



1	New Approaches to <mark>Seismic Microzonation Modelling of Ground Shaking</mark> Using Direct
2	Characteristics of Influencing Criteria: Case Study of Bam City, Iran
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13	Abstract
14	This paper proposes a new model in evaluating seismic microzonation of ground shaking by considering
15	direct characteristics of influencing criteria and dealing with uncertainty of modelling through production
16	of fuzzy membership functions for each criterion. The relevant criteria were explored by reviewing
17	previous literature and interviewing 10 specialized experts. Analytic Hierarchy Process (AHP) and Fuzzy
18	Logic (FL) methods were applied in order to define priority rank of each criteria and to fuzzify sub-criteria
19	of each criterion by interviewing 10 experts, respectively. Applying Fuzzy Logic method to deal with
20	uncertainties of sub criteria of each criterion and using direct characteristics of each criterion are the new
21	approaches in designing a new model. The criteria and sub-criteria were combined in GIS to develop a
22	model for assessing microzonation of ground shaking in the study area of Bam city, Iran. The model's
23	output shows high to very high ground shaking levels were happened in central, east, and northeast to
2 4	north part of the area, The validation results based on overall accuracy and Kappa statistics showed 80%
25	to 82% accuracy, 0.74 and 0.75 Kappa indicating a good fit to the model's output. This model assists
26	planners and decision makers to produce seismic microzonation of ground shaking to be incorporated in
27	designing new development plans of urban and rural areas, and to facilitate making informed decision
28	regarding safety measures of existing buildings and infrastructures.
29	Keywords: Seismic Microzonation, Site Effects, Ground Shaking, Spatial Modelling, Analytic Hierarchy
30	Process, Fuzzy Logic and GIS.
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34 1. Introduction

This paper explores direct characteristics of influencing criteriarand dealing with uncertainty of 35 modelling through production of fuzzy membership functions for each criterion for the assessment 36 of ground shaking amplification in a study area. MERM microzonation manual (2003) sets 37 different factors effecting on the amplitude and duration of ground shaking at a specific site. 38 These include "the magnitude of the earthquake, focal point and depth of the earthquake, 39 directivity of the energy release, distance of rapture from the site, geological condition from the 40 site to the location of the earthquake, and local geology and topographical condition of the site" 41 (SM Working Group, 2015;Boore, 2003;Hassanzadeh et al., 2013). It has long been known that 42 local conditions of foundation soils have a significant impact on the effects of an earthquake, as 43 it was demonstrated in previous earthquakes such as Mexico City, 1985 (Beck and Hall, 1986), 44 Kobe, 1995 (Wald, 1996), Izmit, 1999 (Tang, 2000) and Umbria-Marche earthquake, 1997 (Moro 45 et al., 2007). It was witnessed in the Bam earthquake, 2003, that buildings located on 46 47 unconsolidated sediments had greater destruction levels (Ramazi and Jigheh, 2006). The aim of seismic microzonation studies is to prepare ground-shaking map that can communicate efficient 48 data to planners and policy makers in a geographic area for making informed decision regarding 49 development policies in urban areas. Therefore, this community require accurate and certain 50 51 information for developing mitigation plans and strategies. In the spite of this, there are 52 uncertainties in estimating seismic microzonation of ground shaking at a site, as this can be influenced by complex factors such as the estimates of earthquake source, wave propagation, and 53 54 site condition. Uncertainty in these criteria results in uncertain ground-motion estimate from earthquakes (Wang et al., 2017; Wang et al., 2016; Petersen et al., 2016). 55 Probabilistic Seismic Hazard Analysis (PSHA) (Cornell, 1968) has been used to assess ground-56 57 motion hazards from earthquakes (Atkinson et al., 2015;Petersen et al., 2016). This method 58 dependent on "the length of the causative faults and depth of the earthquake", which are generally unknown that cause uncertainty in assessing ground-motion of earthquakes (Wang et al., 2017). 59 60 In deterministic seismic hazard analysis (DSHA) (Campbell, 2003;Atkinson and Boore, 2006)

- absent of relevant ground-motion attenuation relationship for specific geographic areas can cause 61
- 62 uncertainty in applying DSHA for assessing ground motions of an earthquake (Wang et al.,
- 63 2017). Scenario-based seismic hazard analysis (SSHA) (Panza et al., 2012) applies groundmotion simulations of a scenario earthquake using specified source, path and site parameters. By 64





conducting many simulations, earthquake variability of different sources, ground-motion
propagation characteristics, and local site effects can be considered. Therefore, uncertainties
using SSHA are quantified explicitly (Wang et al., 2017).

Accurate measurement and communication of uncertainties are critical in ground-motion hazard 68 assessment for earthquakes. Thus, other approach in microzonation studies is the use of multi-69 criteria decision-making methods (MCDM). According to these methods after identifying 70 71 potential criteria, experts evaluate and choose among qualitative and quantitative criteria. Since 72 experts' judgments can be subjective and imprecise; uncertainty also exists in the analysis. 73 Uncertainty stems mainly from sources such as the lack of the incomplete data availability, 74 vagueness, and linguistic expert view. Such uncertainties and vagueness can be dealt with fuzzy 75 logic principles (Zadeh, 1965) and inference systems (Klir, 2004;Zadeh, 1975). Based on fuzzy 76 logic method, the content of each sentence implies logical rules, which constitute the foundation of fuzzy system modeling and inference procedures. In comprehensive decisions, an expert's 77 78 heuristic knowledge or empirical information is used frequently for better conclusions. For these 79 reasons, Fuzzy Logic is used for evaluating of seismic microzonation of ground shaking 80 amplification.

