

# Introduction to fuzzy logic

## *Notes of class*

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## 1 Fuzzy logic

linguistic variable of the following elements:

- $X$  variable name
- $T(x)$  set of terms
- $\varpi$  universe of discourse
- $G$  syntax rules
- $M$  semantic rules

degree of belonging  $x \in A = \mu_A(x) = 1$ , otherwise  $x \notin A = \mu_A(x) = 0$ . now we can said that  $\mu(x) \in [0, 1]$  the degree of belonging of  $x$  to a any set is a number in the closed interval  $[0, 1]$ .

we can said that  $\varpi$  discrete (countable), a set could be written as  $\tilde{A} = \{\frac{\mu_{\tilde{A}}(x_1)}{x_1} + \frac{\mu_{\tilde{A}}(x_2)}{x_2} + \dots + \frac{\mu_{\tilde{A}}(x_n)}{x_n}\}$  or in a compact way in  $\tilde{A} = \{\sum_i^n \frac{\mu_{\tilde{A}}(x_i)}{x_i}\}$ . in continuous  $\varpi$  (uncountable) we can write  $\tilde{A} = \{\int \frac{\mu_{\tilde{A}}(x)}{x}\}$ .

Until now the major point is the breaking the law of "tercero excluido".

### 1.1 Support

The support of a fuzzy set  $S()$ , those elements that have a degree of belonging greater than zero  $\mu_{\tilde{A}}(x) > 0$ .

$$S(A) = \{x | \mu_A(x) > 0\} \quad (1)$$

remember that a set is considered regarding a  $\varpi$ .

#### 1.1.1 Singleton

The support is defined only for one value of the all domain for example  $x_0$  and  $\mu_A(x_0) = 1$ .

### 1.2 $\alpha$ cut-off

suppose that  $A$  is fuzzy set of  $\varpi$  then by  $\alpha \in [0, 1]$   $A_\alpha$  is:

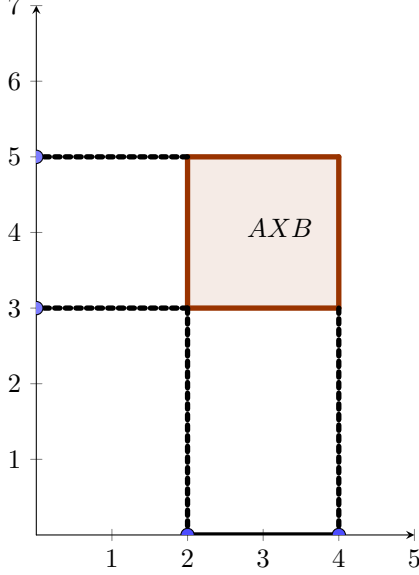
$$A_\alpha = \begin{cases} x \in \varpi | \mu_A(x) \geq \alpha, & \text{if } \alpha \geq 0 \\ S(A), & \text{if } \alpha = 0 \end{cases} \quad (2)$$

## 2 Cartesian product

by two sets  $A$  and  $B$  the Cartesian product is defined as all possible pairs formed with the sets.

$$AXB : \{(a, b) | a \in A \text{ and } b \in B\} \quad (3)$$

for instance  $A = \{a_1, a_2\}$  and  $B = \{b_1, b_2, b_3\}$  therefore  $AXB = \{(a_1, b_1), (a_1, b_2), (a_1, b_3), (a_2, b_1), (a_2, b_2), (a_2, b_3)\}$ . when you think in two intervals for instance  $A = [2, 4]$  and  $B = [3, 5]$  we can interpreted as region in  $\mathbb{R}^2$ .



## 3 Orthogonal extension

we have two universes of discourses  $\varpi_1$  and  $\varpi_2$  we can think in the cartesian product of them  $\varpi_1 X \varpi_2$ .

