

**School of Information Technology
Indian Institute of Technology, Kharagpur**

**IT 60108: Soft Computing Applications
Mid-Semester Examination
Spring, 2015-2016**

Maximum Marks: 80

Time: 2 hours

Instructions:

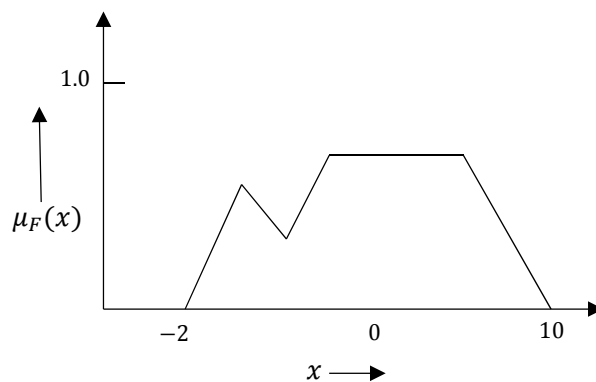
1. Attempt **ALL** questions. There is **NO** negative marking.
 2. The question paper consists of three pages and a total of **six** questions with some questions having sub-parts.
 3. Marks allotted for each question have been clearly mentioned, students are advised to give brief, precise and to-the point answers to all questions, strictly keeping the marks allotted in consideration.
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1. Which of the following is/ are fuzzy set(s). Justify your answer

[4 x 2=8]

- (a) $A = \{(x_1, 0), (x_2, 0), (x_3, 0), (x_4, 0)\}$ defined over a universe of discourse $X = \{x_1, x_2, x_3, x_4, x_5, x_6\}$
- (b) $B = \{(x, \mu_B(x)) \mid x \in Z, \text{ set of all integers and } \mu_B(x) = \frac{1-x}{1+x}\}$
- (c) $C = D \times E$ where D and E are two fuzzy sets and \times denotes the Cartesian product of two fuzzy sets.

(d)



2. Find the results of the fuzzy operations as instructed in the following:

[2 x 4=8]

(a) $R = A \times B$ where

$$A = \left\{ \frac{0.1}{x_1}, \frac{0.2}{x_3}, \frac{0.5}{x_5} \right\}$$

$$B = \left\{ \frac{0.6}{x_2}, \frac{0.8}{x_3}, \frac{1.0}{x_6} \right\}$$

(b) λ cut of the implication, *If x is A then y is B*, where

$$A = \{(x_1, 0.1), (x_2, 0.3), (x_3, 0.2)\}$$

$$B = \{(x_1, 0.2), (x_2, 0.3), (x_3, 0.1)\}$$

and both are defined over $X = \{x_1, x_2, x_3, x_4\}$ and $\lambda = 0.7$

3. Any road is characterized with two fuzzy linguistics WIDE and NARROW whereas a journey is characterized with two fuzzy linguistics HIGH RISK and LOW RISK. The universe of discourses of road and journey are $\{Large, Medium, Small\}$ and $\{High, Moderate, Low\}$, respectively.

A road and journey are associated with the following fuzzy implication:

If road is WIDE then driving is RISKY.

For the MG Road, it is given that

$$\text{Road is WIDE} = \left\{ \frac{0.3}{Large}, \frac{0.5}{Medium}, \frac{0.7}{Small} \right\}$$

$$\text{driving is RISKY} = \left\{ \frac{0.9}{High}, \frac{0.7}{Moderate}, \frac{0.6}{Low} \right\}$$

$$\text{Driving on M.G. road is RISKY} = \left\{ \frac{0.7}{High}, \frac{0.6}{Moderate}, \frac{0.5}{Low} \right\}$$

What is the fuzzy set that, M.G. road is NARROW ?