81 There are many MCDM tools in the literature but Analytical Hierarchal Process (AHP) (Saaty, 82 1980) is one of the most useful techniques, and plays an important role in calculating criteria's weights and selecting optimized alternatives. Sitharam and Anbazhagan (2008) applied AHP and 83 GIS for seismic microzonation studies in Bangalore, India. Furthermore, AHP and GIS was 84 85 applied to produce seismic microzonation map of Dehli (Mohanty et al., 2007), Haldia, Bengal Basin (India) (Mohanty and Walling, 2008), Erbaa (Turkey) (Akin et al., 2013) and Al-Madinah 86 (Moustafa et al., 2016). Fuzzy Logic method was used for evaluation of earthquake damage to 87 buildings (Sen, 2010), and quick seismic microzonation (Teramo et al., 2005;Nath and 88 89 Thingbaijam, 2009; Boostan et al., 2015). Although there were a number of publications evaluating the seismic microzonation of ground shaking amplification in the literature, but there 90 91 is lack of evidence in using the Fuzzy Logic method for producing seismic microzonation of ground shaking amplification. Moreover, few researchers have considered direct characteristics 92 93 of each criteria in local ground shaking analysis. Additionally, in order to remove uncertainties 94 regarding source of probable earthquake, magnitude and rapture length, therefore these criteria was not considered for producing seismic microzonation of ground shaking in this study. 95





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	The main purpose of this paper is to develop a model for evaluation of seismic microzonation of
97	ground shaking amplification using AHP, Fuzzy Logic and Weighted Linear Combination
98	(WLC) methods in GIS. At this stage, model inputs are direct characteristics of local geology,
99	hydrology, sedimentology, and topographical factors that should be taken into consideration.
100	First all selected criteria were weighted using AHP method by interviewing 10 experts, then all
101	criteria are converted into fuzzy sets and fuzzy membership functions (MFs) were produced, then
102	WLC and fuzzy inference rules are used to develop a model for producing seismic microzonation
<mark>103</mark>	of ground shaking amplification for a study area.
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106	2. Material and methods
107	This study investigates the importance of influencing factors on seismic microzonation of ground
<mark>108</mark>	shaking. These criteria are identified by reviewing previous literature. Analytic Hierarchy
109	Process (AHP) and Fuzzy Logic (FL) Methods are applied to deal with selection, weighting and
110	fuzziness of criteria due to associated uncertainties in the decision-making process of seismic
<mark>111</mark>	microzonation of ground shaking amplification by interviewing experts. Combining the criteria
112	and sub criteria is done based on WLC method. Finally, the developed model is validated using
113	Overall Accuracy (OA) and Kappa statistics methods. The study has been conducted in four steps
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 115 116 117 118 119 120 121 122 123 	Figure 1. The methodological approach of the model 2.1. Identification, Weighting and Fuzzification of Criteria Seismic microzonation of ground shaking can be influenced by several criteria. These criteria need to be identified by reviewing literature and interviewing experts in data gathering step.





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128	2.2.1.	Analytical	methods
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129 2.2.1. Analytic Hierarchy Process (AHP) method

Several methods have been developed to deal with ranking of criteria and solving a problem, 130 131 such as Regime (Hinlopen et al., 1983), ELECTRE family (Figueira et al., 2005), Analytical Hierarchy Process (AHP) (Saaty, 1980), and Multiple Attribute Utility approach (MAUT) 132 (Keeney and Raiffa, 1993). AHP is one of the most commonly used multi-criteria decision 133 making (MCDM) tools, and allows the consideration of both objective and subjective factors in 134 135 ranking alternatives in a hierarchical decision model (Saaty, 1980;Saaty, 1990). This method is applied to convert the experts' view on the importance of each criterion and sub-criterion to a 136 numerical value by comparing them to one another, one pair at a time (pair-wise comparison) 137 138 (Saaty, 1980).

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AHP matrix (A) is developed from the pair-wise comparison of the relative importance of criterion A_i to criterion A_j (α_{ij} , represents a quantified judgment on a pair of criteria C_i, C_j) (Figure 2), as it was explained above. The values assigned to α_{ij} according to the Saaty's scale (1980) are usually in the interval of 1 to 9 or their reciprocals. In order to calculate the priority ranking of each criterion (weight), Saaty (1990) suggested the mathematical computation of eigenvector (Eq. 1& 2).

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Figure 2.AHP matrix (A)

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$$\lambda_{max} = \sum_{j=1}^{n} a_{ij} \, \frac{W_j}{W_i} \tag{Eq. 1}$$

151 Where: λ_{max} = the largest eigenvalue; α_{ij} = judgment; $W_i \& W_j$ = numerical weights for judgment 152 α_{ij} .

153

$$(A - \lambda_{max} I)X = 0 \tag{Eq. 2}$$

154 Where: A= AHP matrix; λ_{max} = the largest eigenvalue; I= Unique matrix; X= eigenvector.





155 In addition, the assignment of weights (the degree of importance) to each criterion relates to the 156 process of the experts' logical and analytical thinking, which is tested for each matrix with 157 Consistency Ratio (CR) statistics. In case, this statistics is less than 0.1 (CR < 0.1) the experts' 158 answers are logical. Following the testing for consistency, the weights are aggregated to 159 determine ranking of decision alternatives (the weights) for each criteria. Therefore, in this 160 research, AHP method is applied to calculate the degree of importance of each criterion 161 influencing on seismic microzonation level of ground shaking in a region using interview data of_1 162 163 10 specialized experts in seismology, earthquake engineering, geology, tectonics and structural 164 engineering. 165 166 167 2.2.2. Fuzzy Logic (FL) method 168 Fuzzy logic is a method of "approximating modes of reasoning" (Novák et al., 2012), and it is a 169 mathematical tool that deals with uncertainty in a different way that can relate independent to 170 171 dependent variables. Zadeh (1965) introduced Fuzzy set theory Indicating that the boundary is not precise and the gradual change is expressed by a membership function, and it changes from 172 non-membership to membership in a fuzzy set (Eq. 3). The characteristic function can be 173 174 assigned a value between 0 to $\frac{1}{2}$. Each membership function is represented by a curve that 175 indicates the assignment of a membership degree in a fuzzy set to each value of a variable. Curves of the membership functions can be linear, triangles, trapezoids, bell-shaped, or have more 176 complicated shapes (Figure 3) depend on the purpose of the subject (Demicco and Klir, 2003). 177 178 179 $A_a = \{ x \in X \mid \mu_A(x) \ge a \}$ (Eq. 3)Where A_a is called the a-cut or a-level set of A, and $\mu_A(x)$ represents membership degree of the 180 181 element x. 182 183 Figure 3. Fuzzy membership functions (After Mancini, 2012) 184