$\varpi_1/\varpi_2$	$\tilde{A}$	$a_1$	$a_2$	$a_3$
$b_1$	0.3	$(b_1, a_1)$	$(b_1, a_2)$	$(b_1, a_3)$
$b_2$		$(b_2, a_1)$	$(b_2, a_2)$	$(b_2, a_3)$
$b_3$	0.9	$(b_3, a_1)$	$(b_3, a_2)$	$(b_3, a_3)$

now the extension is:

$\varpi_1/\varpi_2$	$\tilde{A}$	$a_1$	$a_2$	$a_3$
$b_1$	0.3	0.3	0.3	0.3
$b_2$	0	0	0	0
$b_3$	0.9	0.9	0.9	0.9

we can establish that  $\mu_{\tilde{A}}(x, y) = \mu_{\tilde{A}}(x)$

## 4 Orthogonal projection

This problem is inverse to the problem defined previously.

$\varpi_1/\varpi_2$	$a_1$	$a_2$	$a_3$
$b_1$	0.8	0.9	1
$b_2$	0	0	0.8
$b_3$	0.4	0.5	0.7

this is only take the major value for each element in the different combinations, for instance for  $b_1 = 1$  and for  $a_1 = 0.8$ . We can establish that  $\mu_{\tilde{A}}(x) = \max \mu_A(x, y)$ .

ua	ub	U
0	0	0
0	1	1
1	0	1
1	1	1

## 4.1 basic operations

the Union operation with membership functions.

in summary we can use  $\mu_{A \cup B}(x) = \max(\mu_A(x), \mu_B(x))$ .

make the table for intersection and complement.

## 4.2 T-norm and S-norm

remember the unitary cubic

## 4.3 fuzzy relations

We have  $A \subset \varpi_1$  and  $B \subset \varpi_2$  a  $R$  is a subset of pairs, the fuzzy relation  $\tilde{R} = \frac{\mu_{a_1 b_1}}{(a_1 b_1)} + \dots +$ .

in mathematical terms we can say that a fuzzy relation between two fuzzy sets is a subset of the Cartesian product among the universe of discourses.  $\tilde{R} = \tilde{A}X\tilde{B}$  is a subset of the cartesian product of the universe of discourses, and the degree of belonging is defined as  $\mu_{\tilde{R}}(x, y) = \min(\mu_{\tilde{A}}(x), \mu_{\tilde{B}}(y))$ . Remember that we can use any t-norm.

### 4.3.1 Compound relations

Suppose  $\tilde{A}\tilde{R}_1\tilde{B}$  and  $\tilde{B}\tilde{R}_2\tilde{C}$  how we can establish a relation between  $\tilde{A}$  and  $\tilde{C}$ .

$$\tilde{R}_1 = \begin{bmatrix} 0.3 & 0.9 & 0.1 \\ 0.1 & 0 & 0.4 \\ 0.8 & 0.6 & 0.7 \end{bmatrix}, \tilde{R}_2 = \begin{bmatrix} 0.9 & 0.5 & 0.1 \\ 0.3 & 0.7 & 0.3 \\ 0.4 & 0.9 & 0.5 \end{bmatrix}$$

Think in the composition of two functions  $f \circ g(x) = f(g(x))$  therefore  $\tilde{R}_1 \circ \tilde{R}_2$  is defined in the a similar way with a multiplication matrix take the  $i$ -th row of  $\tilde{R}_1$  and compare with the  $i$ -th column of  $\tilde{R}_2$  instead of apply the multiplication uses the min operator for instance the 1 row of  $\tilde{R}_1$  is  $[\tilde{r}_{111}, \tilde{r}_{112}, \tilde{r}_{113} \dots \tilde{r}_{11n}]$  and the 1 column of  $\tilde{R}_2$  is  $[\tilde{r}_{211}, \tilde{r}_{221}, \tilde{r}_{231} \dots \tilde{r}_{2m1}]^T$ . the first element of the composition matrix will be  $\max[\min(\tilde{r}_{111}, \tilde{r}_{211}), \min(\tilde{r}_{112}, \tilde{r}_{221}) \dots \min(\tilde{r}_{11n}, \tilde{r}_{2m1})]$ . applied the steps over all rows and columns respectively.

$$\tilde{R}_1 \circ \tilde{R}_2 = \begin{bmatrix} 0.4 & 0.7 & 0.3 \\ 0.4 & 0.4 & 0.5 \\ 0.8 & 0.7 & 0.5 \end{bmatrix}$$

## 5 Linguistic variables

## 6 fuzzy inference

## 7 Fuzzy C-means