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# DETERMINISTIC AND FUZZY-BASED METHODS TO EVALUATE COMMUNITY RESILIENCE BASED ON THE PEOPLES FRAMEWORK

## ABSTRACT

Community resilience is becoming a growing concern for authorities and decision makers. This paper introduces two indicator-based methods to evaluate the resilience of communities based on the PEOPLES framework. PEOPLES is a multi-layered framework that defines community resilience using seven dimensions. Each of the dimensions is described through a set of resilience indicators collected from literature and they are linked to a measure allowing the analytical computation of the indicator's performance. The first method proposed in this paper requires data on previous disasters as an input and returns as output a performance function for each indicator and a performance function for the whole community. The second method exploits a knowledge-based fuzzy modeling for its implementation. This method allows a quantitative evaluation of the PEOPLES indicators using descriptive knowledge rather than deterministic data including the uncertainty involved in the analysis. The output of the fuzzy-based method is a resilience index for each indicator as well as a resilience index for the community. The paper also introduces an open source online tool in which the first method is implemented. A case study illustrating the application of the first method and the usage of the tool is also provided in the paper.

**Keywords:** *Deterministic approach, resilience indicators, Fuzzy method, , PEOPLES framework, Earthquake resilience*

## 1. INTRODUCTION

The number of natural disasters and the corresponding number of people affected, with related economic losses, have shown an upward trend in the last years. This implies that communities are often not sufficiently resilient to natural catastrophes. Consequently, the concept of resilience has been deepened in the engineering field to assess the ability of a community to recover after an undesirable event. Indeed, since the adoption of the Hyogo framework in Manyena (2006), strategies involved in hazard planning and disaster risk reduction have experienced a paradigm shift from a vulnerability assessment approach to a resilience-based approach (Mayunga, 2007).

Since the concept of resilience is applicable in several disciplines, different definitions are available in the literature. Cimellaro et al. (2016a) have conducted a comprehensive review on this topic. In their work, the resilience definition provided by Bruneau et al. (2003) emerges: resilience is “the ability of social units to mitigate hazards, contain the effects of disasters when they occur, and carry out recovery activities in ways to minimize social disruption”. This definition has been later improved by Cimellaro et al. (2010) who define resilience as: “a function indicating the capability to sustain a level of functionality or performance for a given building, bridge, lifeline network, or community, over a period defined as the control time ( $T_C$ ) that is usually decided by owners, or society (usually is the life cycle, life span of the system etc.)”. Thus, resilience can be defined analytically as the area under the serviceability performance curve  $Q(t)$  of a system, normalized accordingly to the considered control time ( $T_C$ ):

$$R = \int_{t_l}^{t_r} \frac{Q(t)}{T_C} dt \quad (1)$$

where  $R$  is the resilience index,  $Q(t)$  is the system functionality at time  $t$ ,  $t_l$  is the moment when the disturbance occurs and the system functionality drops,  $t_r$  is the moment when the initial serviceability is completely recovered;  $T_C$  is the control time. The serviceability  $Q(t)$  ranges between 0% and 100% to indicate the complete absence of functionality of the service and its complete effectiveness, respectively.

1 Several solutions for measuring resilience are available in the literature (Abeling et al., 2014, Cutter et al.,  
2 2008, Cimellaro et al., 2016b, Cimellaro et al., 2015, Cimellaro et al., 2014). Liu et al. (2017) introduced a  
3 method that combines dynamic modeling with resilience analysis. Interdependent critical infrastructures have  
4 been analyzed in terms of design, operation, and control using this method by performing a numerical analysis.  
5 Kammouh et al. (2017b) have introduced a quantitative method to assess the resilience at the state level based  
6 on the Hyogo Framework for Action (UNISDR, 2011). The approach introduced was an evolution of the risk  
7 assessment concept. The resilience of 37 countries has been evaluated and a resilience score between 0 and  
8 100 has been assigned to each of them (Kammouh et al., 2017c). Cutter et al. (2014) clarified that research on  
9 measuring community resilience is still in the early stages of development. Although many attempts have been  
10 made to consolidate research on community resilience, no accepted method exists so far and there are still  
11 difficulties in developing concrete assessment approaches and reliable indicators.

12 Here, community resilience is evaluated exploiting two novel methodologies, which benefit from the  
13 PEOPLES framework (Cimellaro et al., 2016a, Renschler et al., 2010). PEOPLES is a layered framework:  
14 each dimension is divided into components. However, the framework does not identify a clear procedure to  
15 quantitatively compute resilience, but rather a qualitative assessment and description of resilience. The goal of  
16 this paper is to use the structure of PEOPLES framework to come up with a quantitative framework that allows  
17 evaluation of the resilience of communities. To do so, two different new methodologies to analytically quantify  
18 the resilience of communities are proposed. The first method is deterministic and requires data on past  
19 earthquake events in the form of indicators. This method turns a resilience index and a performance function  
20 for the community as an output. However, in specific scenarios, some indicators may be difficult to obtain and  
21 quantify, as well as the interdependency among them. In order to track and represent such uncertainties,  
22 another method based on a fuzzy-logic modeling is proposed. This method does not require deterministic data  
23 but rather expert knowledge to determine the different parameters involved in the resilience evaluation. It also  
24 accounts for the uncertainties involved in the assessment process. This method returns a resilience index for  
25 each indicator and a resilience index for the analyzed community. This paper also introduces a free open  
26 source tool in which the deterministic resilience method is implemented. A case study illustrating the use of  
27 the tool is also presented.

## 28 **2. PEOPLES: A COMPREHENSIVE MULTI-LAYERED RESILIENCE FRAMEWORK**

29 PEOPLES is a framework for identifying the different resilience characteristics of a community (both in time  
30 and space) and for providing new ways through which decision makers can take actions under emergencies  
31 (Cimellaro et al., 2016a). The framework allows modeling possible responses of a community considering the  
32 interdependency between the different community layers. The acronym PEOPLES stands for seven community  
33 dimensions including:

- 34 1 Population and demographics: it includes parameters that describe the social-economic composition of the  
35 community. This dimension measures the social vulnerability that could hinder the functionality of the  
36 emergency and recovery systems (e.g. population density, age distribution, presence and integration of  
37 minorities and socio-economic status).
- 38 2 Environment and ecosystem: it estimates the capability of the environment and of the ecosystem to get  
39 back to its pre-hazard conditions. It includes water, air and soil assessments as well as a measure of the  
40 biodiversity and the sustainability relations.
- 41 3 Organized government services: it covers the services that the government guarantees before and after an  
42 extreme event. A great importance is given to the mitigation and recovery processes, which include the  
43 preparedness to hazards and all disaster risk reduction measures.
- 44 4 Physical infrastructure: it considers the buildings and facilities that are the prevalent interests of civil  
45 engineers and traditional resilience analysis. Particularly, two different aspects are analyzed in this  
46 dimension: facilities, which includes housing and services which are not crucial for the emergency  
47 response, and lifelines, which instead consists of the services that are of vital importance for the  
48 management of critical situations.
- 49 5 Lifestyle and community competence: this dimension takes into account the capability of a community to  
50 face problems by means of political partnerships. This includes both the abilities of a community (i.e. the

1 skills of their components) and its perceptions (i.e. the judgements and feelings that a community has on  
2 itself).

