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Application of intuitionistic fuzzy sets in high school determination via normalized Euclidean distance method

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Abstract: Intuitionistic fuzzy set is very profitable model to elaborate uncertainty and vagueness involved in decision making. In this paper, we have proposed an application of intuitionistic fuzzy set in school determination using normalized Euclidean distance method. Normalized Euclidean distance method has been utilized in order to measure the distance between each student and each school. The schools, in which each of the students have enrolled, have been determined via normalized Euclidean distance method depending upon examination that is performed for transition to high school education. Solution has been determined by measuring the smallest distance between each student and each school.

Keywords: Intuitionistic fuzzy sets, High school determination, Distance measures. **AMS Classification:** 03E72.

1 Introduction

The notion of fuzzy logic was firstly defined by L. A. Zadeh in 1965 [1]. The membership function of an element to a fuzzy set is a value between zero and one, the non-membership function of an element to a fuzzy set is equal to 1 minus the membership degree in fuzzy set theory. Then, Intuitionistic fuzzy sets (shortly IFS) were defined by K. Atanassov in 1986 [2]. Intuitionistic fuzzy sets form a generalization of the notion of fuzzy sets. In intuitionistic fuzzy set theory, sum of the membership function and the non-membership function is a value between zero and one. The hesitation degree is defined as 1 minus the sum of membership and non-membership degrees respectively in intuitionistic fuzzy sets. In fuzzy set theory, the hesitation degree is zero since sum of the membership function and the non-membership function is 1. The intuitionistic fuzzy set theory is useful in various application areas, such as algebraic structures, robotics, control systems, agriculture areas, computer, irrigation, economy and various engineering fields. The knowledge and semantic representation of intuitionistic fuzzy set become more meaningful, resourceful and applicable since it includes the degree of membership, the degree of non-membership and the hesitation margin [3]. Szmidt and Kacprzyk have showed that intuitionistic fuzzy sets are pretty useful in situations when description of a problem by a linguistic variable given in terms of a membership function only seems too rough [6]. Due to the flexibility of intuitionistic fuzzy set in handling uncertainty, it is a tool for a more human consistent reasoning under imperfectly defined facts and imprecise knowledge [7].

Various applications of intuitionistic fuzzy set have been carried out through distance measures approach. Many researchers have explored various applications of intuitionistic fuzzy set such as medical diagnosis, medical application, career determination, real life situations [6–12]. In this paper, we proposed an application of intuitionistic fuzzy set in school determination via normalized Euclidean distance method. Normalized Euclidean distance method was utilized in order to measure the distance between each student and each school. The schools in which each of the students has enrolled were determined via normalized Euclidean distance method depending upon examination that is performed for transition to high school education. Solution is determined by measuring the smallest distance between each student and each school.

For this paper; high schools in Kahramanmaraş city in Turkey have been researched. There is an exam in Turkey performed for transition to high school education. Each of the students has been enrolled in each high school taking their examination scores into account. Students' school achievement scores do not have a huge impact on their examination scores. Therefore; their school achievement scores have been accepted as fixed. Besides; success score of the schools has been accepted fixed. While searching database which is used in this paper; socio-economic status of students, student psychology, success of schools, teacher factor, order of preference, different city preference of students are ignored. As factors such as success of student, examination difficulty vary every year, these data will show variability for every year. In this paper; each high school base point has been calculated depending on student examination score (over 100 marks total). This research has utilized official data that were obtained from the Ministry of Education. Approximately 7000 students have been searched for 2016–2017 academic year. Afterwards, some students who were randomly selected have been searched depending on scores of the exam. We have used intuitionistic fuzzy sets as a tool since it incorporates the membership degree (the marks of the questions that have been correctly answered by the student, the nonmembership degree (the marks of the questions that have been wrongly answered by the student) and the hesitation degree (the marks of the questions that are free from any answer). Solution has been assumed by measuring the smallest distance between each student and each school via normalized Euclidean distance method. In solution; high school has been intuitionally assumed.

2 Preliminaries

Definition 1. [1] Let X be a nonempty set. A fuzzy set A drawn from X is defined as

$$A = \{ \langle x, \mu_A(x) \rangle | x \in X \},\$$

where $\mu_A(x): X \to [0,1]$ is the membership function of the fuzzy set A.

