

Computational Intelligence Lecture 10:Fuzzy Sets

Farzaneh Abdollahi

Department of Electrical Engineering

Amirkabir University of Technology

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Classical Set

Fuzzy Set Basic Concepts in Fuzzy Sets Operations on Fuzzy Sets Fuzzy Complement Fuzzy Union Fuzzy Intersection Averaging Operator

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Classical Set

- ► A classical (crisp) set A in the universe of discourse U: can be defined by
 - List method: listing all of its members
 - Rule method: specifying the properties that must be satisfied by the members of the set

 $A = \{x \in U | x \text{ meets some conditions} \}$

 Membership method: introduces a zero-one membership function (also called characteristic function, discrimination function, or indicator function)

$$\mu_{A} = \begin{cases} 1 & \text{if } x \in A \\ 0 & \text{if } x \notin A \end{cases}$$

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Example: cars in Tehran

- ► The universe of discourse *U*.
- Set A is the cars with 4 cylinders:

$$A = \{x \in U | x \text{ has 4 cylinders} \} OR$$

$$\mu_A = \begin{cases} 1 & \text{if } x \in U \& x \text{ has 4 cylinders} \\ 0 & \text{if } x \in U \& x \text{ does not have 4 cylinder} \end{cases}$$

- Set D is the car made in Iran
- BUT the distinction between an Iranian car and a non-Iranian a car is not crisp:(
 - Most of them are not completely made in Iran
- So what should we do??!!





Fuzzy Set

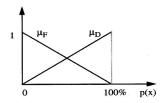
- some sets do not have clear boundaries.
- ► Fuzzy set: in a universe of discourse U is characterized by a membership function µ_A(x) that takes values in the interval [0, 1].
- In classical sets the membership function of a classical set can only take zero and one
- ▶ In fuzzy set the membership function is a continuous function with range [0, 1].
- A fuzzy set A in U is represented by:
 - a set of ordered pairs of a generic element x and its membership value: $A = \{(x, \mu_A(x)) | x \in U\}$
 - for continuous U: $A = \int_U \mu_A(x)/x$.
 - for discrete U: $\mu_A(x)$: $A = \sum_U \mu_A(x)/x$
 - \int and \sum do not represent integral and summation.
 - ► They denote collection of all points x ∈ U with the associated membership function µ_A(x)



Example: cars in Tehran (Cont'd)

- D: The set "Iranian cars in Iran,"
- $\mu_D = p(x)$
 - p(x) is the percentage of the parts of car x made in Iran
 - it takes values from 0% to 100%.
- F: The set "non-Iranian cars in Iran,"

$$\blacktriangleright \ \mu_F(x) = 1 - p(x)$$



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- Different membership functions can be defined to characterize the same description.
- ► The membership functions are not fuzzy, themselves.
- They are precise mathematical functions.
- ► Fuzzy sets are used to defuzzify the world.
- How to determine the membership functions?
 - Formulate human knowledge
 - Usually, gives a rough formula of the membership function
 - fine-tuning is required.
 - Data collected from various sensors
 - specify the structures of the membership functions and then fine-tune the parameters based on the data.
- A fuzzy set has a one-to-one correspondence with its membership function

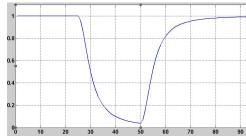
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Fuzzy Set



Example: Old and Young [?]

- U is in the interval of [0, 100]
- ► young = $\int_{0}^{25} \frac{1}{x} + \int_{25}^{100} \left(1 + \left(\frac{x-25}{5}\right)^{2}\right)^{-1} / x$ ► old = $\int_{50}^{100} \left(1 + \left(\frac{x-50}{5}\right)^{-2}\right)^{-1} / x$



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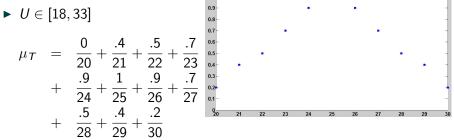
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Fuzzy Set



Example: A Digital Thermometer

 T: the set for desirable temperature



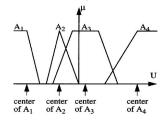
Basic Concepts in Fuzzy Sets

Support of a fuzzy set A in the universe of discourse U is a crisp set that contains all the elements of U that have nonzero membership values in A: supp_A = {x ∈ U |µ_A > 0}

- ► In the digital thermometer example: supp_A = [21, 30]
- empty fuzzy set: support is empty
- fuzzy singleton: support is a single point

Center of a fuzzy set:

- If the mean value of all points at which the membership function of the fuzzy set achieves its maximum value is finite, then this mean value is the center
- If the mean value equals positive (negative) infinite, then the center is the smallest (largest) among all points that achieve the maximum membership value.