3 6 Economic development: it describes the economic situation of the community. It can be easily divided in  
4 two terms, a static component, which measures the present economic condition, and a dynamic one, which  
5 instead takes into account the development and economic growth of the community.

6 7 Social-cultural capital: this last dimension contains an evaluation of the community's attitude to react to  
7 disasters and to return to the pre-event conditions. It includes a lot of subcategories that measure the  
8 people's commitment in the community and the social-cultural heritage.

### 9 3. METHOD 1: INDICATOR-BASED DETERMINISTIC APPROACH

#### 10 3.1 *PEOPLES' hierarchy: dimension, components, indicators, and measures*

11 PEOPLES is a multi-layered framework containing seven dimensions, each of which is divided into a set of  
12 components. To quantify the PEOPLES framework, a list of 115 resilience indicators describing the different  
13 aspects of a community has been identified and allocated to the PEOPLES' components. A full list of the  
14 components and indicators is provided in the free tool web page (more details are provided later in the paper).  
15 Each indicator is accompanied with a *measure* to allow the analytical evaluation of the indicator's  
16 performance. The measures are presented in the form of continuous functions instead of scalar values (crisp  
17 values). This allows identifying the performance of the indicator during an interval of time (i.e. the period  
18 following the disaster) rather than at a specific instance of time. Each measure is normalized with respect to a  
19 fixed quantity, the standard value (*SV*). The standard value is an essential quantity that provides the baseline to  
20 measure the resilience of a system. The system's existing serviceability at any instance of time is compared  
21 with the standard value to know how much serviceability deficiency is experienced by the system. In addition,  
22 measures are classified in two different categories: "static measures (*S*)", assigned to the measures that are not  
23 affected by the disastrous event, and "dynamic measure (*D*)" or event-sensitive measures, assigned to the  
24 measures whose values change after a hazard takes place.

#### 25 3.2 *Interdependency factor*

26 Interdependencies between the different variables of PEOPLES framework can highly affect the resilience  
27 result. Generally, the interdependency depends on several factors such as the disaster event and the type of  
28 analyzed community (rural, urban and industry). To include interdependencies, weighting factors are allocated  
29 to each variable through an interdependency analysis. In the analysis, the variables of PEOPLES are classified  
30 in three major groups as follows:

- 31 1. Indicators that fall within a component are considered as a group (totally 29 groups);
- 32 2. Components classified under a dimension are taken as a group (totally 7 groups);
- 33 3. PEOPLES seven dimensions fall in one group (totally 1 group).

34 A square matrix for each group of variables is created (Figure 1), where each cell in the matrix represents the  
35 level of interdependency between two variables. Each cell ( $a_{ij}$ ) can take the values 0 or 1 indicating full  
36 independence and full interdependence, respectively. This value can be identified either using descriptive  
37 knowledge in the form of a questionnaire filled by a group of experts or by relating to past data on previous  
38 events. Figure 2 shows a sample questionnaire to portray, as an example, the interdependencies that exist  
39 among the indicators under the component "lifeline". This questionnaire form can be filled at least by one  
40 expert who has enough knowledge about the dynamics that exist between the variables. The expert  
41 responsibility is to identify whether two indicators have "low" or "high" correlation. These descriptors are  
42 translated to 0 and 1 respectively in the matrix cells.

43 [Figure 1. near here]

44 [Figure 2. near here]

45

1 The interdependency matrix is not symmetrical because if a variable  $i$  is dependent on a variable  $j$ , the reverse  
 2 is not necessarily true. The interdependency factor of a variable  $i$  is obtained by normalizing the summation of  
 3 the cells' values in column  $i$  with respect to maximum value among all the columns' summations. The  
 4 interdependency factor for a variable  $i$  is mathematically calculated as follows:

$$\lambda_i = \frac{\sum_{j=1}^n a_{ji}}{\max(\sum_{j=1}^n a_{j1}, \dots, \sum_{j=1}^n a_{jn})} \quad (2)$$

6 where  $\lambda_i$  is interdependency factor for variable  $i$ ,  $a_{ji}$  is the interdependency level that variable  $j$  has on a  
 7 variable  $i$ ,  $n$  is the number of variables in the studied group. To better consider the uncertainties and reduce the  
 8 subjectivity, the questionnaire can be filled by a group of experts. In this case, a statistical analysis is carried  
 9 out. A probability distribution function (PDF) is considered for each variable based on the data collected from  
 10 the experts (Figure 2), and three values are used in the subsequent analysis to address the uncertainties in final  
 11 resilience output: (1) the mean ( $\sigma$ ), (2) mean + standard deviation, (3) mean – standard deviation (Figure 3).  
 12 A total of 37 matrices are needed to perform a complete interdependency analysis at a community level using  
 13 the hierarchy of PEOPLES framework.

14 [Figure 3. near here]

### 15 3.3 Importance factor

16 Variables do not contribute equally to the overall resilience output. Thus, each of the dimensions, components  
 17 and indicators is given an importance factor ( $I$ ) ranging from 1 to 3, where 1 means low importance and 3  
 18 means high importance. This factor represents the extent to which a variable (component, sub-component, or  
 19 indicator) contributes towards achieving resilience. This factor can be chosen by experts or decision makers.

### 20 3.4 Weighting factor

21 The final weighting factor for each variable ( $w_i$ ) is calculated considering both interdependency and  
 22 importance factors. Equation 3 translates an interdependency factor ( $\lambda_i$ ) and importance factor ( $I_i$ ) of variable  $i$   
 23 into a final weighting factor ( $w_i$ ).

$$w_i = \frac{\lambda_i \cdot I_i}{\text{mean}(\lambda_1 \cdot I_1, \dots, \lambda_n \cdot I_n)} = \frac{\lambda_i \cdot I_i}{\sum_{j=1}^n \lambda_j \cdot I_j} \cdot n \quad (3)$$

25 where  $w_i$  is weighting factor of variable  $i$ ,  $I_i$  is importance factor of variable  $i$ ,  $\lambda_i$  is interdependency factor of  
 26 variable  $i$ ,  $n$  is the number of variables in the studied group.