Definition 2. [2, 3] Let X be a nonempty set. An intuitionistic fuzzy set A in X is an object having the form

$$A = \{ \langle x, \mu_A(x), \nu_A(x) \rangle | x \in X \},\$$

where the functions

$$\mu_A(x), \nu_A(x) : X \to [0, 1]$$

define respectively, the degree of membership and degree of nonmembership of the element $x \in X$, to the set A, which is a subset of X, and for every element $x \in X$,

$$0 \le \mu_A(x) + \nu_A(x) \le 1.$$

According to Fuzzy Set Theory, if the membership degree of an element x is $\mu(x)$, if the nonmembership degree of an element x is $1 - \mu(x)$

Furthermore, we have

$$\pi_A(x) = 1 - \mu_A(x) - \nu_A(x)$$

called the intuitionistic fuzzy set index or hesitation on margin of x in A. $\pi_A(x)$ is degree of indeterminacy of $x \in X$ to the IFS A and $\pi_A(x) \in [0, 1]$ i.e.,

$$\pi_A: X \to [0,1]$$

for every $x \in X$. $\pi_A(x)$ expresses the lack of knowledge of whether x belongs to IFS A or not.

Definition 3. Let X be nonempty. Intuitionistic fuzzy sets $A, B, C \in X$. The distance measure d between intuitionistic fuzzy sets A and B is a mapping $d : X \times X \rightarrow [0, 1]$; if d(A, B) satisfies the following axioms:

$$\begin{array}{l} A1) \ 0 \leq d(A,B) \leq 1 \\ A2) \ d(A,B) \ \textit{if and only if } A = B \\ A3) \ d(A,B) = d(B,A) \\ A4) \ d(A,C) + d(B,C) \geq d(A,B) \\ A5) \ \textit{if } A \subseteq B \subseteq C, \ \textit{then } d(A,C) \geq d(A,B) \ \textit{and } d(A,C) \geq d(B,C) \end{array}$$

Distance measure is a term that describes the difference between intuitonistic fuzzy sets and can be considered as a dual concept of similarity measure. Distance measures between intuition-istic fuzzy sets are proposed [4,5].

Definition 4. Let

$$A = \{ \langle x, \mu_A(x), \nu_A(x), \pi_A(x) \rangle | x \in X \}$$

and

$$B = \{ \langle x, \mu_B(x), \nu_B(x), \pi_B(x) \rangle | x \in X \}$$

be two intuitionistic fuzzy sets in $X = x_1, x_2, ..., x_n, i = 1, 2, ..., n$. Based on the geometric interpretation of intuitionistic fuzzy set, Szmidt and Kacprzyk [4, 5] proposed the following four distance measures between A and B:

The Hamming distance;

$$d_H(A,B) = \frac{1}{2} \sum_{i=1}^n (|\mu_A(x_i) - \mu_B(x_i)| + |\nu_A(x_i) - \nu_B(x_i)| + |\pi_A(x_i) - \pi_B(x_i)|)$$

The Euclidean distance;

$$d_E(A,B) = \sqrt{\frac{1}{2} \sum_{i=1}^{n} \left[(\mu_A(x_i) - \mu_B(x_i))^2 + (\nu_A(x_i) - \nu_B(x_i))^2 + (\pi_A(x_i) - \pi_B(x_i))^2 \right]}$$

The Normalized Hamming distance;

$$d_{n-H}(A,B) = \frac{1}{2n} \sum_{i=1}^{n} (|\mu_A(x_i) - \mu_B(x_i)| + |\nu_A(x_i) - \nu_B(x_i)| + |\pi_A(x_i) - \pi_B(x_i)|)$$

The Normalized Euclidean distance;

$$d_{n-E}(A,B) = \sqrt{\frac{1}{2n} \sum_{i=1}^{n} \left[(\mu_A(x_i) - \mu_B(x_i))^2 + (\nu_A(x_i) - \nu_B(x_i))^2 + (\pi_A(x_i) - \pi_B(x_i))^2 \right]}$$

3 Application of intuitionistic fuzzy set via normalized Euclidean distance method

Let $H = \{H_1, H_2, H_3, H_4, H_5\}$ be a set of high schools, $L = \{Turkish, Mathematics, Science, Social, English, Religion\}$ be a set of lessons, $S = \{S_1, S_2, S_3, S_4, S_5\}$ be a set of students.