- Crossover point of a fuzzy set: the point in U whose membership value in A equals 0.5.
- Height of a fuzzy set: the largest membership value attained by any point.
 - Normal fuzzy set: the height of fuzzy set equals to one (digital thermometer).
- α-cut of a fuzzy set A a crisp set A_α contains all the elements in U that have membership values in A greater than or equal to α:
 A_α = {x ∈ U | μ_A(x) ≥ α}
 - In digital thermometer for $\alpha = 0.7$, $T_{\alpha} = [23, 24, 25, 26, 27]$
 - A fuzzy set A is convex iff its α -cut is a convex set for $\forall \alpha \in (0, 1]$.

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Fuzzy Set

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 - A fuzzy set A is convex iff its α -cut is a convex set for $\forall \alpha \in (0, 1]$.
 - ► In Euclidean space, an object is convex if for every pair of points within the object, every point on the straight line segment that joins them is also within the object.





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 - A fuzzy set A is convex iff its α -cut is a convex set for $\forall \alpha \in (0, 1]$.
 - ► Let *C* be a set in a real or complex vector space. *C* is convex if, $\forall x, y \in C$ and all $\lambda \in [0, 1] \rightsquigarrow, \lambda x + (1 - \lambda)y \in C$

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 - A fuzzy set A is convex iff its α-cut is a convex set for ∀α ∈ (0,1].
 - ► Lemma: A fuzzy set $A \in \mathcal{R}^n$ is convex iff $\mu_A[\lambda x_1 + (1 - \lambda)x_2] \ge \min[\mu_A(x_1), \mu_A(x_2)] \quad \forall x_1, x_2 \in \mathcal{R}^n, \lambda \in [0, 1].$

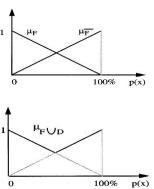
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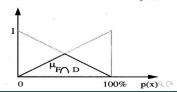


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Operations on Fuzzy Sets

- ► Sets *F* and *D* are equal iff $\mu_F(x) = \mu_D(x), \forall x \in U$
- ► Set *D* contains set *F* (*F* ⊂ *D*), iff $\mu_F(x) \le \mu_D(x), \forall x \in U$
- Complement of F is a fuzzy set F ∈ U whose membership function is μ_F(x) = 1 − μ_F(x)
- ▶ Union of sets *F* and *D* (*F* \cup *D*) is a fuzzy set in *U*: $\mu_{F \cup D} = \max[\mu_F(x), \mu_D(x)]$
 - F ∪ D is the smallest fuzzy set containing both F and D.
- ▶ Intersection of *F* and *D* (*F* ∩ *D*) is a fuzzy set in $U:\mu_{F\cap D} = \min[\mu_F(x), \mu_D(x)]$
 - *F* ∩ *D* is the smallest fuzzy set contained by *F* and *D*.





Lecture 10

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► The De Morgan's Laws are true for fuzzy sets:

$$\overline{F \cup D} = \overline{F} \cap \overline{D}$$
$$\overline{F \cap D} = \overline{F} \cup \overline{D}$$

► For Iranian Cars example:

•
$$\mu_{F \cup D} = \begin{cases} \mu_D & \text{if } 0 \le p(x) \le 0.5 \\ \mu_F & \text{if } 0.5 \le p(x) \le 1 \end{cases}$$

• $\mu_{F \cap B} = \begin{cases} \mu_F & \text{if } 0 \le p(x) \le 0.5 \\ \mu_D & \text{if } 0.5 \le p(x) \le 1 \end{cases}$



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Further Operations

- An other difference between fuzzy sets and crisp sets:
 - ▶ for crisp sets only one type of operation is defined for complement, union, and intersection
 - for fuzzy sets, we can define several operations for them based on the given axioms.
- Why do we need different type of operations?
 - Some operations may not be satisfactory in some situations.





Fuzzy Complement

- Let c : [0,1] → [0,1] be a mapping that transforms the membership function of fuzzy set A into the membership function of the complement of A: c[µ_A(x)] = µ_Ā(x)
- It was defined: $c[\mu_A(x)] = 1 \mu_A$
- Let $a = \mu_A(x_1)$ and $b = \mu_A(x_2)$
- ▶ the function *c* is qualified as a complement if:
 - Axiom c1: c(0) = 1 and c(1) = 0 (boundary condition)
 - ▶ Axiom c2: $\forall a, b \in [0, 1]$, if a < b, then $c(a) \ge c(b)$ (nonincreasing condition)
 - an increase in membership value must result in a decrease or no change in membership value for the complement