### 27 3.5 Deriving the final resilience curve

28 After obtaining weighting factors for the variables of the PEOPLES framework, a serviceability function is  
 29 built for each variable: uniform function for event-non-sensitive measures “static measures”, and non-uniform  
 30 function for event-sensitive measures “dynamic measures”, as shown in Figure 4. The serviceability function  
 31 can be defined using a set of parameters that mark the outline of the serviceability function (e.g. initial  
 32 serviceability  $q_0$ , post disaster serviceability  $q_1$ , restoration time  $T_r$ , recovered serviceability  $q_f$ ). These  
 33 parameters can be obtained from the past events and/or by performing a hazard analysis specific to each  
 34 variable. In addition, the shape of restoration curve (sometimes referred to as *slope* or *rapidity*) during the  
 35 recovery affects the resilience quantity and therefore it should be taken into account in the resilience  
 36 computation. However, the restoration rapidity depends on many variables such as the spatial dimension, the  
 37 temporal dimension, the hazard type, the available resources (including financial and human resources), the  
 38 restoration plan, etc. Thus, modeling the restoration curve of a single system is complex and it could be  
 39 defined graphically in countless shapes (Kammouh et al., 2017b). Different types of restoration curves such as  
 40 linear, exponential, step function, trigonometric and random function can be selected based on the available

1 system information. For example, the exponential shape can be selected when the initial speed recovery is high  
2 due to an initial inflow of resources and it decreases as the recovery reaches the end (Kafali and Grigoriu,  
3 2005). HAZUS (FEMA, 2011) adopted the linear trend as a restoration curve that is generally used when there  
4 is not enough available data regarding the system resources and recovery plans (Whitman et al., 1997). In this  
5 study, as not much information about the restoration rapidity is available, the linear shape for restoration curve  
6 is selected to build the serviceability function. All serviceability functions are weighted using the weighting  
7 factors described before.

8 [Figure 4. near here]

9 Prior to obtaining the weighted serviceability function for each indicator, the final resilience function is  
10 obtained through a hierarchical aggregation procedure (Figure 5). The average of the weighted serviceability  
11 functions of the variables in the same group is considered to move to an upper layer. That is, to obtain the  
12 serviceability function of component  $i$ , the average of the weighted serviceability functions of the indicators  
13 under component  $i$  is considered. Similarly, to obtain the serviceability function of dimension  $j$ , the average of  
14 the weighted serviceability functions of the components under dimension  $j$  is considered. Finally, the  
15 serviceability function of the community is the average of the weighted serviceability functions of the seven  
16 dimensions. The resilience index of the community is then evaluated as the area under the final serviceability  
17 function using Equation 1.

18 [Figure 5. near here]

### 19 3.6 Open source online tool

20 The use of the open source online tool (<http://borispio.ddns.net/PEOPLES/login.php>) is illustrated here. A  
21 Login/Register window appears when accessing the tool (Figure 6a). The user must register prior to using the  
22 tool. Once registered, the user can start a new scenario for which the resilience is to be evaluated (Figure 6b).  
23 The scenario is composed of two main parts: (1) the analyzed community (i.e. city, country, etc.), and (2) the  
24 hazard considered (e.g. earthquake, tsunami, fire, etc.).

25 [Figure 6. near here]

26 After defining the scenario, a data-entry page that displays the various variables of the PEOPLES framework  
27 appears (Figure 7). On the left side of the webpage, the seven dimensions of PEOPLES are listed. A separate  
28 page for each dimension can be accessed by clicking on the dimension. For each dimension, a list of  
29 components and indicators is shown with blank spaces to insert the data of the parameters required for the  
30 resilience evaluation. A pop-up description is triggered when hovering the mouse over a parameter in the  
31 window. This is to get extra information that helps the user identify what kind of information they need to  
32 insert. The parameters involved in the resilience evaluation are:

- 33 • Importance factor ( $I$ ): a value between 1 and 3 representing the weight of the indicator towards the  
34 resilience output (Note: in the current version of the tool only the importance factor procedure is  
35 implemented. The new version will include the interdependency analysis described above)
- 36 • Indicator nature ( $Nat$ ): the indicators are classified according to their nature: “Static ( $S$ )”, assigned to  
37 the measures that are not affected by the disastrous event, and “Dynamic ( $D$ )” or event-sensitive  
38 measures, assigned to the measures whose values change after a hazard takes place;
- 39 • Un-normalized serviceability before the event ( $q_{0u}$ ): is the un-normalized initial serviceability of the  
40 measure;
- 41 • Standard value ( $SV$ ): represents the optimal quantity for the indicator in order to be considered as fully  
42 resilient;
- 43 • Normalized serviceability before the event ( $q_0$ ): is the normalized initial serviceability of the measure.  
44 It is obtained automatically by the tool by dividing the un-normalized serviceability  $q_{0u}$  over the  
45 standard value  $SV$ ;
- 46 • Serviceability after the event ( $q_l$ ): The residual serviceability after the disaster. This quantity should be  
47 normalized by the user with respect to  $SV$ ;
- 48 • Serviceability after recovery ( $q_r$ ): it is the recovered serviceability, which can be equal, higher, or lower  
49 than the initial serviceability ( $q_0$ ). In this paper. The recovered serviceability  $q_r$  is assumed equal to the

1 initial serviceability  $q_0$ ;

- 2 • Restoration time ( $T_r$ ): it is the time needed to finish the recovery process. This value is usually  
3 determined using probabilistic or statistical approaches.

4 A list of importance factors ( $I$ ) has been set as default in the tool; however, the user can change the numerical  
5 values in the list according to their preference. The importance factors can be set all to “1” in case the user  
6 finds no justification to assign weights to the indicators; in this case, the indicators will be equally weighted.  
7 The nature of the indicator “*Nat*” can also be changed by the user because this parameter depends on the type  
8 of hazard and type of community considered in the analysis. If the indicator is *Static* ‘ $S$ ’, it is enough for the  
9 user to insert data about the initial serviceability of the system  $q_{0u}$ , and the standard value  $SV$ . If otherwise the  
10 indicator is *Dynamic* ‘ $D$ ’, the user should proceed and insert data about the post-event damage  $q_1$ ,  
11 serviceability level after restoration  $q_r$ , and restoration time  $T_r$ . The parameter  $q_{0u}$  is inserted as un-normalized  
12 value while the other serviceability parameters  $q_1$  and  $q_r$  have to be normalized by the user with respect to  $SV$   
13 (divide over  $SV$ ). A serviceability curve for each component is shown at the bottom of the page after inserting  
14 the indicators’ data. The serviceability curve of the analyzed dimension, which is the weighted average of all  
15 serviceability functions of the components, is also shown on the same graph.

16 After inserting the required data for all PEOPLES seven dimensions, the user will be able to see the  
17 serviceability curve of the community by clicking on the ‘The community resilience curve’ on the left side of  
18 the screen. For each of the serviceability curves, the tool automatically evaluates the  $LOR$ , which is an  
19 indicator for the serviceability loss incurred during the event.

