EARTHQUAKE VULNERABILITY ASSESSMENT FOR HOSPITAL BUILDINGS USING A GIS-BASED GROUP MULTI CRITERIA DECISION MAKING APPROACH: A CASE STUDY OF TEHRAN, IRAN

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ABSTRACT:

Nowadays, urban areas are threatened by a number of natural hazards such as flood, landslide and earthquake. They can cause huge damages to buildings and human beings which necessitates disaster mitigation and preparation. One of the most important steps in disaster management is to understand all impacts and effects of disaster on urban facilities. Given that hospitals take care of vulnerable people reaction of hospital buildings against earthquake is vital. In this research, the vulnerability of hospital buildings against earthquake is related to a number of criteria including age of building, number of floors, the quality of materials and intensity of the earthquake. Therefore, the problem of seismic vulnerability assessment is a multicriteria assessment problem and multi criteria decision making methods can be used to address the problem. In this paper a group multi criteria decision making model is applied because using only one expert's judgments can cause biased vulnerability degree of buildings. Fuzzy capacities which are similar to layer weights in weighted linear averaging operator are calculated using particle swarm optimization. Then, calculated fuzzy capacities are included into the model to compute a vulnerability degree for each hospital.

1. INTRODUCTION

Tehran is situated in a seismically active region in Middle East and it is dramatically threatened by earthquake (M Moradi, Delavar, Moshiri, & Khamespanah, 2014). In recent years, Tehran became a centre for many financial, educational and recreational activities. Rapid urban growth lack of urban planning and urban sprawl render the city extremely vulnerable to earthquake. Among all facilities in Tehran metropolitan area, hospitals have a remarkable value because they are related to a vulnerable category of people. A number of researches have been undertaken in order to assess the vulnerability of habitats and to estimate the number of deaths and casualties in the future earthquake (Mansouri, Kiani, & Amini-Hosseini, 2014; SHAYANNEJAD & ABEDINI ANGERABI, 2014; Sinha, Priyanka, & Joshi, 2014). Considering the fact that many factors are affecting seismic vulnerability of hospital buildings, this problem is a multi-criteria problem (Rashed & Weeks, 2003). This paper applies a multi-criteria method in order to assess the vulnerability of hospital buildings against earthquake. In addition, given that multi-criteria algorithms are ill-structured (Chakraborty & Joshi, 2014) which means that different multicriteria models may produce different results, in this paper a model based on Sugeno integral is applied to analyse the vulnerability of hospital buildings. Moreover, more than one expert's judgments are included into the model to reduce the uncertainty of single expert systems. Firstly, a vulnerability degree is calculated based on every expert's judgments. Following that, a group multi criteria decision making model is

1.1 Related works

Seismic vulnerability assessment is a core process in disaster management and frequent efforts have been undertaken to address this problem. A number of these researches were related to physical vulnerability which indicates the vulnerability of buildings and structures against earthquake (Milad Moradi, Delavar, & Moshiri, 2015). Milad Moradi et al. (2015) proposed a GIS-based model using ordered weighted averaging operator that is able to analyse different seismic vulnerability maps based on optimistic and pessimistic scenarios. A group multi criteria decision making model was proposed by Peng (2012) for seismic vulnerability assessment. They assessed seismic vulnerability using a number of different MCDM methods in order to compare the results. A number of

used to aggregate different experts' vulnerability degrees and produce one overall answer. Sugeno integral which is able to aggregate attribute values with respect to the interaction among them is employed in this research. In fact, Sugeno integral is employed to serve two purposes in the current research. First of all, to aggregate attribute values and produce a vulnerability map based on each expert's point of view and for group decision making. In this paper 5 experts are asked to evaluate the vulnerability of hospital buildings. Then, PSO is used to extract the optimum value of fuzzy capacities which are the unknowns of the Sugeno model. Finally, these fuzzy capacities which are obtained through training process are used for computing a vulnerability degree for each hospital building.