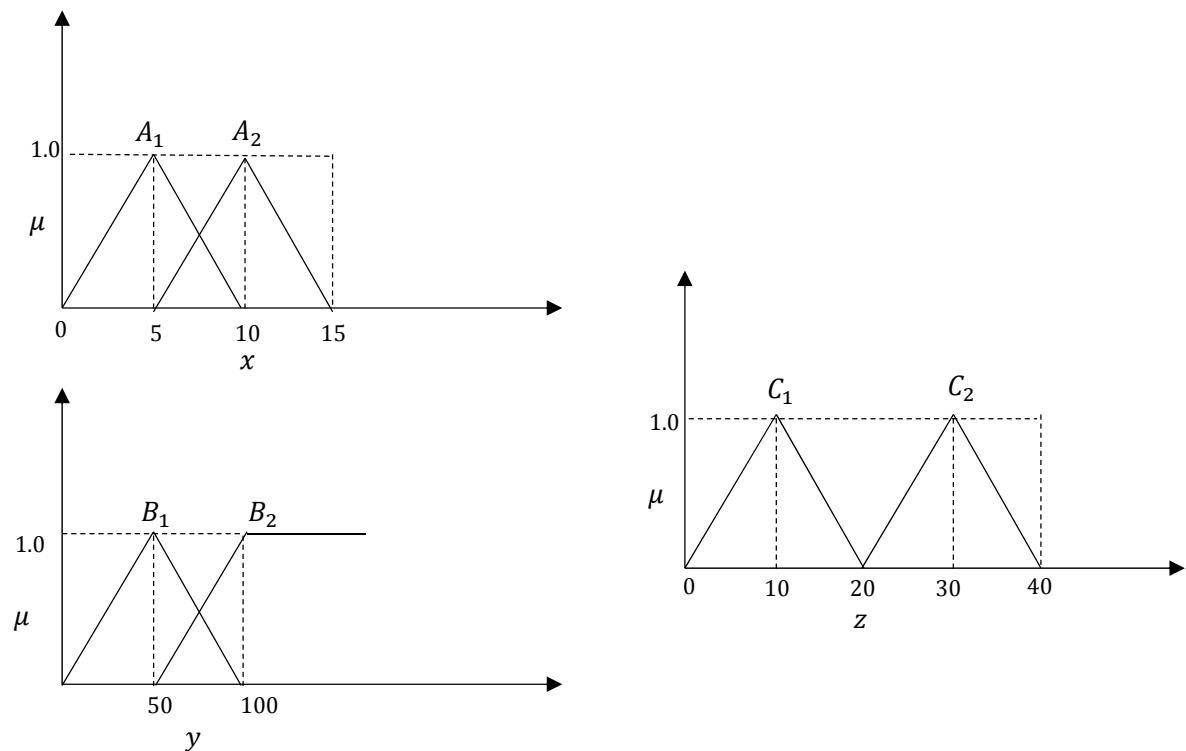
[10]

4. In a fuzzy controller for two input $x = 6$ and $y = 25$, two fuzzy rules are fired as below:

R_i : IF x is A_1 AND y is B_1 THEN z is C_1

R_j : IF x is A_2 AND y is B_2 THEN z is C_2

The fuzzy sets involved in R_i and R_j are known as given below:



- (a) Graphically show the combined output due to R_i and R_j for $x = 6$ and $y = 25$.
- (b) Apply COS (Center of Sum) defuzzification method to obtain the crisp value of the output when $x = 6$ and $y = 25$. **[2 x 7=14]**
5. Write brief answers to the following questions: **[4 x 4=16]**
- (a) Draw an ANN with the minimum number of perceptron which would classify input pattern 00, 01, 10 and 11 into two classes 0 and 1 following OR-logic.
- (b) The *Tanh – sigmoid* transfer function ϕ is defined as follows:
- $$\phi(I) = \frac{e^{\theta I} - e^{-\theta I}}{e^{\theta I} + e^{-\theta I}}$$
- where symbols bear usual meaning.
- Prove that $\frac{\partial \phi}{\partial I} = \theta(1 + \phi(I))(1 - \phi(I))$
- (c) State the Delta rule, which is usually followed in Back propagation algorithm. Is the rule applicable to any type of ANN?
- (d) Draw a symbolic diagram (also called bubble diagram) of a perceptron and clearly show the different unknown parameters in it.
6. Answer the following: **[8 x 3=24]**
- (a) Draw a MLFFNN having $l - m - n$ configuration. Clearly show its network parameters.
- (b) Suppose, $T = \langle T_o, T_I \rangle$ is a training data for the supervised learning of a $l - m - n$ network. If $I_i \in T_I$ is the i^{th} input applied to the network, then express the error at the k^{th} perceptron in the output layer. Also, obtain the expression for total error E due to all $I_i \in T_I$.
- (c) If w_{ij} and v_{jk} denote the weight of the link from an i^{th} neuron in the input layer to the j^{th} neuron in the hidden layer and from the j^{th} neuron in the hidden layer to the k^{th} neuron in the output layer, and e_K denotes the error of the k^{th} neuron in the output layer, then write down the chain rule of differentiation to calculate the following:
- $$\frac{\partial e_K}{\partial w_{ij}} \text{ and } \frac{\partial e_K}{\partial v_{ij}} .$$
- Clearly mention all the symbols used in the rule you have stated.