20 [Figure 7. near here]

### 21 3.7 Case study

22 The resilience of the city of San Francisco is evaluated using the proposed method. The case study shows the  
23 applicability of the proposed methodology and not the actual evaluation of the resilience of San Francisco. The  
24 1989 Loma Prieta earthquake, which characterized by a moment magnitude of 6.9  $M_w$ , is considered as the  
25 disaster event. The introduced tool has been used to carry out the case study, but only the ‘Physical  
26 Infrastructure’ dimension is considered in the case study. Table 1 shows the list of the components and  
27 indicators that are grouped under the dimension ‘Physical Infrastructure’. Each indicator is linked to a measure  
28 that describes the indicator numerically using a set of parameters. In this study, the parameters have been  
29 obtained using open database sources (see notes under Table 1). The case study can be replicated by inserting  
30 the data in Table 1 in their corresponding online fields (see Figure 8).

31 In Table 1,  $q_{0u}$  is the un-normalized initial serviceability of the measure. The normalization of this quantity is  
32 necessary to combine it with the other measures that fall in a same group. This is done by dividing the un-  
33 normalized serviceability  $q_{0u}$  over the standard value  $SV$  described before. Right after the disaster, the  
34 serviceability function of a dynamic measure drops to  $q_1$  (see Figure 4b). In this example, the recovered  
35 serviceability  $q_r$  is assumed equal to the initial serviceability  $q_0$ . It is worth to note that not all facilities can be  
36 restored immediately after the disaster due to limitation of resources (financial, man power, etc.) and due to the  
37 lack of recovery plans. In addition, restoring some facilities is sometimes done in series with (after the  
38 completion of) other facilities, which poses some delay to the restoration process. The restoration time  $T_r$  is  
39 usually determined using probabilistic or statistical approaches. In this case study, the restoration fragility  
40 curves recently developed have been used to determine the restoration time for the different variables  
41 (Kammouh and Cimellaro, 2017a). In their work, they have introduced an empirical probabilistic model to  
42 estimate the downtime of the lifelines following an earthquake. Different restoration functions were derived for  
43 different earthquake magnitudes using a large earthquake database that contains data on the downtime of the  
44 infrastructures. The functions were presented in terms of probability of recovery versus time. The downtime  
45 corresponding to 95% of exceedance probability of recovery has been used as a deterministic downtime for the  
46 considered infrastructure. As for the rate of restoration, in this paper a linear interpolation is assumed for all  
47 the measures.

48 [Table 1. near here]

49 [Figure 8. near here]

1 Data collection was the most challenging part of the analysis since data about the serviceability of community  
2 systems is scarce and not shareable with the public. However, this does not imply that data is not available but  
3 rather is not accessible. Interested parties, such as decision makers and authorities, can use the framework with  
4 its full potential since data is usually available to them.

5 The tool combines the serviceability functions as described, while it evaluates the loss of resilience of the  
6 physical infrastructure using Equation 1. The time interval for the calculation of resilience is considered from  
7 the time that the event occurs ( $t_0=0$ ) until full recovery is achieved (i.e. the time corresponding to the instance  
8 where the curve reaches its pre-disaster level, which coincides with the maximum restoration time among all  
9 indicators;  $t_r=700$  days). The control time  $T_c$  is determined based on the user's period of interest and so it can  
10 take any value. In the tool,  $T_c$  is set equal to  $t_r$  automatically. Figure 9 shows the resilience curve of the case  
11 study obtained using the tool. The obtained *LOR* value (25.6%) corresponds only to the physical infrastructure  
12 dimension of the community. In order to have a resilience index for the whole community, the serviceability  
13 functions of other dimensions have to be similarly evaluated and combined in the same way. It is also  
14 interesting to compare the resilience of the two components *facilities* and *lifelines* shown in Figure 9. It is clear  
15 that the city of San Francisco has more problems in facilities ( $LOR=31.29\%$ ) than lifelines ( $LOR=21.85\%$ );  
16 therefore, it is suggested that the authority focuses more on enhancing their facilities.

17 [Figure 9. near here]

## 18 4. METHOD 2: SIMPLIFIED FUZZY LOGIC RESILIENCE FRAMEWORK

19 The methodology previously described can serve as a tool in preliminary decision-making processes related to  
20 natural catastrophic events. Nevertheless, this method is operable only if indicators can be numerically  
21 quantified, which may not be the case in some scenarios. In this section, a method that does not require  
22 deterministic data to compute the resilience of a community is proposed. The method exploits a fuzzy logic-  
23 based modeling of PEOPLES indicators to deal with uncertainties and missing knowledge. In the following  
24 sections, the fuzzy modeling of PEOPLES indicators and the evaluation of community resilience using  
25 information gathered through the fuzzy inference system are discussed. Different approaches are proposed to  
26 match different levels of complexity, starting from a two- and four-parameter approaches and ending with a  
27 full translation of the PEOPLES framework. The proposed methodologies are not fully interchangeable and so  
28 only one of them should be selected in accordance with the level of details needed.

### 29 4.1 Fuzzy logic

30 Zadeh (1965) introduced the concept of fuzzy set and the theory behind it. This theory comes with the absence  
31 of any mathematical framework that is able to describe the complexity and vagueness included in processes  
32 where human intervention is significant. While in the crisp logic the variables belong only to one class, in the  
33 fuzzy logic a variable  $x$  can be a member of several classes (fuzzy sets) with different membership grades ( $\mu$ ).  
34 Thus, each fuzzy set is characterized by a membership function that associates to any input  $x$  a real number ( $\mu$ )  
35 ranging between 0 ( $x$  does not belong to the fuzzy set) and 1 ( $x$  completely belongs to the fuzzy set) (Zadeh,  
36 1965). The strength of inference systems based on fuzzy logic relies on the following two main aspects:

- 37 • fuzzy inference systems can handle both descriptive (linguistic) knowledge and numerical data;
- 38 • fuzzy inference systems exploit approximate reasoning algorithm to formulate relationships between  
39 inputs by which uncertainties can be propagated throughout the whole process (Tefamariam and  
40 Saatcioglu, 2008a)

41 Designing a fuzzy logic-based system follows two fundamental steps: 1) defining the membership functions  
42 and the fuzzification process; 2) designing the fuzzy inference system. Fuzzy methods have been widely  
43 developed and applied in several fields (Ross, 2009). In the context of earthquake engineering, fuzzy methods  
44 have been exploited in different applications (e.g. Tefamariam and Saatcioglu, 2008b, Tefamariam and  
45 Saatcioglu, 2010, Tefamariam and Wang, 2011). Fuzzy methods have been widely used also for developing  
46 structural control systems. A clear procedure for the application of fuzzy logic can be found in Tefamariam  
47 and Saatcioglu (2008a, b).