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researches addressed the problem of social vulnerability which indicates the degree to which human of a specific urban area are vulnerable against earthquake. Ge et al. (2013) analysed social vulnerability of urban areas against natural disasters with a multi criteria approach. Rashed and Weeks (2003) proposed an integrated multi criteria decision making model which is able to evaluate both social and physical vulnerabilities. They employed a fuzzy model to manage fuzzy uncertainties associated with input data. Earthquake vulnerability of urban facilities is assessed in some other researches. Considering the fact that children are a vulnerable part of any society, Panahi, Rezaie, and Meshkani (2014) designed and developed a seismic vulnerability assessment model for schools based on fuzzy analytical hierarchy process. They evaluated the impact of peak ground acceleration on schools buildings in Tehran. Rezaie and Panahi (2015) developed a model for including geotechnical parameters in a seismic vulnerability model based on multi criteria analysis. Although previous researches have investigated a wide range of aspects of this problem, vulnerability of hospital buildings have not been assessed before.

2. METHODOLOGY

2.1 Multi criteria decision making

Multi criteria decision making is a powerful tool that provides a variety of analyses in order to assess alternatives according to some aspects (Malczewski & Liu, 2014; Rinner, Keßler, & Andrulis, 2008). MCDM enables the decision makers to store, analyse and visualize the data, while GIS facilitates an environment to analyse the desirability of alternatives. In order to define the desirability relationships, expert's judgment is obligatory. Given that usually there is uncertainty among expert's judgments, in a number of decision making problems more than one expert is employed which is called group decision making (Jankowski & Nyerges, 2001; Ouma, Kipkorir, & Tateishi, 2011). Decision making is a framework in which useful information regarding with selecting an alternative from a set of alternatives is driven from row data layers. In spatial decision making, alternatives are points, lines and polygons which have a number of attribute values (Meng & Malczewski, 2015).

2.2 Sugeno integral

Sugeno integral is an aggregation operator which is able to compute an overall fitness measure regarding the interaction among criteria. The Sugeno integral applies fuzzy measures instead of simple weights. Fuzzy measure is a set function that assigns a value to each member of the power set of criteria. in fact, Sugeno integral need a measure indicating the importance of each set of criteria instead of the importance of each criteria which is indicated by traditional layer weights. Therefore, fuzzy capacities can produce a lattice for any problem which shows the relative importance of every criterion and every set of criteria. μ is a fuzzy measure on X the power set of criteria (Murofushi & Sugeno, 1989):

$$\mu: X \to [0,1] \tag{1}$$

where

 $\mu =$ fuzzy capacity X = the power set of criteria Considering the fact that importance of a larger set is more than the importance of any of its subsets, following equation should always be satisfied (Grabisch, 1995):

$$\mu(\varphi) = 0, \quad \mu(X) = 1$$

$$A \subseteq B \Rightarrow \mu(A) < \mu(B)$$
(2)

where

$$\mu = 1022y$$
 capacity
A, B = two subset of criteria

In this notation sugeno integral is as follows (Grabisch, 1995):

$$S_{\mu}(\mathbf{f}(x_{1}), \mathbf{f}(x_{2}), \dots, \mathbf{f}(x_{n})) = \prod_{i=1}^{n} (f(x_{i}) \wedge \mu(\mathbf{A}_{(i)}))$$
(3)

S = Sugeno integral of input vector X

where

$$\mu = \text{fuzzy capacity} \\ A_{(i)} = \{x_{(i)}, ..., x_{(n)}\}$$

The unknowns of this model are fuzzy capacities or fuzzy measures. In order to calculate them, experts' opinion should be employed (Grabisch & Labreuche, 2010). These fuzzy capacities indirectly implies the importance of each criterion in the decision making process. But they cannot provide explicit explanation because criterion i has an element of the fuzzy capacity lattice and the coalition of i and j also has another element, accordingly, it is impossible to understand the role of each criterion from the lattice itself. By defining other measures such as interaction index and Shapley index the role of each criterion in decision making can be expressed.

