



Decision Making Problem for Medical Diagnosis Using Hexagonal Fuzzy Number Matrix

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Abstract

Fuzzy set theory plays a vital role in medical fields. There are varieties of models involving fuzzy matrices to deal with different complicated aspects of medical diagnosis. Fuzzy set theory is highly suitable and applicable for developing knowledge based system in medicine for the tasks of medical findings. The field of medicine and decision making are the most fruitful and interesting area of applications of fuzzy set theory. In this paper, we have applied the notion of Hexagonal fuzzy membership matrix in a medical diagnostic model. The advantage of this model is, if the patient-matrices are known, then it is possible to find which patient is suffering from what kind of disease. Most probably the fuzzy decision model in which overall ranking or ordering of different fuzzy sets are determined by using comparison matrix.

Keywords: Decision making, Hexagonal fuzzy number, Hexagonal fuzzy number matrix, Comparison matrix

1. Introduction

Fuzzy set theory introduced by professor Zadeh [12] in 1965 acts as a qualitative computational approach which describes uncertainty. As fuzzy decision making is a most important scientific, social and economic endeavour, there exist several major approaches within the theories of fuzzy decision making. The field of medicine and decision making are the most fruitful and interesting area of applications of fuzzy set theory. In real life situations, the imprecise nature of medical documentation and uncertain information gathered for decision making requires the use of fuzzy. Sánchez [10] formulated the diagnostic models involving fuzzy matrices representing the medical knowledge between the symptoms and diseases. Meenakshi and Kaliraja [7] have extended Sanchez's approach for medical diagnosis using the representation of interval valued fuzzy matrix. Elizabeth and Sujatha [2] have extended Sanchez's approach of medical diagnosis using the representation of triangular fuzzy membership matrix. In this paper, we have applied the notion of Hexagonal fuzzy membership matrix in a medical diagnostic model. The advantage of this model is, if the patient-matrices are known, then it is possible to find which patient is suffering from what kind of disease. Most probably the fuzzy decision model in which overall ranking or ordering of different fuzzy sets are determined by using comparison matrix.

2. Preliminaries

2.1 Definition (Fuzzy Set)

Let X be a set. A fuzzy set A on X is defined to be a function $A: X \rightarrow [0,1]$ or $\mu_A: X \rightarrow [0,1]$. Equivalently, a fuzzy set A is defined to be the class of objects having the following representation $A = \{(x, \mu_A(x)) : x \in X\}$ where $\mu_A: X \rightarrow [0,1]$, is a function called the membership function of A .

2.2 Definition (Fuzzy Number)

The fuzzy number A is a fuzzy set whose membership function $\mu_A(x)$ satisfies the following conditions:

1. $\mu_A(x)$ is piecewise continuous;
2. A fuzzy set A of the universe of discourse X is convex;
3. A fuzzy set of the universe of discourse X is called a normal fuzzy set if $\exists x_i \in X, \mu_A(x_i) = 1$.

3. Hexagonal Fuzzy Number

A fuzzy number is the normal Hexagonal fuzzy number is denoted by $(a_1, a_2, a_3, a_4, a_5, a_6)$ where $a_1 \leq a_2 \leq a_3 \leq a_4 \leq a_5 \leq a_6$ are real numbers and its membership function $\mu_A(x)$ is given below

$$\mu_{\tilde{A}}(x) = \begin{cases} 0 & \text{for } x < a_1 \\ \frac{1}{2} \left(\frac{x-a_2}{a_2-a_1} \right) & \text{for } a_1 \leq x \leq a_2 \\ \frac{1}{2} + \frac{1}{2} \left(\frac{x-a_2}{a_3-a_2} \right) & \text{for } a_2 \leq x \leq a_3 \\ 1 & \text{for } a_3 \leq x \leq a_4 \\ 1 - \frac{1}{2} \left(\frac{x-a_4}{a_5-a_4} \right) & \text{for } a_4 \leq x \leq a_5 \\ \frac{1}{2} \left(\frac{a_6-x}{a_6-a_5} \right) & \text{for } a_5 \leq x \leq a_6 \\ 0 & \text{for } x \geq a_6 \end{cases}$$

3.1 Definition (Hexagonal Fuzzy Number Matrix).

The elements of Hexagonal fuzzy number matrix are defined as $A = (a_{ij})_{m \times n}$, where $a_{ij} = (a_{ijL}, a_{ijM}, a_{ijM}, a_{ijN}, a_{ijN}, a_{ijU})$ is the ij^{th} element of fuzzy number matrix of A. Then $0 \leq a_{ijL} \leq a_{ijM} \leq a_{ijM} \leq a_{ijN} \leq a_{ijN} \leq a_{ijU} \leq 30$, where a_{ijL} is the lower bound, a_{ijM} , a_{ijN} is the moderate value, and a_{ijU} is the upper bound.