1 A fuzzy logic system (FLS) can be defined as the nonlinear mapping of an input data set to a scalar output  
2 data. A FLS consists of four main parts: fuzzifier, rules, inference engine, and defuzzifier (Figure 10). The  
3 process of fuzzy logic is: a crisp set of input data are gathered and converted to a fuzzy set using fuzzy  
4 linguistic variables, fuzzy linguistic terms and membership functions. This step is known as fuzzification.  
5 Afterwards, an inference is made based on a set of rules. Lastly, the resulting fuzzy output is mapped to a crisp  
6 output using the membership functions, in the defuzzification step.

7 [Figure 10. near here]

#### 8 4.1.1 Fuzzification

9 The basic input parameters have a range of values that can be clustered into linguistic quantifiers, for instance,  
10 very low (VL), medium low (ML), medium (M), medium high (MH) and very high (VH). The process of  
11 assigning linguistic values is a form of data compression and it is called *granulation*. The fuzzification step  
12 converts the input values into a homogeneous scale by assigning corresponding membership functions with  
13 respect to their specified granularities (Tesfamariam and Saatcioglu, 2008b).

14 Membership functions are used in the fuzzification and defuzzification steps of a FLS, to map the non-fuzzy  
15 input values to fuzzy linguistic terms and vice versa. A membership function is used to quantify a linguistic  
16 term. Note that, an important characteristic of fuzzy logic is that a numerical value does not have to be  
17 fuzzified using only one membership function. In other words, a value can belong to multiple sets at the same  
18 time. There are different forms of membership functions such as triangular, trapezoidal, piecewise linear,  
19 Gaussian, or singleton. The most common types of membership functions are triangular, trapezoidal, and  
20 Gaussian shapes. The type of the membership function can be context dependent and it is generally chosen  
21 arbitrarily according to the user experience (Mendel, 1995).

#### 22 4.1.2 Fuzzy rules

23 The fuzzy rule base (FRB) is derived from heuristic knowledge of experts or historical data to define the  
24 relationships between inputs and outputs. The most common type is the *Mamdani* type, which is a simple *if-  
25 then* rule with a condition and a conclusion. For instance, considering two inputs, the  $i^{\text{th}}$  rule has the following  
26 formulation:

$$27 \quad R: \text{if } x_1 \text{ is } A_1 \text{ and } x_2 \text{ is } A_2 \text{ then } y \text{ is } B \quad (4)$$

28 where  $R$  is the rule number,  $x_1$  and  $x_2$  are the inputs variable,  $A_1$  and  $A_2$  are input sets,  $y$  is the output and  $B$  is  
29 the output set. The completeness of a fuzzy model is determined by the description of the behaviour for all  
30 possible input values and requires a large number of rules. The rule base is the union of all the rules:

$$31 \quad R = \bigcup_{i=1}^n R_i = R_1 \text{ also } R_2 \text{ also } \dots \text{ also } R_n \quad (5)$$

32 In some cases it is possible to regulate the degree of influence of each rule on the final output. This can be  
33 done by adding weightings based on priority or consistency, in a static or in a dynamic way.

#### 34 4.1.3 Fuzzy inference system (FIS)

35 After evaluating the result of each rule, the results are combined to obtain a final output. This process is called  
36 *inference*. Several accumulation methods can be used to combine the results of the individual rules. The  
37 maximum algorithm is generally used for accumulation. The evaluations of the fuzzy rules and the  
38 combination of the results of the individual rules are performed using fuzzy set operations. The operations on  
39 fuzzy sets are different with respect to the operations on non-fuzzy sets.

#### 1 4.1.4 Defuzzification

2 After the inference step, the overall result is a fuzzy value. This result should be defuzzified to obtain a final  
3 crisp output. This is the purpose of the defuzzifier component of an FLS. The defuzzification represents the  
4 inverse of the fuzzification process. It is performed according to the membership function of the output  
5 variable. There are several techniques to perform the defuzzification such as centre of gravity, centre of area,  
6 and mean of maximum methods.

#### 7 4.2 Two-parameter approach

8 This approach adopts only two of the four serviceability parameters described before, namely serviceability  
9 initial drop  $q^*$  (previously referred to as  $q_0$ ) and recovery time  $T^*$  (previously referred to as  $t_0$ ). Fuzzy  
10 parameters have been chosen based on the research by Bruneau et al. (2003) who describes the resilience of a  
11 system using the following three indicators: reduced failure probability; reduced consequences from failure;  
12 reduced time to recovery. The reduced failure probability has not been taken into account as it is not easily  
13 related to the herein adopted mathematical definition of resilience, which considers only the failure  
14 consequence  $q^*$  and the repair time  $T^*$ . Figure 11 presents a hierarchy of the two-parameter approach where  
15 both *time* and *initial drop* variables are used as inputs for the fuzzy system. The inputs are combined using a  
16 set of rules to obtain the output variable *fuzzy resilience*. The fuzzy output is defuzzified to get a crisp value  
17 that serves as a resilience index for the corresponding indicator.

18 [Figure 11. near here]

##### 19 4.2.1 Evaluating initial serviceability drop ( $q^*$ )

20 Two trapezoidal membership functions can be reasonably adopted in the present case. They are termed as  
21 “High” and “Low”. The fuzzification used for  $q^*$  is [High; Low]  $\rightarrow$  [(0, 0, 0.3, 0.7); (0.3, 0.7, 1, 1)]. The  
22 membership functions are graphically shown in Figure 12.

23 [Figure 12. near here]

##### 24 4.2.2 Evaluating recovery time ( $T^*$ )

25 When speaking of recovery, the intention is *full* recovery. Outperforming, or non-complete recovery, as  
26 indicated by Cimellaro et al. (2010), are not generally predictable and therefore they are not included here. For  
27 the time variable  $T^*$ , three membership functions are suggested by the authors, namely: “short”; “long”; and  
28 “very long”. The time variable is normalized based on a 3-year time span, which is normally the time reference  
29 for civil applications (i.e., 3 years corresponds to 1 on the horizontal axis). Figure 13 shows the membership  
30 functions chosen by the authors. The membership functions are not symmetrical as they have been constructed  
31 to the favor of the “Long” and “Very Long” memberships. That is, high range of values of the restoration time  
32  $T^*$  variable corresponds to the membership functions “Long” and “Very Long”.

33 [Figure 13. near here]

34 The aim is to translate the given input variables ( $q^*$  and  $T^*$ ) into one resilience measure  $R$ . This measure is itself  
35 fuzzy and so it is defined by a membership function. The chosen membership functions are depicted in Figure  
36 14. Following the fuzzy approach, it is possible to define an output value calculated from the provided inputs  
37 on basis of a set of rules. The rules adopted in this study to relate the inputs and the output are shown in Table  
38 2.