	Turkish	Mathematics	Science	Social	English	Religion
H_1	(0.965, 0.028, 0.007)	(0.985, 0.012, 0.003)	(0.995, 0.004, 0.001)	(0.990, 0.008, 0.002)	(0.975, 0.020, 0.005)	(0.995, 0.004, 0.001)
H_2	(0.910, 0.080, 0.010)	(0.945, 0.044, 0.011)	(0.945, 0.044, 0.011)	(0.920, 0.064, 0.016)	(0.900, 0.080, 0.020)	(0.990, 0.008, 0.02)
H_3	(0.845, 0.124, 0.031)	(0.800, 0.180, 0.020)	(0.825, 0.140, 0.035)	(0.850, 0.120, 0.030)	(0.855, 0.126, 0.019)	(0.950, 0.040, 0.010)
H_4	(0.710, 0.261, 0.029)	(0.570, 0.387, 0.043)	(0.750, 0.200, 0.050)	(0.825, 0.140, 0.035)	(0.650, 0.280, 0.070)	(0.690, 0.248, 0.062)
H_5	(0.640, 0.324, 0.036)	(0.400, 0.480, 0.120)	(0.550, 0.405, 0.045)	(0.610, 0.351, 0.039)	(0.570, 0.345, 0.085)	(0.830, 0.153, 0.017)

Table 1.

Table 2 below has been determined depending on students' examination score.

	Turkish	Mathematics	Science	Social	English	Religion
S_1	(0.95, 0.04, 0.01)	(0.98, 0.01, 0.01)	(0.99, 0.005, 0.005)	(0.95, 0.03, 0.02)	(0.98, 0.01, 0.01)	(0.95, 0.04, 0.01)
S_2	(0.90, 0.08, 0.02)	(0.98, 0.01, 0.01)	(0.95, 0.03, 0.02)	(0.99, 0.008, 0.002)	(0.98, 0.015, 0.005)	(0.99, 0.006, 0.004)
S_3	(0.65, 0.28, 0.07)	(0.45, 0.44, 0.11)	(0.60, 0.32, 0.08)	(0.40, 0.48, 0.12)	(0.35, 0.52, 0.13)	(0.65, 0.30, 0.05)
S_4	(0.60, 0.32, 0.08)	(0.45, 0.44, 0.11)	(0.70, 0.24, 0.06)	(0.80, 0.16, 0.04)	(0.80, 0.17, 0.03)	(0.80, 0.16, 0.04)
S_5	(0.70, 0.24, 0.06)	(0.80, 0.16, 0.04)	(0.85, 0.12, 0.03)	(0.75, 0.20, 0.05)	(0.90, 0.08, 0.02)	(0.95, 0.04, 0.01)
S_6	(0.85, 0.12, 0.03)	(0.95, 0.04, 0.01)	(0.95, 0.02, 0.03)	(0.95, 0.05, 0.00)	(0.85, 0.11, 0.04)	(0.95, 0.02, 0.03)
S_7	(0.90, 0.08, 0.02)	(0.95, 0.04, 0.01)	(0.95, 0.02, 0.03)	(0.95, 0.015, 0.035)	(0.90, 0.09, 0.01)	(0.95, 0.04, 0.01)

Table 2.

Table 3 below has been calculated shortest distance between each student (i.e., Table 2) and each high school (i.e., Table 1) using normalized Euclidean distance method.

	H_1	H_2	H_3	H_4	H_5
S_1	0.0231	0.0450	0.09420	0.26458	0.36903
S_2	0.0308	0.04279	0.10847	0.26316	0.35264
S_3	0.44333	0.48796	0.32449	0.20901	0.13595
S_4	0.29438	0.37369	0.17615	0.17789	0.14159
S_5	0.22332	0.12040	0.06946	0.17097	0.24231
S_6	0.27333	0.03055	0.08150	0.22134	0.33569
S_7	0.05008	0.03291	0.09209	0.23234	0.33903

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Table 3 depicts that the shortest distance between each student and each high school has given that the student will enroll in the high school. According to Table 3, the student S_1 is to enroll in H_1 high school, the student S_2 is to enroll in H_1 high school, the student S_3 is to enroll in H_5 high school, the student S_4 is to enroll in H_5 high school, the student S_5 is to enroll in H_3 high school, the student S_6 is to enroll in H_2 high school, the student S_7 is to enroll in H_2 high school.

4 Conclusion

In this paper, seven students who were randomly selected have been researched in order to form the above table. Moreover, this research can be applicable to larger groups of students. Socioeconomic status and psychology of the students as well as success of schools, teacher role, order of preference and different city preference are ignored, because these factors will have positive or negative effects on high school determination. As factors such as success of student, examination difficulty vary every year, these data will show variability for every year. This application of intuitionistic fuzzy set in high school determination is very useful; because by calculating distance between each student and each school, the most proper school for each student has been intuitionally determined.

Available system results with the method which we have used are compatible. This paper has shown that used method could be applied to evaluation system through various arrangements.

This method is suitable in order to achieve more sensible results. Available evaluation system could be renewed by using intuitionistic fuzzy logic.

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