3.2 Definition (Hexagonal Fuzzy Number Matrix into Membership Function).

Let $\tilde{A} = (a_1, a_2, a_3, a_4, a_5, a_6)$ be a Hexagonal fuzzy number. Then $\mu_{\tilde{A}} = (a_1/30, a_2/30, a_3/30, a_4/30, a_5/30, a_6/30)$, where $0 \leq a_1 < a_2 < a_3 < a_4 < a_5 < a_6 < 30$. Thus $0 \leq a_1/30 < a_2/30 < a_3/30 < a_4/30 < a_5/30 < a_6/30 \leq 1$. Let Membership function of $a_{ij} = (a_{ijL}, a_{ijM}, a_{ijM}, a_{ijN}, a_{ijN}, a_{ijU})$ is defined as $(a_{ijL}/30, a_{ijM}/30, a_{ijM}/30, a_{ijN}/30, a_{ijN}/30, a_{ijU}/30)$ where $0 \leq a_{ijL} \leq a_{ijM} \leq a_{ijM} \leq a_{ijN} \leq a_{ijN} \leq a_{ijU} \leq 30$ $0 \leq a_{ijL}/30 \leq a_{ijM}/30 \leq a_{ijM}/30 \leq a_{ijN}/30 \leq a_{ijN}/30 \leq a_{ijU}/30 \leq 1$ is called a Hexagonal fuzzy number matrix into its membership function.

3.3 Arithmetic Mean of a Hexagonal Fuzzy Membership Number

Let $\mu_{\tilde{A}} = (a_1/30, a_2/30, a_3/30, a_4/30, a_5/30, a_6/30)$ be a Hexagonal fuzzy membership number, Then $AM(\mu_{\tilde{A}}) = \frac{a_1+a_2+a_3+a_4+a_5+a_6}{180}$

3.4 Definition (Relativity Function).

Let x and y be variables defined on a universal set X . The relativity function is denoted as $f(x/y)$ and is defined as

$$f(x/y) = \frac{\mu_y(x) (-) \mu_x(y)}{\max \{ \mu_y(x), \mu_x(y) \}}$$

Where $\mu_y(x)$ is the membership function of x with respect to y , $\mu_x(y)$ is the membership function of y with respect to x and $\max \{ \mu_y(x) - \mu_x(y) \}$ is maximum operation on Octagonal fuzzy number.

3.5 Definition (Comparison Matrix).

Let $A = \{x_1, x_2, x_3, \dots, x_{i-1}, x_i, x_{i+1}, \dots, x_n\}$ be a set of n variables defined on X . Then form a matrix of relativity values $f \left(\frac{x_i}{x_j} \right)$, where x_i 's for $i = 1$ to n , are n variables defined on an universe X .

The matrix $C = (C_{ij})$ is a square matrix of order n is called the comparison matrix

$$AM(f(x_i/y_j)) = \frac{AM(\mu_{x_j}(x_i) (-) \mu_{x_i}(x_j))}{AM(\max \{ \mu_{x_j}(x_i), \mu_{x_i}(x_j) \})}$$

4. Working Rule

Step 1: We can find membership function for Hexagonal fuzzy number matrix by using relativity function.

Step 2: We shall calculate all the relative values of the function.

Step 3: In the comparison matrix, the upper triangular part and lower triangular part are same

With opposite sign.

Step 4: In comparison matrix the maximum value in each row of the matrix will have the maximum possibility for ranking purpose.

5. Numerical Examples

Suppose there are 3 patients P_1, P_2, P_3 in a hospital with possible symptoms relating to the diseases like cancer, jaundice and dengue which are represented by D_1, D_2, D_3 . Here patient – disease fuzzy matrix can be represented as hexagonal fuzzy number matrix and a scale from 0 – 30.