- Fuzzy systems are mainly based on expert knowledge to formalize reasoning in natural language
 mostly using sets of fuzzy inference rules or *"if-then"* rules (Eq. 4).
- 188 If x is A then y is B (Eq. 4)
- 189

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190 As membership functions curve can easily be changed by small increments based on expert knowledge, therefore, fuzzy logic can characterize and model geologic systems in an efficient 191 way (Klir, 2004;Demicco and Klir, 2003). Therefore, in this research using Fuzzy set, the 192 uncertainties in producing microzonation map of ground shaking can be managed by defining 193 fuzzy membership functions for each criterion. This happens by assigning meaningful values (0 194 195 to 1) to each individual (sub criteria) of each criterion through interviewing 10 specialized experts. For the purpose of defuzzification, largest of maximum method was used that the precise 196 value of the variable output is one of which the fuzzy subset has the maximum truth-value 197 198 (Mancini et al., 2012).

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200 2.3. Data gathering

In order to identify influencing criteria in seismic microzonation of ground shaking the required data were collected through a literature review, and semi-structured interviews with 10 experts who were involved in the geology, seismology, tectonic and structural engineering, and geomorphology fields. They were asked about the criteria that can influence seismic microzo nation level of ground shaking, and then these data were analyses using AHP and FL methods as explained in the following:

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208 2.3.1. Determining the relevant criteria by reviewing literature

The potential criteria influencing seismic microzonation of ground shaking were determined through reviewing previous research. By reviewing documents on earthquake engineering, seismology, geology, tectonic and structural engineering, geomorphology and seismic microzonation reports and guidelines (Fäh et al., 1997;Ding et al., 2004;Molina et al., 2010;Mundepi et al., 2010;Marulanda et al., 2012;Hassanzadeh et al., 2013;Federal Emergency Management Agency (FEMA), 2014;Fraume et al., 2014;Grelle et al., 2016;Grelle et al., 2014;SM





- 215 Working Group, 2015;Rehman et al., 2016;Nwe and Tun, 2016;Global Earthquake Model (GEM),
- 216 2017;CAPRA, 2017;Michel et al., 2017;Trifunac, 2016;Hassanzadeh and Nedovic-Budic, 2016),
- 217 in total 14 criteria were recognized that can influence seismic microzonation levels in a study area
- 218 (Table 1).
- 219
- 220 Table 1.Relevant criteria that influence on seismic microzonation
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- 223 2.3.2. Experts' Knowledge data
- a) Interviewing disaster managers (semi-structured interviews) to determine the
 important criteria

The most important criteria were determined by conducting a semi-structured interview with 10 experts using the snowball sampling or chain-referral sampling method (Biernacki and Waldorf, 1981). In this study, all 10 interviewees were highly experienced and had been involved in seismic microzonation studies. The average age of the sampled individuals was 43 years, and all of them had a postgraduate degree.

231 A list of criteria that were identified by reviewing previous studies were given to the experts and they were requested to add other criteria if they thought they were applicable. They were asked 232 to rank each criterion using a five-point Likert Scale (Likert, 1932), so respondents could choose 233 234 the option that best reflected their opinion on each criterion. When surveying many people on 235 the same criterion, the five codes could be summed up, averaged or calculate the mode, indicating 236 overall positive or negative orientation towards that criterion. This was the basis from which this method was used to identify the degree of importance for each criterion in seismic microzonation 237 of ground shaking in a region. Therefore, in order to elicit the most relevant criteria, the 238 239 significance of specific factors were measured on a five-point Likert Scale where, 1 represents 'not important at all', 3 'of little importance', 5 'of average Importance', 7 'very important', and 240 9 'extremely important' (Likert, 1932; Jamieson, 2004). The collected data from experts were 241 analysed and criteria with mean ratings above '5' ('of average important') were selected (Table 242 243 2). These are considered for further analysis using the Analytic Hierarchy Process (AHP) method.





244	
245	Table 2. The average importance criteria based on 5-point Likert Scale
246	b) Interviewing disaster managers (structured interviews) in order to collect data for
247	computing the relative importance (weights) of the criteria
248	A questionnaire based on AHP matrix (A) was developed for a pair-wise comparison of the
249	relative importance of the criteria for calculating the weights (priority ranking) of each criterion.
250	As AHP is a subjective method therefore a large sample size is not needed (Cheng and Li,
251	2002;Lam and Zhao, 1998). Therefore; data were collected by interviewing 10 experts (the same
252	experts who were interviewed in the first round) based on the structured questionnaire (closed-
253	ended questions). They were asked to compare the relative importance of each criterion against
254	all others, based on Saaty's scale by verbal preferences (Saaty, 1980). A pair-wise comparison
255	that was carried out with an expert is shown in Table 3. These data are used by the AHP method
256	to compute the weight of each criterion as explain previously.
257	
258	Table 3. The results of pair-wise comparisons of the selected criteria with each other based on
259	the AHP matrix
260	
261	c) Determining fuzzy set and fuzzification of thresholds of sub-criteria for each criterion
262	In the next step, since each criterion and its sub-criteria has different effect on the seismic
<mark>263</mark>	microzonation of ground shaking level in a region, fuzzy membership functions (MFs) for sub
264	criteria of each criterion are defined in that numerical analyses of their effect would be computed.
265	As, designed parameters of each membership function depends on experts knowledge, then
266	number of memberships, the shape, the positioning, and the overlay area of memberships of each
267	MFs for each criterion would be different. To conduct this analysis, 10 experts were interviewed
268	regarding membership degree of sub criteria of each criterion, and mode of each sub criteria was
269	calculated and MFs for each criterion was depicted as descried in the following:
270	- Thickness of soil and sediments: an effective factor in site effect assessment is the thickness of
271	sediments. Rezaei et al. (2009) (2009) state that the soil thickness shows a direct relationship to
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damage rate observations in the Bam earthquake. This layer was produced by 245 geophysical, 272 273 geotechnical, and sedimentological sample sites across the city. The alluvial thickness varies in different parts of the city. In the northern part of the city, the sediment thickness ranges from 0 m, 274 275 where bedrock is exposed beneath Arg-e-Bam, to 90 m across most of the northern half of the study area. Toward the south and center of the study area, sediment thickness increases over a 276 short distance, to more than 270 m. This defines a subsurface of high sediment thickness that 277 extends across the entire study area from west to east and underlies south-central Bam. Therefore, 278 279 based on a direct relationship between the damage rate and alluvial thickness (Rezaei et al., 280 2009; Marie Nolte, 2010). MF for this criterion is depicted in figure 4a.