39 [Figure 14. near here]

40 [Table 2. near here]

1 4.2.3 Defuzzification

2 The fuzzy output variable is translated (defuzzified) into a numerical value that serves as a measure for  
3 resilience. Different methods for defuzzification can be used (Manyena, 2006) such as center of gravity,  
4 weighted average, mean-max, center of largest area etc. The use of one method rather than another is  
5 dependent on the application. Here, the center of gravity method given in Equation 6 is used.

6 
$$CoA = \frac{\int_{x_{min}}^{x_{max}} f(x).x dx}{\int_{x_{min}}^{x_{max}} f(x) dx} \quad (6)$$

7 where  $f(x)$  is the function that shapes the output fuzzy set after the aggregation process and  $x$  stands for the real  
8 values inside the fuzzy set support ( $[0, 1]$ ). Practical examples on the application of the fuzzy method to  
9 several case studies can be found in Tesfamariam and Saatcioglu (2008a,b).

10 4.2.4 Importance factor

11 The fuzzy logic introduced above applies to each indicator apart. It is often the case to aggregate different  
12 indicators into a single measure (i.e. community resilience) through a hierarchical structure. Usually, indicators  
13 are not equally important because they contribute differently towards resilience and this necessitates weighting  
14 them according to their contribution. The weighting scheme used in Kammouh et al. (2017d) is here adopted.  
15 It can be performed by simply allocating an importance factor ( $I$ ) ranging between 1 and 3 to each indicator  
16 then applying the weighted average rule, as follows:

17 
$$R = \frac{\sum_{i=1}^N I_i R_i}{\sum_{i=1}^N I_i} = \sum_{i=1}^N w_i R_i \quad (7)$$

18 where  $R$  is the community resilience measure,  $R_i$  is the resilience measure of the  $i^{th}$  indicator,  $I_i$  is the  
19 corresponding importance factor (which can be scaled to preference), and  $w_i$  is the weighting factor of the  $i^{th}$   
20 indicator. The difference with what has been proposed previously in the paper or in Kammouh et al. (2017d) is  
21 that in this methodology the serviceability functions are translated into resilience values before applying the  
22 weighting method (Figure 15). This simplifies the fuzzy system as it reduces the number of variables that need  
23 to be handled.

24 [Figure 15. near here]

25 4.2.5 Interdependency factor

26 While the next step of the research is to inject an interdependency methodology within the fuzzy system (with  
27 its own membership functions); currently, interdependencies are accounted for as weighting factors that  
28 increase or decrease the contribution of each variable according to the level of interdependency of the variable  
29 with other variables. Since the introduced fuzzy-based method deals with indicators, the interdependency  
30 methodology presented previously in this paper can be applied in full potential. The interdependency  
31 coefficient can be combined with the importance factor to obtain one single weighting factor for each  
32 indicator. This can be done using Equation 3 or by simply taking the average of the two coefficients.

33

### 1 4.3 *Four-parameter approach*

2 Considering only two parameters to represent a resilience indicator may in some cases be insufficient, thus  
3 affecting the mentioned benefits of using the Fuzzy approach. Moreover, this may oversimplify the problem  
4 especially when specific information about the structure itself is available and to be added. For this reason, in  
5 certain cases it may be beneficial to build up the resilience curve from fuzzy parameters other than the  
6 recovery time and the initial drop. In fact, it has been pointed out by Comerio (2006) that further distinction in  
7 the repair time is possible. According to his work, the following parameters should be taken into account:

- 8 • Construction repair time;
- 9 • Mobilization time;
- 10 • Economic conditions of the interested region;

11 The mobilization time in particular, labeled as “irrational” in Comerio (2006) (e.g. financing, workforce  
12 availability, relocation of functions or regulatory changes), is often not properly accounted for and therefore it  
13 should be given a special attention when evaluating downtime. These three indicators may be fuzzyfied with a  
14 structure similar to the one adopted for the recovery time  $T^*$ . The result is similar to what shown previously  
15 with the only difference that new rules and membership functions are to be assigned to the new variables.  
16 When resilience measures are calculated, weighting is performed to obtain the system (community) resilience.

### 17 4.4 *Full PEOPLES*

18 Most of the concepts described previously remain valid here. The only difference is that the approach  
19 introduced in this section includes the weighting of the variables within the fuzzy system. Normally, choosing  
20 adequate weighting factors is subjective and includes uncertainty. Although the inclusion of the weighting  
21 factors within the fuzzy system may add additional complexity as more variables are considered, it is certainly  
22 beneficial as it solves the uncertainty problem related to the weighting factors. To do that, two alternatives are  
23 proposed:

- 24 • Include the importance factor in the definition of the rules governing the fuzzy logic. In other words,  
25 assign rules such that the output is strongly related to the indicators with highest importance;
- 26 • Translate the importance factor into a fuzzy variable itself and include rules for it.

27 In both cases, rules have to be adapted to account for the importance factors. In the former case, rules are  
28 firmly tight to the particular application (i.e. hard to modify and not flexible); the latter case is, in this respect,  
29 more flexible but at the cost of additional complexity since additional rules have to be added to include the  
30 effect of the importance factor. This approach will be further developed and case studies will be added in  
31 future work. Figure 16 shows the logic flow where the weighting process is included as a separate variable in  
32 the first step before fuzzification.

33 [Figure 16. near here]

## 34 5. DISCUSSION

35 The two methodologies, although applied in the same context, are used under two different conditions. The  
36 first “deterministic” method is used only when data on a previous disaster is available. Applying the  
37 methodology to a real disaster allows the user to assess the loss of resilience following that particular disaster,  
38 which may help to prepare for future disasters by focusing on the weak aspects. This method also has the  
39 potential of being probabilistic when data on many previous events are available. However, this goes beyond  
40 the scope of the paper since the methodology will be the same regardless if the data are deterministic or  
41 probabilistic. The second methodology is used for both assessment and planning when data does not actually  
42 exist (which is often the case). Of course, this method would yield less accurate results since it relies on expert  
43 judgments instead of actual data.

44 Choosing between the two methods depends on what we really need. If it is an actual assessment for a previous  
45 disaster then the first method should be used (given that data is available). If the goal is to plan for future

1 events without having data on previous events then the fuzzy-based method should be used. The results of the  
2 first method are more accurate but it can be rather challenging to obtain data.