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Fuzzy Complement

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- ▶ the function *c* is qualified as a complement if:
 - Axiom c1: c(0) = 1 and c(1) = 0 (boundary condition)
 - ► Axiom c2: ∀a, b ∈ [0,1], if a < b, then c(a) ≥ c(b) (nonincreasing condition)</p>
 - an increase in membership value must result in a decrease or no change in membership value for the complement
- Some types of fuzzy complement:
 - ▶ Sugeno class: $c_{\lambda}(a) = \frac{1-a}{1+\lambda a}, \quad \lambda \in (-1,\infty)$
 - $\lambda = 0 \rightsquigarrow$ basic fuzzy complement

► Yager class:
$$c_w(a) = (1 - a^w)^{1/w}$$
, $w \in (0, \infty)$

► w = 1 → basic fuzzy complement Computational Intelligence

- Let s : [0, 1] × [0, 1] → [0, 1] be a mapping that transforms the membership functions of fuzzy sets A and B into the membership function of the union of A and B, that s[μ_A(x), μ_B(x)] = μ_{A↓JB}.
- ▶ the function S to be qualified as an union

• Let
$$a = \mu_A(x)$$
 and $b = \mu_B(x)$

- Axiom s1.s(1,1) = 1, s(0,a) = s(a,0) = a (boundary condition).
- Axiom s2. s(a, b) = s(b, a) (commutative condition).
- ► Axiom s3. If a ≤ a' and b ≤ b', then s(a, b) ≤ s(a', b') (nondecreasing condition).

Fuzzy Set

- Axiom s4. s(s(a, b), c) = s(a, s(b, c)) (associative condition).
- Popular types of s-norm
 - ► Dombi class: $s_{\lambda}(a, b) = \frac{1}{1 + [(\frac{1}{a} 1)^{-\lambda} + (\frac{1}{b} 1)^{-\lambda}]^{-1/\lambda}}, \ \lambda \in (0, \infty)$
 - ► Dobios-Prade class: $s_{\alpha}(a,b) = \frac{a+b-ab-\min(a,b,1-\alpha)}{\max(1-a,1-b,\alpha)}, \alpha \in [0,1]$
 - ► Yager class: $s_w(a, b) = min[1, (a^w + b^w)^{1/w}], w \in (0, \infty)$

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Other type of s-norm

• Drastic sum:
$$s_{ds}(a, b) = \begin{cases} a & if b = 0 \\ b & if a = 0 \\ 1 & otherwise \end{cases}$$

• Einstein sum:
$$s_{es}(a,b) = \frac{a+b}{1+ab}$$

- Algebric sum: $s_{as}(a, b) = a + b ab$
- ► Theorem: For any s-norm s, that is for any function s : [0,1] × [0,1] → [0,1] that satisfies Axioms s1-s4, the smallest s-norm is maximum and the largest is drastic s-norm
- ► Proof:
 - Axioms s1 and s3 \Rightarrow $s(a, b) \ge s(a, 0) = a$
 - Axiom s2 \Rightarrow s(a, b) = s(b, a) \ge s(b, 0) = b

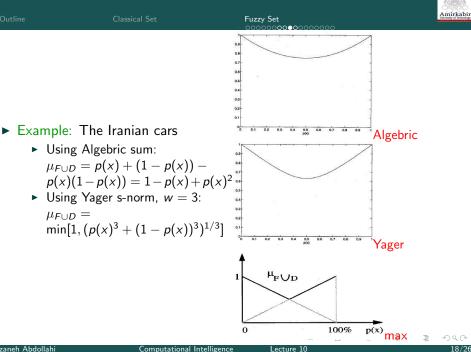
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 $(a, b) \geq max(a, b)$

► If
$$b = 0$$
, Axiom s1 \Rightarrow $s(a, b) = s(a, 0) = a \rightsquigarrow s(a, b) = s_{ds}(a, b)$

• If
$$a = 0$$
, Axiom s2 \Rightarrow $s(a, b) = s_{ds}(a, b)$

- If $a \neq 0 \& b \neq 0$, $s_{ds}(a, b) = 1 \ge s(a, b)$
- :: $s(a, b) \leq s_{ds}(a, b), \forall a, b \in [0, 1]$

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Lemma 1: For Dombi class s-norm and Drastic class s-norm it can be defined

$$\lim_{\lambda \to \infty} s_{\lambda}(a, b) = max(a, b)$$

 $\lim_{\lambda \to 0} s_{\lambda}(a, b) = s_{ds}(a, b)$

Lemma 2: For Yager class s-norm and Drastic class s-norm it can be defined

$$\lim_{w \to \infty} s_w(a, b) = max(a, b)$$
$$\lim_{w \to 0} s_w(a, b) = s_{ds}(a, b)$$