281

282 - Consolidation and strength of soil and sediments: It has been frequently observed that earthquake 283 damage is greater in settlements located on unconsolidated and soft soils than in those sited on 284 stiff soils or hard rock. For example, in Bam earthquake strong amplification occurred due to the extremely soft clay layers that caused high-rise buildings to collapse (Jafari et al., 2005). Another 285 286 example was the Loma Prieta earthquake that happened in 1989, where much of the damage occurred in the central San Francisco Bay area at sites underlain by thick deposits of soft clay soils 287 288 (Stewart, 1997). The soil classification has been done based on different thresholds for the average shear wave velocity (Vs) to a depth of 30m by the National Earthquake Hazard Reduction Program 289 (NEHRP) to characterize sites for purposes of estimation amplification of seismic motions. This 290 standard has applied in Unified Building Code (Dobry et al., 2000) and Eurocode8 (Sabetta and 291 292 Bommer, 2002;Kanlı et al., 2006). Based on this classification in areas on unconsolidated sediments, shear wave velocity reduces, and expected amplification during earthquakes cab be 293 increased. Therefore, according to this MFs for each class have been calculated as shown in figure 294 295 4b.

296

- Type of soil and particle size distribution of sediments: It has long been recognized that the destructiveness of ground shaking during earthquakes can be significantly worsened by the type of local soil and subsurface sediment conditions. In past events, the observed variability in seismic intensity and structural damage severity has often been attributed to the variability of soil and subsurface sediment stratigraphy in a given area. Among the geotechnical properties of soil and sediments, grain size is one of the most important criteria (Assimaki et al., 2006;Phoon et al.,





2006). In the study area, Rezaei et al. (2009) identified eight sediment types: clay, silt, sand, 303 304 granules, pebbles, cobbles, and boulders. They stated that the grain size in the shallow subsurface (<10 m) decreases across the city from south to north and increases with depth. Their investigation 305 306 showed that fine-grained soils and sediments (clay, clayey sand, cohesive sandy mud, and cohesive 307 muddy sand) dominated the northern part of the city at shallow depths. In the central part of the city, fine-grained sediments changed laterally to coarse-grained sediments (poorly sorted sand, 308 well-rounded gravel, poorly sorted gravel, and muddy or sandy gravel) which dominated in the 309 south part of the city. As a rule, it can be assumed that, the smaller the grain size of sediments, the 310 less the shear waves velocity and therefore the greater the effect of the seismic wave on the 311 312 destruction level of buildings (Rezaei et al., 2009;Assimaki et al., 2006;Phoon et al., 2006). 313 Therefore, the MFs for each specific grain size are calculated in Figure 4c.

314

315 - Depth of groundwater: Research on the effects of groundwater shows it can magnify an earthquake's damage. The most well known effect is liquefaction. The geologic and hydrologic 316 317 factors that affect liquefaction susceptibility are the age and the type of sedimentary deposits, the looseness of cohesions less sediments and the depth to the groundwater table (Tinsley et al., 1985). 318 319 The liquefaction is mostly limited to water-saturated, cohesions less sediments, and granular sediments at depths less than 15m (Iguchi and Tainosho, 1998;Sitharam, 2010). Noack and Fah 320 (2001) categorized it by the depth of the water table, which is split into three classes where the 321 weight of the class increases while the groundwater table decreases (Fah et al., 1997). Therefore, 322 323 due to the geological conditions in Bam, liquefaction is considered of minor importance because Talebian et al. (2004) and Rezaei et al. (2009) found water saturated sands in very few places, 324 however, measured microtremore data demonstrated more applification in areas with high 325 groundwater levels. Accordingly, MFs for each class of groundwater depth are computed as shown 326 327 in figure 4d.

- Type of surficial rock: Type of surficial rocks can effect on seismic microzonation level of ground
 shaking in each region. Three main types of rock based on their formation process include igneous,
 metamorphic, and sedimentary rocks. Each type has its own sub-categories and what matter in this
 research is how hard or soft and how dense the specific type of rock is in compare with the other
 types. Geological Strength Index (Geological Survey of Iran (GSI)) of "rock masses depends on





rock's material, the amount of joints and their relations, alteration, and presence of water" (Hoek 333 and Brown, 1997). There are many rock types in the nature that GSI can be calculated for any of 334 them based on their condition, and then can be fuzzified addressing their effect on seismic 335 microzonation level of ground shaking. There are five classes of GSI including very good, good, 336 fair, poor and very poor based on their surface quality and interlocking of rock pieces from 337 massive, blocky, very blocky, disintegrated, and laminated/ sheered (Marinos et al., 2007). The 338 GSI values categorized in five classes including very low, low, medium, high and very high levels. 339 These classes shows the geological strength of rocks that the high and very high GSI demonstrate 340 341 high to very high strength of rocks. Therefore, previous studies demonstrates that in massive rocks, high GSI values, seismic waves passes quickly and therefore have small influence in seismic 342 microzonation level of ground shaking, and vice versa if GSI value gets to the lower values. Thus, 343 344 in fuzzyfication process of surficial rocks, the rock with very high GIS assign 0 and the rocks with 345 very low GSI assign 1 (Figure 4e). Furthermore, the criterion of type of bedrock acts the same as surficial rock type criterion as explained above. Type of bedrock rarely changed over a small extent 346 347 with homogenous lithology. However, it was concern of experts in determining seismic microzonation of ground shaking. 348