The hexagonal fuzzy number matrix represents a patient-disease matrix form is given below

| | | | | | | | | | | | | | | |
|----------------------|--|----------------------|---------------------|---------------------|----------------------|----------------------|---------------------|---------------------|---------------------|---------------------|----------------------|-------------------|--|--|
| | D_1 | D_2 | D_3 | | | | | | | | | | | |
| $A =$ | P_1 | | | | | | | | | | | | | |
| | P_2 | | | | | | | | | | | | | |
| | P_3 | | | | | | | | | | | | | |
| | <table border="0" style="width: 100%; border-collapse: collapse;"> <tr> <td style="padding-right: 20px;">$(20,16,10,4,18,30)$</td> <td style="padding-right: 20px;">$(4,6,12,16,20,26)$</td> </tr> <tr> <td style="padding-right: 20px;">$(2,6,12,16,22,26)$</td> <td style="padding-right: 20px;">$(8,10,16,20,26,30)$</td> </tr> <tr> <td style="padding-right: 20px;">$(8,10,16,20,26,30)$</td> <td style="padding-right: 20px;">$(6,8,14,20,28,30)$</td> </tr> <tr> <td style="padding-right: 20px;">$(4,8,14,18,24,28)$</td> <td style="padding-right: 20px;">$(2,8,12,18,24,26)$</td> </tr> <tr> <td style="padding-right: 20px;">$(2,8,12,18,24,26)$</td> <td style="padding-right: 20px;">$(8,10,16,22,26,28)$</td> </tr> <tr> <td style="padding-right: 20px;">$(0,0,2,4,16,16)$</td> <td></td> </tr> </table> | $(20,16,10,4,18,30)$ | $(4,6,12,16,20,26)$ | $(2,6,12,16,22,26)$ | $(8,10,16,20,26,30)$ | $(8,10,16,20,26,30)$ | $(6,8,14,20,28,30)$ | $(4,8,14,18,24,28)$ | $(2,8,12,18,24,26)$ | $(2,8,12,18,24,26)$ | $(8,10,16,22,26,28)$ | $(0,0,2,4,16,16)$ | | |
| $(20,16,10,4,18,30)$ | $(4,6,12,16,20,26)$ | | | | | | | | | | | | | |
| $(2,6,12,16,22,26)$ | $(8,10,16,20,26,30)$ | | | | | | | | | | | | | |
| $(8,10,16,20,26,30)$ | $(6,8,14,20,28,30)$ | | | | | | | | | | | | | |
| $(4,8,14,18,24,28)$ | $(2,8,12,18,24,26)$ | | | | | | | | | | | | | |
| $(2,8,12,18,24,26)$ | $(8,10,16,22,26,28)$ | | | | | | | | | | | | | |
| $(0,0,2,4,16,16)$ | | | | | | | | | | | | | | |

Step 1:-

$$(A)_{mem} = \begin{bmatrix} (0.6,0.5,0.3,0.1,0.6,1) & (0.1,0.2,0.4,0.5,0.6,0.8) \\ (0.06,0.2,0.4,0.5,0.7,0.8) & \\ (0.26,0.3,0.5,0.6,0.8,1) & (0.2,0.26,0.4,0.6,0.9,1) \\ (0.1,0.26,0.4,0.6,0.8,0.9) & \\ (0.06,0.26,0.4,0.6,0.8,0.86) & (0.26,0.3,0.53,0.73,0.86,0.9) \\ (0,0,0.06,0.1,0.5,0.5) & \end{bmatrix}$$

$$\mu_{P_1}(D_1) = (0.6,0.5,0.3,0.1,0.6,1)$$

$$\mu_{P_1}(D_2) = (0.1,0.2,0.4,0.5,0.6,0.8)$$

$$\mu_{P_1}(D_3) = (0.06, 0.2, 0.4, 0.5, 0.7, 0.8)$$

$$\mu_{P_2}(D_1) = (0.26, 0.3, 0.5, 0.6, 0.8, 1)$$

$$\mu_{P_2}(D_2) = (0.2, 0.26, 0.4, 0.6, 0.9, 1)$$

$$\mu_{P_2}(D_3) = (0.1, 0.26, 0.4, 0.6, 0.8, 0.9)$$

$$\mu_{P_3}(D_1) = (0.06, 0.26, 0.4, 0.6, 0.8, 0.86)$$

$$\mu_{P_3}(D_2) = (0.26, 0.3, 0.53, 0.73, 0.86, 0.9)$$

$$\mu_{P_3}(D_3) = (0, 0, 0.06, 0.1, 0.5, 0.5)$$

Step 2:-

$$f\left(\frac{P_1}{D_1}\right) = \frac{\mu_{D_1}(P_1) (-) \mu_{P_1}(D_1)}{\max\{\mu_{D_1}(P_1), \mu_{P_1}(D_1)\}}$$