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Fuzzy Intersection- T-Norm

- Let t : [0, 1] × [0, 1] → [0, 1] be a mapping that transforms the membership functions of fuzzy sets A and B into the membership function of the union of A and B, that t[μ_A(x), μ_B(x)] = μ_{A∩B}.
- ▶ the function T to be qualified as an intersection
 - Axiom t1.t(0,0) = 0, t(1,a) = t(a,1) = 1 (boundary condition).
 - Axiom t2. t(a, b) = t(b, a) (commutative condition).
 - ► Axiom t3. If a ≤ a' and b ≤ b', then t(a, b) ≤ t(a', b') (nondecreasing condition).
 - Axiom t4. t(t(a, b), c) = t(a, t(b, c)) (associative condition).
- Popular types of t-norm
 - ► Dombi class: $t_{\lambda}(a, b) = \frac{1}{1 + [(\frac{1}{a}-1)^{\lambda} + (\frac{1}{b}-1)^{\lambda}]^{1/\lambda}}, \ \lambda \in (0, \infty)$
 - ► Dobios-Prade class: $t_{\alpha}(a, b) = \frac{ab}{\max(a, b, \alpha)}, \alpha \in [0, 1]$
 - Yager class:

 $t_w(a,b) = 1 - min[1,((1-a)^w + (1-b)^w)^{1/w}], w \in (0,\infty)$

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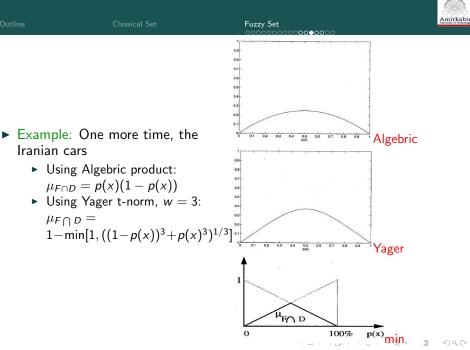
Other type of t-norm

• Drastic product:
$$t_{ds}(a, b) = \begin{cases} a & if \ b = 1 \\ b & if \ a = 1 \\ 0 & otherwise \end{cases}$$

- Einstein product: $t_{ep}(a,b) = \frac{ab}{2-(a+b-ab)}$
- Algebric product: $t_{ap}(a, b) = ab$
- ► Theorem: For any t-norm t, that is for any function t : [0,1] × [0,1] → [0,1] that satisfies Axioms t1-t4, the largest t-norm is minimum and the smallest is drastic t-norm
- prove it.
- Lemma 3: For Dombi class t-norm and Drastic class t-norm it can be defined

$$\lim_{\lambda \to \infty} t_{\lambda}(a, b) = \min(a, b)$$
$$\lim_{\lambda \to 0} t_{\lambda}(a, b) = t_{dp}(a, b)$$

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If the s-norm s(a, b), t-norm t(a, b) and fuzzy complement c(a) satisfy the following equation, they form an associated class (DeMorgan's Law)

$$c[s(a,b)] = t[c(a),c(b)]$$

Example: Show that the Yager s-norm and t-norm with the basic complement are associated

•
$$c[s_w(a,b)] = 1 - \min[1, (a^w + b^w)^{1/w}]$$

• $t_w[c(a), c(b)] = 1 - \min[1, ((1 - 1 + a)^w + (1 - 1 + b)^w)^{1/w}]$

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tline	Classical Set	Fuzzy Set 000000000000000000000000000000000000	S
minimum drastic product Einstein product		maximum drastic sum Einstein sum	
algebraic product		algebraic sum	
$\int_{0}^{\text{Dombi t-norm}} \lambda \int_{0}^{\text{Dombi t-norm}} \lambda$	22	$ \bigcup_{\alpha \to -\infty}^{\text{Dombi s-norm}} \lambda \longrightarrow_{0}^{-\infty} $	
Yager t-norm	55	Yager s-norm	
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_{dp} (a,b) mit	n(a,b)	max(a,b) s _{ds} (a,b)	
intersection		union operators	

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Averaging Operator

- This operator fills the gap between min(a, b), and max(a, b)
- Some average operators:
 - Max-min average: $v_{\lambda}(a, b) = \lambda \max(a, b) + (1 \lambda) \min(a, b), \ \lambda \in [0, 1]$
 - Generalized means: $v_{\alpha}(a,b) = (\frac{a^{\alpha}+b^{\alpha}}{2})^{1/\alpha}, \ \alpha \in R, \ \alpha \neq 0$
 - Fuzzy and: $v_p(a, b) = pmin(a, b) + \frac{(1-p)(a+b)}{2}, \ p = \in [0, 1]$
 - Fuzzy or: $v_{\gamma}(a,b) = \gamma \max(a,b) + \frac{(1-\gamma)(a+b)}{2}, \ \gamma \in [0,1]$

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