- Slop surface; Bisch et al. (2012) reported that the effects of slope angle on topographic 349 350 amplification factor. They classified the slop angle to three categories: 0-15 with no effect, 15-30 degree with 1.2 and more than 30 degree with 1.4 amplification coefficients. Bouckovalas and 351 352 Papadimitriou (2005) investigated that the influence of slope in amplifying the peak horizontal seismic ground acceleration in front and behind the crest. Grelle et al. (2016) presented formulae 353 354 for topographic amplification on slope surface. These studies indicated that with the increase in slope angle the amplification factor would be increased. This can be a basis for depicting MFs of 355 356 this criterion (Figure 4f).

Topography irregularities: Seismic amplification has been witnessed in several earthquakes due
to topographical changes (Geli et al., 1988;Paolucci, 2002). Bisch et al. (2012) classified the site
in two classes of "isolated cliff and ridge with crest width significantly less than base width" (CEN
European Committee for Standardisation, 1994, p 93). However, this seems simplistic, as it does
not consider the elevation differences. Furthermore, Grelle et al. (2016) presented an equation that
considered the local slope height, relief height, regional share wave velocity and relief ratio. In





addition, several calibration constants should be calculated using 2d numerical analysis for each 363 364 study area to compute topographic effects on seismic microzonation of ground shaking. Lee et al. (2009) found out that the amplification on top of elevated surfaces with small extent was much 365 higher than valleys and flat areas. Therefore, the elevation differences (dH m) between the bases 366 of a hill with the top of the hill and also the area (A m^2) of the top part of the hill are the main 367 driver in computing the amount of amplification of seismic waves and can effect on seismic 368 microzonation level of ground shaking. Therefore, the higher the elevation differences and the 369 370 smaller the area of the elevated surface, the ground in this part will be more amplified. Here, using 371 fuzzy logic and experts' knowledge the effect of topography in terms of elevation differences in 372 determining seismic microzonation of ground shaking in the study area is defined (Figure 4g).

- Figure 4. Membership functions (MFs) based on fuzzy logic system: Thickness of soil and
 sediments (a), Consolidation and strength of soil and sediments (b), Type of soil and particle
 size distribution of sediments (c), Depth of groundwater (d), Type of surficial rock and
 bedrock (e), Slope surface (degree) (f), Topography irregularities (g).
- 377 2.3.3. Preparing thematic data

The required data were collected from relevant organizations and documents and they were converted to GIS files. These thematic data included: thickness of soil and sediments (Figure 5a), consolidation and strength of soil and sediments (Figure 5b), type of soil and particle size distribution of soil and sediments (Figure 5c and d), depth of groundwater (Figure 5e), type of surficial rock (Figure 5f), topography of surface (Figure 5g), and slop surface (Figure 5h) layers.

Figure 5. Thematic Layers of Bam city: Thickness of soil and sediments (m) (a), Consolidation
and strength of soil and sediments, (b), Sediment type at depth of 1 meter (c) and at depth of 9
meters (d), Groundwater level (e), Type of surficial rock(f), Topography (g) and Slop (h) layers.

387 2.3.4. Preparing control data

National Cartographic Center (2003) and Hisada et al.(2005) were collected data on the destruction
level of buildings after math of the barn earthquake (Figure 6a and b). Lashkari Pour et al. (2006)





390	and Askari et al. (2004) were collected data on the dominant frequency of soil (Figure 6c and d)
391	using microtremor measurements in Bam city. These datasets were used to validate the model.
392	
393	Figure 6. Control data: Actual building destruction level (Hisada et al., 2005) (a), percentage of
394	damage to buildings caused by the Bam earthquake in 2003 (National Cartographic Center
395	(NCC), 2003) (b), Dominant frequency by(LashkariPour et al., 2006) (c) and by (Askari et al.,
396	2004) (d) using Microtremor field measurement.
397	
398	2.3. Spatial combination methods and overlay rules
399	The spatial Multi Criteria Decision Making (MCDM) approach is a decision-aid and a
400	mathematical tool that combines and transforms spatially referenced data into a raster layer with
401	a priority score. (Roy, 1996;Malczewski, 2006). Several combination methods have been
402	developed, such as Boolean operations (Malczewski, 1999), weighted linear combination (WLC:
403	combining the normalized criteria based on overlay analysis) (Voogd, 1983;Drobne and Lisec,
404	2009;O'Sullivan and Unwin, 2010) (Eq. 5), ordered weighted averaging (OWA) (Yager,
405	1988;Rinner and Malczewski, 2002), and Analytical Hierarchy Process (AHP) based on the
406	additive weighting methods (Zhu and Dale, 2001). In this research, the AHP method (Saaty,
407	1980) was used to derive the weights associated with criteria and Fuzzy Logic method was
408	applied to compute sub-criteria's membership functions (MFs) in order to produce the seismic
409	microzonation of ground shaking. Then, the degree of membership of each sub-criteria
410	(calculated by Fuzzy Logic method) is assigned to the corresponding sub-criteria. Next, this is
411	multiplied by the weight of corresponding criteria (calculated by AHP method). Finally, they are
412	summed up in a linear manner using WLC method (Eq. 5) to develop the model (Larzesh model)
413	for production of the seismic microzonation of ground shaking in the study area.

$$Ai = \sum W_j * X_{ij} \tag{Eq. 5}$$

414 Where: w_j = the calculated weight of criteria j, and Xij = the degree of memebrship of the ith

sub-criteria with respect to the *j*th criteria, and *Ai* = the seismic microzonation of ground

416 *shaking index in ith location.*

417

418 **2.4.** Validation and comparison methods





419 In order to validate the model, as categorical variables are the main driver of model development in

- 420 this research, therefore relevant measures such as Overall Accuracy and Kappa statistic will be
- 421 applied to measure the performance of the model.