$$= \frac{(0.6, 0.5, 0.3, 0.1, 0.6, 1) (-) (0.6, 0.5, 0.3, 0.1, 0.6, 1)}{\max\{(0.6, 0.5, 0.3, 0.1, 0.6, 1), (0.6, 0.5, 0.3, 0.1, 0.6, 1)\}}$$

$$AM\left(f\left(\frac{P_1}{D_1}\right)\right) = 0$$

$$AM\left(f\left(\frac{P_1}{D_2}\right)\right) = 0.244$$

$$AM\left(f\left(\frac{P_1}{D_3}\right)\right) = 0.107$$

$$AM\left(f\left(\frac{P_2}{D_1}\right)\right) = -0.244$$

$$AM\left(f\left(\frac{P_2}{D_2}\right)\right) = 0$$

$$AM\left(f\left(\frac{P_2}{D_3}\right)\right) = 0.141$$

$$AM\left(f\left(\frac{P_3}{D_1}\right)\right) = -0.107$$

$$AM\left(f\left(\frac{P_3}{D_2}\right)\right) = -0.141$$

$$AM\left(f\left(\frac{P_3}{D_3}\right)\right) = 0$$

Step 3:-

The comparison matrix = AM $\left(f\left(\frac{P_i}{D_j}\right)\right)$ is given by

$$A = \begin{bmatrix} 0 & 0.244 & 0.107 \\ -0.244 & 0 & 0.141 \\ -0.107 & -0.141 & 0 \end{bmatrix}$$

Step 4:-

In Comparison matrix, Maximum of Ist row = 0.244

Maximum of IInd row = 0.141

Maximum of IIIrd row = 0

Here P₁ is affected by jaundice, P₂ is affected by dengue and P₃ is affected by jaundice

6. Conclusion

Fuzzy set framework has been utilized in several different approaches to model the medical diagnostic process and decision

making process. From the above analysis it is obvious that, the patient P₁ is affected by jaundice, P₂ is affected by dengue and P₃ is affected by jaundice. In this paper, we have applied the notion of Hexagonal fuzzy membership matrix in a medical diagnostic model. The advantage of this model is, if the patient-disease matrices are known, then it is possible to find which patient is suffering from what kind of disease.

References

- [1] Bellman.R. and Zadeh.L.A. – “Decision making in a fuzzy environment.” Management Science.
- [2] S.Elizabeth and L. Sujatha – Application of Fuzzy Membership Matrix in Medical Diagnosis and Decision Making
- [3] Kim, K.H and Roush, F.W – “Generalized fuzzy matrices.” Fuzzy Sets Sys.
- [4] M.Pal, Fuzzy matrices with fuzzy rows and fuzzy columns, Journal of intelligent and fuzzy systems, 30(1) (2015) 561-573.
- [5] M.Pal, Interval-valued fuzzy matrices with interval-valued fuzzy rows and columns, Fuzzy Information and Engineering, 7(3) (2015) 335-368.
- [6] Meenakshi A.R.(2008), “ Fuzzy Matrix” Theory and Application, MJP publishers
- [7] A.R.Meenakshi and M. Kaliraja, An application of interval valued fuzzy matrices in medical diagnosis, International journal of Analysis, 5(36) (2011) 1791 – 1802.
- [8] A.K.Shyamal and M.Pal, Two new operators on fuzzy matrices, J. Applied Mathematics and computing, 15 (1-2) (2004) 91-107.
- [9] A.K.Shyamal and M.Pal, Triangular fuzzy matrices, Iranian Journal of Fuzzy systems, 4(1) (2007) 75-87.
- [10] Sanchez, E., “Inverse of fuzzy relations, application to possibility distribution and medical diagnosis”, Fuzzy Sets and Systems, Vol. 2, pp. 75-86, 1979.
- [11] Vasantha. W.B.Kandasamy “Elementary Fuzzy Matrix Theory and fuzzy models for social scientist.”
- [12] Zadeh, L. A., Fuzzy sets, Information and Control, Vol. 8, pp. 338-353, 1965