i)}{\sum_{z_i \in V} X(z_i)} = \frac{\sum_{z_i \in V} z_i}{\sum_{z_i \in V} 1} = \frac{\sum_{z_i \in V} z_i}{|V|}$$