422 a) Overall accuracy (OA)

423 Accuracy assessments determine the quality of the results derived from data analysis or a model, in comparison with a reference or ground truth data (where ground truth data are assumed to be 424 100% correct) (Congalton and Green, 2009). The accuracy assessment can be obtained by 425 creating a contingency table of counts of observations, with calculated, estimated or predicted 426 data values as rows and with reference data values as columns. The values in the shaded cells 427 428 along the diagonal represent counts for correctly classified observations, where the reference data matches the predicted value. This contingency table is often referred to as a confusion matrix, 429 misclassification matrix, or error matrix (Czaplewski, 1992;Congalton and Green, 2009) (Eq. 6). 430

$$OA = \frac{\sum_{k=1}^{q} n_{kk}}{n} \times 100 \tag{Eq. 6}$$

431

432 Where: OA = Overal Accuracy, $n_{kk} = Values$ in diagonal cell of the matrix (correctly classified 433 observations), and n = number of observations.

434 b) Kappa analysis

The kappa statistic (κ) (Sim and Wright, 2005;Congalton and Green, 2008) calculates degree of agreement between classes of two independent observe measuring the same property. The degree of Kappa would be 0 for a random classifies and 1 for classification. Degree of agreement of Kappa interprets as follows: less than 0.4: poor agreement, 0.4 and 0.8: moderate agreement, and greater than 0.80: strong agreement (Congalton and Green, 2008) (Eq. 7).

440

441
$$k = \frac{P_o - P_e}{1 - P_e}$$
 (Eq. 7)

Where: Po = the relative observed agreement among raters, Pe = the hypothetical probability of chance
agreement.





445 **Results and discussion**

446 In order to produce the seismic microzonation of ground shaking the most important criteria were identified and then were weighted using AHP pair-wise comparison method. The higher weight 447 belong to thickness of soil and sediments (0.271), consolidation and strength of soil and 448 sediments (0.207), type of soil and particle size distribution of sediments (0.177), depth of 449 groundwater (0.171), topography of surface (0.054), type of surficial rock (0.041), slop surface 450 (0.040), and type of bedrock (0.040) were considered. Then, based on Fuzzy Logic method sub-451 criteria of each criterion was fuzzified and membership functions for them was defined. Next, 452 these criteria were combined based on the Weighted Linear Combination (WLC) (Drobne and 453 Lisec, 2009) in GIS to develop the model for producing the seismic microzonation of ground 454 shaking map of the study area, as it is proposed in the following (Eq. 8): 455

456

457
$$A_{j} = \sum (wS_{s} \cdot FS_{ss}) + (wT_{A} \cdot FS_{TA}) + (wS_{A} \cdot FS_{SA}) + (wD_{GW} \cdot FS_{DGW}) + (wT_{R} \cdot FS_{TR})$$

458
$$+(wT_{Br}.FS_{TBR}) + (wT_S.FS_{TS}) + (wS_L.FS_{SL})$$
 (Eq. 8)

Where: A_i = seismic microzonation of ground shaking, weights of each criterion: wS_s = consolidation 459 and strength of soil and sediments, $wT_A = thickness$ of soil and sediments, $wS_A = Type$ of soil and particle 460 461 size distribution of sediments, $wD_{Gw} = depth of groundwater$, $wT_R = type of surficial rock$, $wT_{Br} = type of surficial rock$, wtype of bedrock, $wT_S =$ topography of surface, $wS_L =$ slop surface, and fuzzified sub-criteria of each 462 criterion: FS_{ss} = consolidation and strength of soil and sediments, FS_{TA} = thickness of soil and sediments 463 , $FS_{SA} = Tyep$ of soil and particle size distribution of soil and sediments, $FS_{DGW} =$ 464 depth of groundwater, FS_{TR} = type of surficial rock, FS_{TBR} = type of bedrock, FS_{TS} = topography of 465 surface, $FS_{SL} = slop surface$. 466

467

Figure 7 displays the resulting microzonation map of ground shaking in Bam city. The areas with high to very high susceptibility of amplification are located in the north, east and northeast part of Bam city. This is due to the widespread unconsolidated sediments, low groundwater level in combination with high sediment thickness.

In order to validate the results OA and Kappa methods were applied comparing the output of model with the measured predominant frequency (Askari et al., 2004;LashkariPour et al., 2006) in the study area. The results demonstrated 80% and 82% (Table 4a and b) for OA and 0.74 and





475	0.75 for Kappa (Table 5) indicating a good fit of the model's output with the measured data.
476	Moreover, overlaying the building destructions caused by the Bam earthquake in 2003 (Hisada
477	et al., 2005;National Cartographic Center (NCC), 2003) shows high destruction levels happened
478	in locations with high ground shaking which were located in central, north and northeast part of
479	the city.
480	
481	Figure 7. Seismic microzonation of ground shaking map of Bam city
482	
483	Table 4. Coparesion between the model's output with the measured predominant frequanecy in Bam
484	city by Askari et al. (2004) (a)c and LashkariPour et al. (2006) (b).
485	
486	Table 5. Kappa coefficient and OA
487	
488	In this study, we have focused on the site effect and local geology properties of a site that have a
489	massive influence on seismic microzonation of ground shaking in the study area. To deal with
490	related uncertainties in preparing seismic microzonation, the most important criteria were selected,
491	weighted and the fuzzified. Criteria with high uncertainty degree such as distance of active fault
492	to the site, depth and magnitude of the probable earthquake were not considered because there was
493	no possibility to exactly find out where and how an earthquake will be triggered. Therefore, only
494	the criteria with known location (x and y) and known characteristics were taken into consideration.
495	Furthermore, to deal with uncertainties Fuzzy Logic is a suitable approach as we can define
496	membership function of the effect of each criterion in the amplification of ground shaking by
497	interviewing 10 experts and obtaining expert's knowledge. This can result in realistic output
498	regarding the behavior of each criterion in ground shaking calculation.
499	The newly developed model uses AHP and Fuzzy Logic (Zadeh, 1965) to deal with complexities
500	and uncertainties in data analyses in weighting the criteria and fuzzifying the sub-criteria of each
501	criterion. Although, in studies for evaluating seismic microzonation in Bangalore (India) (Sitharam
502	and Anbazhagan, 2008), Dehli (Mohanty et al., 2007), Haldia (India) (Mohanty et al., 2007), Erbaa
503	(Turkey) (Akin et al., 2013) and Al-Madinah (Moustafa et al., 2016) only AHP method was applied
504	to weight the criteria, and none of these studies considered weighting of sub criteria for each
505	criterion even using other methods.