### 3 **6. CONCLUDING REMARKS**

4 This paper introduces two methods to compute the resilience of communities based on the PEOPLES  
5 framework. An indicator-based approach has been implemented as the core engine of both methodologies. The  
6 significance of the first resilience method lies in its graphical representation, which helps decision makers take  
7 proper actions to improve their resilience. While all previous works generally provide a single index to  
8 measure community resilience, the proposed method indicates in detail whether the resilience deficiency is  
9 caused by the system's lack of robustness or by the slow restoration process. It identifies where exactly  
10 resources should be spent to efficiently improve resilience. This method has been implemented in a user-  
11 friendly tool that allows the user to insert data of the different community indicators and get a resilience curve  
12 as output. On the other hand, the second method does not need deterministic data for its implementation but  
13 rather descriptive knowledge, which can come in the form of a questionnaire. This method makes use of the  
14 fuzzy logic modeling to account for the uncertainty involved in the resilience parameters assessment. Choosing  
15 between the two methods depends on the availability of data and on the level of complexity sought. The  
16 interdependency among the resilience variables in both methods has been considered by performing an  
17 interdependency analysis, which resulted in an importance factor allocated to each variable.

### 18 **ACKNOWLEDGEMENTS**

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21 Control of Sustainable Communities during Emergencies.

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**FIGURES**

<i>Var.</i>	$V_1$	$\dots$	$V_n$	→ Variables: Group of Dimensions/ Components/Indicators
$V_1$	$a_{11}$	$\dots$	$a_{1n}$	→ Interdependency Matrix
$\vdots$	$\vdots$	$\ddots$	$\vdots$	
$V_n$	$a_{n1}$	$\dots$	$a_{nn}$	
$\lambda_i$	$\lambda_1$	$\dots$	$\lambda_n$	→ Interdependency Factor

Figure 1. Interdependency matrix for the variables in a same group.

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Name:  
 Designation:  
 Company:  
 Date:

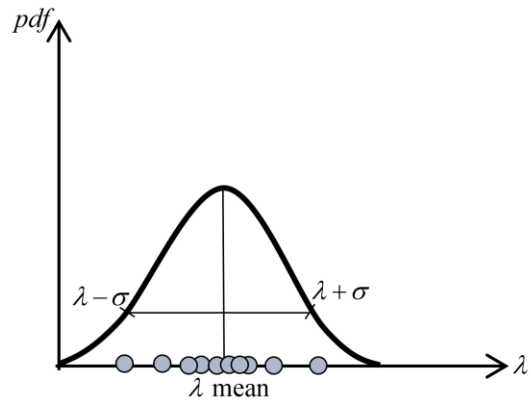
Question: please fill the following table based on your expertise. Each cell represents the level of interdependency and its cascading effect of each indicator upon the other one (across each row). Please use "L" for Low level of interdependency and "H" to display High level of interdependency.

Indicator	Telecommunication	Mental health support	Physician access	Medical care capacity	Evacuation routes	Industrial re-supply potential	High-speed internet infrastructure	Efficient energy use	Efficient Water Use	Gas	Access and evacuation	Transportation	Waste water treatment
Telecommunication	H												
Mental health support		H											
Physician access			H										
Medical care capacity				H									
Evacuation routes					H								
Industrial re-supply potential						H							
High-speed internet infrastructure							H						
Efficient energy use								H					
Efficient Water Use									H				
Gas										H			
Access and evacuation											H		
Transportation												H	
Waste water treatment													H

The level of interdependency on other indicators (read across each row)

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Figure 2. Sample questionnaire to build the interdependency matrix for indicator under component "lifeline".



●: interdependency factor ( $\lambda$ ) given by each expert

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Figure 3. Statistical analysis for the expert responses about the interdependency factor of each variable.



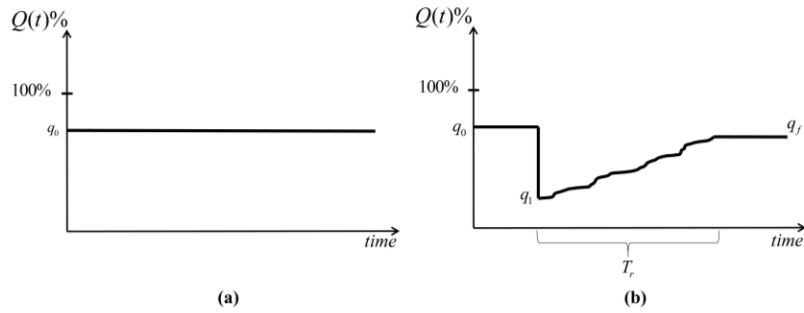


Figure 4. Serviceability functions (a) static, (b) dynamic.

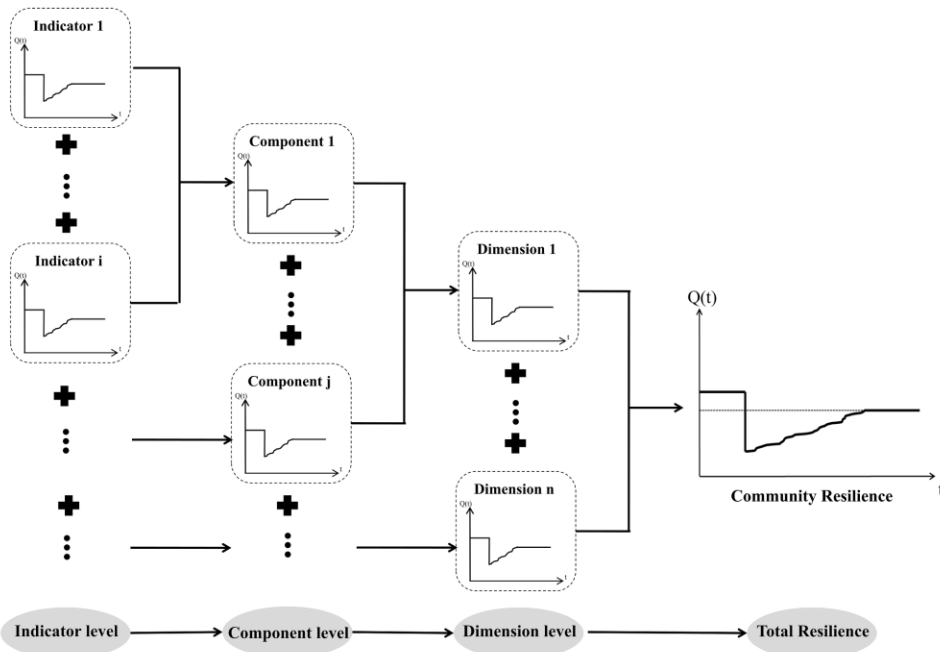


Figure 5. Hierarchical scheme of the adopted indicator-based resilience methodology.

PEOPLES Framework login

Username

Password

(a)

PEOPLES Framework

Logged as Test

Saved Scenario

OR

(b)

Figure 6. (a) Registration/login page, (b) new scenario definition/load scenario.