2.2.1 Interaction Index

In real world situation, the criteria are rarely independent and there is interaction among criteria which can be positive or negative. When there is positive interaction among the criteria, the importance of the coalition is greater than the importance of each criterion separately (Smrek, 2015). In terms of negative interaction, the fuzzy capacity associated to each criterion can be larger than the fuzzy capacity of the coalition. The interaction among the criteria can be computed using the following equation (Grabisch, 1996) :

$$I_{ij} = \sum_{K \subset X \setminus \{i,j\}} \frac{(n - |K| - 2)! |K|!}{(n - 1)!}$$
(4)
$$[\mu(K \cup \{i,j\}) - \mu(K \cup \{i\}) - \mu(K \cup \{j\}) + \mu(K))]$$

where $\mu =$ fuzzy measure of the set of criteria n = is the number of criteria I = computed interaction of criterion *i* and *j*.

Positive values of interaction indicate complementary criteria, whereas negative values indicate redundant criteria. It is also possible to calculate the interaction among any given set of criteria using the following equation (Grabisch, Sugeno, & Murofushi, 2000):

$$I(A) = \sum_{B \subset X \setminus A} \frac{(n - |B| - |A|)! |B|!}{(n - |A| + 1)!} \sum_{C \subset A} (-1)^{|A \setminus C|} \mu(C \cup B)$$
(5)

On the other hand, it is possible to calculate the fuzzy measures from a set of training data. In this method, optimum fuzzy capacities can be calculated through an optimization algorithm. Moreover, given that Equation 2 should always be satisfied, the problem of finding fuzzy capacities is a constrained optimization problem. Successful models based on PSO and GA have been proposed for finding fuzzy capacities (Wang, He, Dong, & Zhao, 2011).

2.3 Particle Swarm Optimization

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Particle Swarm Optimization is a swarm optimization algorithm. The first step is to initialize a set of particles. Then, the location of these particles should be updated using the following equation (Wang et al., 2011):

$$X^{t+1}(i) = X^{t}(i) + v^{t+1}(i)$$
(6)

where X = the position of the i th particle in t th iteration v = the velocity of i th particle.

The velocity of each particle tends to move the particle towards the best position in the search space. The velocity is updated using the following equation (Wang et al., 2011):

$$v^{t+1}(i) = w^{*}v^{t}(i) + c_{1}^{*}$$
best-X^t(i)) + c_{2}^{*}(Gbest-X^t(i))
(7)

where Pbest = the location of the best particle in the ith iteration

Gbest = the best location ever searched.

Algorithm: PSO				
1- Start				
2- i=0				
3- initialize 100 particles p(i)				
4- Assess particles				
5- While (the accuracy is not reached)				
6- Start				
7- i=i+1				
8- update the weight vector				
9- select Pbest of particles				
10- update Gbest				
11- update velocity				
12- update particle position				
13- assess particles				
14- terminate				
15- End				

Figure 1. The Pseudo code of PSO

Figure 1 indicates the steps which have been done to extract fuzzy capacities from sample data. In this paper, each element in the lattice of fuzzy capacities is simulated as a particle. Following that, by doing the steps shown in Figure 1, particles move toward the position that has minimum inconsistency with the objective function and the best set of fuzzy measures can be found.

3. IMPLEMENTATION

In this research, a multi criteria evaluation model is developed in Matlab which is an easy programing language. In this model, firstly, five experts are asked to give a score to 20 randomly selected hospitals and health care centres for their vulnerability against earthquake. Then, particle swarm optimization is employed as mentioned in section 2.3 to extract fuzzy capacities from experts' data. Following that, Equation (3) is applied to calculate vulnerability degree of other hospital buildings. Finally, these vulnerability degrees are used to produce seismic vulnerability for hospital buildings.