Few researchers have considered direct properties of influencing factors in assessing ground 506 507 shaking amplification. Even, in evaluating seismic response developed models such as SiSeRHMap v1.0 (Grelle et al., 2016) and GIS Cubic Model (Grelle et al., 2014), the researchers 508 509 have applied only lithodynamic, stratigraphic and topographic effects as influencing factors. The current research considers direct properties of each criteria and tries to manage uncertainties in 510 criteria and sub-criteria of each criterion via weighting and fuzzification process using experts' 511 knowledge and the use of direct properties of criteria. These processes can be extended in more 512 details, which are subject to more investigation in the future. 513

514

515 Conclusions

Larzesh model introduces a new method based on AHP and Fuzzy Logic rules that enables experts to produce seismic microzonation of ground shaking using direct properties of lithological, sediment-logical, geological, hydrological and topographical effects in a study area using experts' knowledge in weighting and fuzzifing criteria and sub criteria that can be readily perceived and consulted.

The application of the model was carried out in the urban area of the Bam city in Iran. The results demonstrated high to very high ground shaking amplifications were located in central, east, and northeast to north part of the city that was confirmed comparing with measured microtremor data on predominate frequency in the study area. However, as the proposed model is a spatial computational tool, the validation of output in producing seismic microzonation of ground shaking strictly dependent on the quality and preparation of input data.

527 In conclusion, the model enable disaster managers, planners, and policy makers in producing 528 seismic microzonation of ground shaking and making informed decision in urban planning and 529 designing appropriate plans for urban development, especially in areas with high seismic activities.

530

531 Acknowledgements

532 The authors would like to express their appreciation to Institute of Science and High Technology

and Environmental Sciences, Graduate University of Advanced Technology, Kerman, Iran for

financial support of this study under reference number of 7/C/95/2053.





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Tables

	Table 1.Relevant criteria that inf	luence	on seismic microzonation
1	Thickness of soil and sediments	9	Thickness of bedrock
2	Consolidation and strength of soil and sediments	10	Morphology of bedrock

	3	Type of soil and particle size distribution of sediments	11	Topography of bedrock
	4	Depth of groundwater	12	Age of alluvial and sediments
-	5	Topography of surface	13	Age of bedrock
-	6	Type of surficial rock	14	Age of surfacial rock
-	7	Slop_surface		
-	8	Type of bedrock		

Table 2. The average importance criteria based on 5-point Likert Scale

	Criteria for	Average degree of
1	Thickness of soil and sediments	8.5
2	Consolidation and strength of soil and sediments	8
3	Type of soil and particle size distribution of sediments	7.5
4	Depth of groundwater	7.25
5	Type of surficial rock	7
6	Topography of surface	5.25
7	Slop surface	5
8	Type of bedrock	5
9	Thickness of bedrock	4.5
10	Morphology of bedrock	4.5
11	Topography of bedrock	4.5
12	Age of alluvial and sediments	3.75
13	Age of bedrock	3.25
14	Age of surfacial rock	2.75





787	Table 3. The results of pair-wise comparisons of the selected criteria with each other based on
788	the AHP matrix

Criteria	1	2	3	4	S	6	7	8	Weights
1-Thickness of soil and sediments	1	1	2	2	5	5	7	4	0.271
2-Consolidation and strength of soil and sediments		1	1	1	5	4	5	5	0.207
3-Type of soil, and particle size distribution of sediments			1	1	5	5	5	7	0.177
4-Depth of groundwater				1	5	7	3	5	0.171
5-Type of surficial rock					1	2	1/2	1/2	0.041
6-Topography of surface						1	1/2	3	0.054
7-Slop surface							1	4	0.040
8-Type of bedrock								1	0.040
Lambda = 8.60 CI = 0.05									

Table 4. Coparesion between the model's output with the measured predominant frequanecy in Bam city by Askari et al. (2004) (a) and LashkariPour et al. (2006) (b).

		a)							
	Predominant Frequency (Measured)								
Predicted		1	2	3	4	5	Total		
1				1				1	
2			2					2	
3		1		1	1	1		4	
4					7			7	
5						9		9	
Total		1	2	2	8	10		23	
$Av_Ac = 82 \%$		1	2	2	ð	10		4	

b)									
	Predominant Frequency (Measured)								
Predicted	1	2	3	4	5	Total			
1	1						1		
2		1					1		
3			3				3		