3.1 Study area

The study area of this research is Tehran metropolitan area which is located in the central part of Iran. Its area is about 600 $\rm km^2$ and is located between 51° 15' and 51° 35'E and 35° 33' and 35° 50'N. Tehran is situated on the southern part of Alborz Mountains which are surrounded by a number of known and unknown faults. Therefore, Tehran is in a seismically active region and its habitants are threatened by earthquake risk. Moreover, the recurrence of earthquake is about 100 years and no large earthquake has stroked the city for more than 100 years. Consequently, an earthquake may be imminent in Tehran based on evidence. A number of hospitals are constructed on or near the faults which can cause a huge damage in case of any possible seismic threat. All the aforementioned situations necessitate researches on seismic hazard mitigation and disaster management.

3.2 Contributing criteria

Due to the fact that seismic damage is associated with a number of different criteria such as age and number of stories of the buildings, seismic vulnerability assessment can be considered as multi criteria problem. Accordingly, this problem should be addressed using multi criteria evaluating methods. Different criteria are included in previous researches on seismic vulnerability assessment including peak ground acceleration, distance to hazardous facilities, soil liquefaction and seismic resonance but because of the fact that authors have limited access to up-to-date data these factors are not included into the model. Six criteria are included in this research including the number of floors of the hospitals, the quality of hospital building's materials, age of the buildings, the capacity of the hospitals, the distance of each hospital to the nearest fault and he average slope of earth's surface in the neighbourhood of the hospitals which are discussed in Table 1.

3.3 Model implementation

The steps shown in Figure are done in this research in order to find the vulnerability degree of hospitals. The resulted vulnerability degrees are then classified into 5 categories and illustrated in ArcGIS software. These steps are for a single expert and in this paper these steps are done for all experts separately.

ID	Criteria	Symbol	Reference
1	number of floors	Flr	(Debnath, 2013; Milad
			Moradi et al., 2015)
2	Building quality	Bql	(Milad Moradi et al.,
			2015; Rashed & Weeks,
			2003)
3	Age of building	Abl	(Huang, Zhou, & Wang,
			2014; Rashed & Weeks,
			2003)
4	Capacity	Cap	(Rashed & Weeks, 2003)
5	Distance to	Dfl	(M Moradi, Delavar, &
	faults		Moshiri, 2013; Nath,
			Adhikari, Devaraj, &
			Maiti, 2014;
			SHAYANNEJAD &
			ABEDINI ANGERABI,
			2014)
6	Slope	Slp	(Barbat et al., 2010; Nath,
			Adhikari, Devaraj, &
			Maiti, 2015)





Figure 2. The steps of the proposed model

4. RESULTS AND DISCUSSION

Figure 3 depicts the spatial distribution of the vulnerability of hospitals in Tehran based on the opinion of all experts. In fact, this map is based on the consensus of the five experts' opinions. Overall, 12%, 29%, 34% and 25% of hospital buildings are categorized in least vulnerable, less vulnerable, vulnerable and very vulnerable classes, respectively.



Figure 3. Vulnerability map of hospital buildings

This data can provide us an insight into the vulnerability conditions of the hospitals buildings. There is considerable number of vulnerable and very vulnerable hospitals in Tehran, according to which the modification and reconstruction of these buildings are obligatory. Results indicate that most of the vulnerable hospitals are located in the central parts of the city where most of the buildings are aged and roads are narrow. Thus, the results necessitate maintenance operations for healthcare centres inside Tehran metropolitan area.

5. CONCOLUSION

A new group MCDM model is proposed in this paper. The proposed model is able to aggregate attribute values with respect to the interaction among criteria. Sugeno integral is able to take the interaction among the criteria into account. In this paper, five experts are asked to specify the degree to which hospitals are vulnerable. Using more than one expert's opinion, the model reduces the uncertainty associated with experts' judgments. The results depict that most hospitals in central and southern part of Tehran are extremely vulnerable and may cause huge damages in case of earthquake. This research also specifies the hospitals that need immediate attention. The multi criteria decision making model proposed in this research can be employed in other disaster management studies with minor modifications.

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