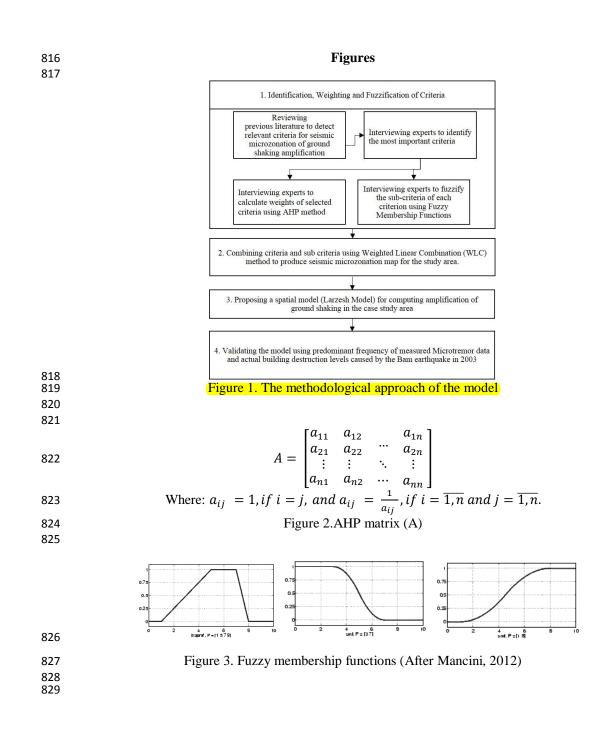


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		4					1		1				
		5		1			1	2	4				
		Total		2	1	3	2	2	10				
	Av_Ac = 80 %												
796													
797 798			Table 5 Varra as	off:		at a	nd (•					
798	Table 5. Kappa coefficient and OA Comparison of the Predominant frequency Predominant frequency												
	model's output	t and	(Askari et al., 200)4)	5			(LashkariPour et al., 2006)					
	measured da	nta											
	Kappa coefficie	ent	0.74 (0.000)				0.75 (0.000)						
700	OA		82%					80%					
799 800													
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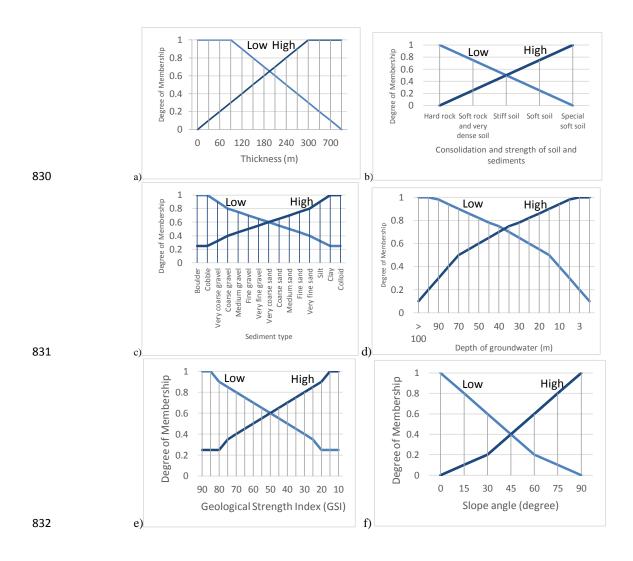






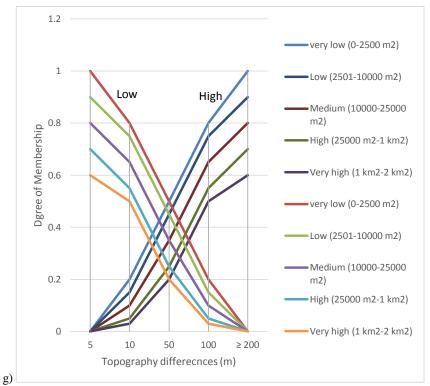


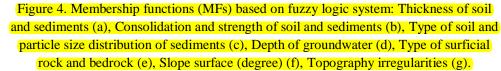


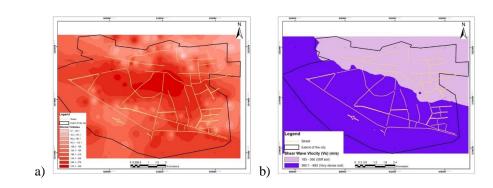








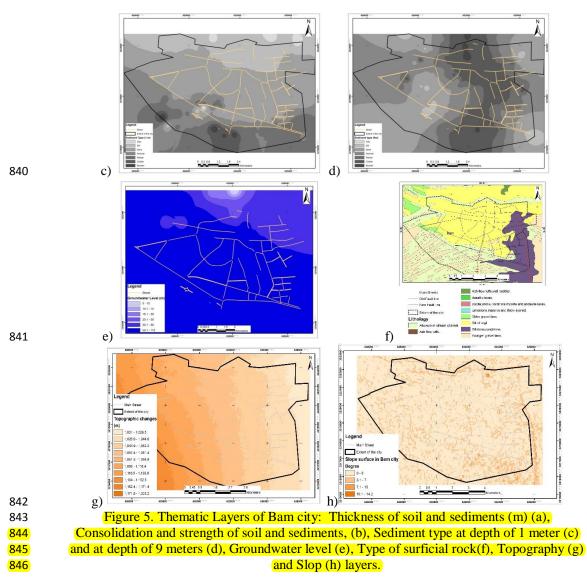




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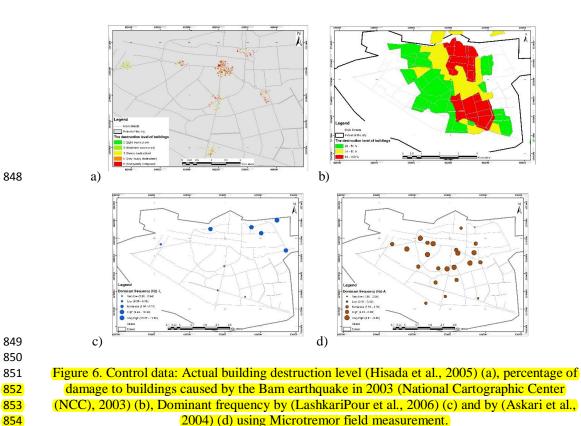




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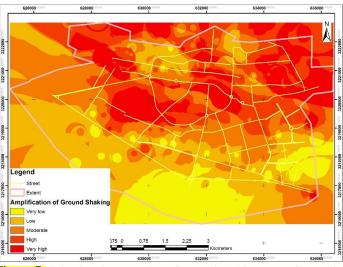


Figure 7. Seismic microzonation of ground shaking map of Bam city