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### 1.1 Distribution/Density

Importance: 3

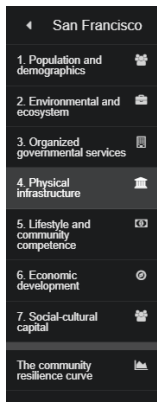
Index	Indicator	Measure	I	Nat	q <sub>0u</sub>	SV	q <sub>0</sub>	q <sub>1</sub>	q <sub>r</sub>	T <sub>r</sub>
1.1.1	Population density	1-(Average number of people per area + SV)	3	D	q <sub>0u</sub>	SV	q <sub>0</sub>	q <sub>1</sub>	q <sub>r</sub>	T <sub>r</sub>
1.1.2	Population distribution	% population living in urban area	2	D	q <sub>0u</sub>	SV	q <sub>0</sub>	q <sub>1</sub>	q <sub>r</sub>	T <sub>r</sub>

### 1.2 Composition

Importance: 2

Index	Indicator	Measure	I	Nat	q <sub>0u</sub>	SV	q <sub>0</sub>	q <sub>1</sub>	q <sub>r</sub>	T <sub>r</sub>
1.2.1	Age	% population whose age is between 18 and 65	3	S	q <sub>0u</sub>	SV	q <sub>0</sub>	-	-	-
1.2.2	Place attachment-not recent immigrants	1-(% population not foreign-born persons who came within previous five years)	1	S	q <sub>0u</sub>	SV	q <sub>0</sub>	-	-	-
1.2.3	Population stability	1-% population change over previous five year period	2	S	q <sub>0u</sub>	SV	q <sub>0</sub>	-	-	-
1.2.4	Equity	% nonminority population - % minority population	3	S	q <sub>0u</sub>	SV	q <sub>0</sub>	-	-	-
1.2.5	Race/Ethnicity	1-Absolute value of (% white - % nonwhite)	1	S	q <sub>0u</sub>	SV	q <sub>0</sub>	-	-	-

Figure 7. User interface and data entry environment.



### 4.1 Facilities

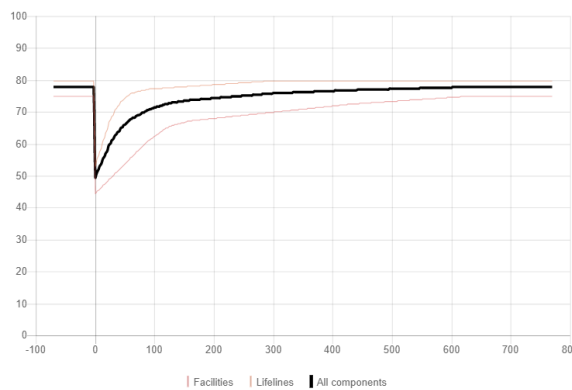
Importance: 2

Index	Indicator	Measure	I	Nat	q <sub>0u</sub>	SV	q <sub>0</sub>	q <sub>1</sub>	q <sub>r</sub>	T <sub>r</sub>
4.1.1	Sturdier housing types	% housing units not manufactured homes	3	D	1	1	1.000	0.599	0.998	120
4.1.2	Temporary housing availability	% vacant units that are for rent	3	D	2.68	5	0.536	0.05	0.536	620
4.1.3	Housing stock construction quality	100-% housing units built prior to 1970	3	D	0.241	1	0.241	0.148	0.241	700
4.1.4	Community services	%Area of community services (recreational facilities - parks - historic sites - libraries - museums) total area - SV	2	D	0.16	0.2	0.800	0.48	0.8	430
4.1.5	Economic infrastructure exposure	% commercial establishments outside of high hazard zones + total commercial establishment	2	S	0.85	1	0.850	-	-	-
4.1.6	Distribution commercial facilities	%Commercial infrastructure area per area - SV	3	D	0.13	0.15	0.867	0.520	0.867	160
4.1.7	Hotels and accommodations	Number of hotels per total area - SV	3	D	102	128	0.797	0.478	0.797	130
4.1.8	Schools	Schools area (primary and secondary education) per population - SV	3	D	134	140	0.957	0.574	0.957	90

Figure 8. Case study input data for Physical Infrastructure component.



### Chart



All components are correctly plotted

LOR Facilities:

$$\int_0^{T_r(max)} \frac{100 - Q(x)}{T_r(max)} dx = 31.29\%$$

LOR Lifelines:

$$\int_0^{T_r(max)} \frac{100 - Q(x)}{T_r(max)} dx = 21.85\%$$

LOR All components:

$$\int_0^{T_r(max)} \frac{100 - Q(x)}{T_r(max)} dx = 25.63\%$$

Figure 9. Serviceability curves of the components “Facilities” and “Lifelines” of the dimension “Physical Infrastructure”.

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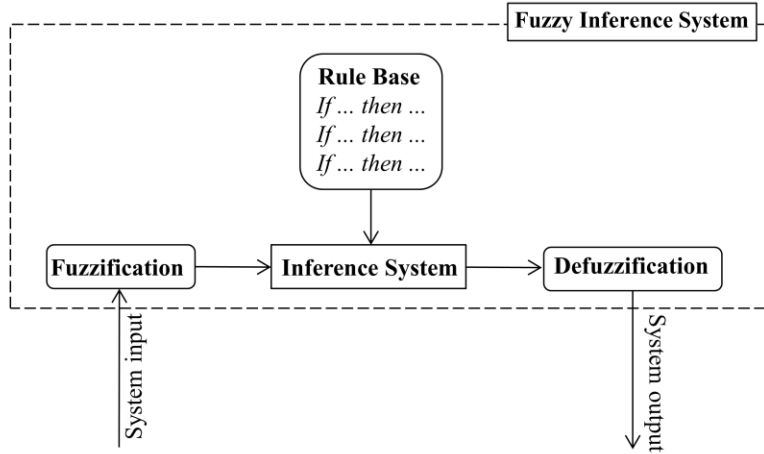


Figure 10. Fuzzy inference system.

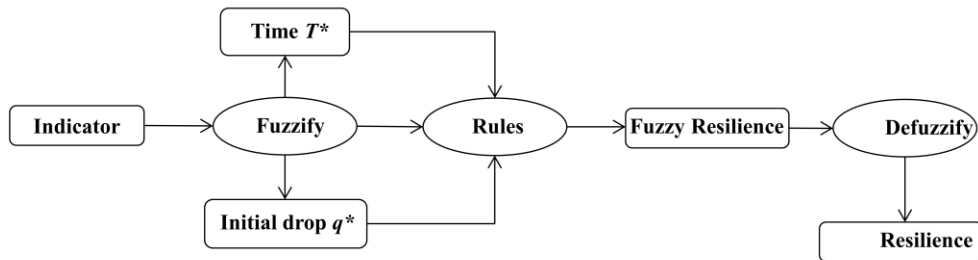


Figure 11. Schematic representation of the two-parameter approach.

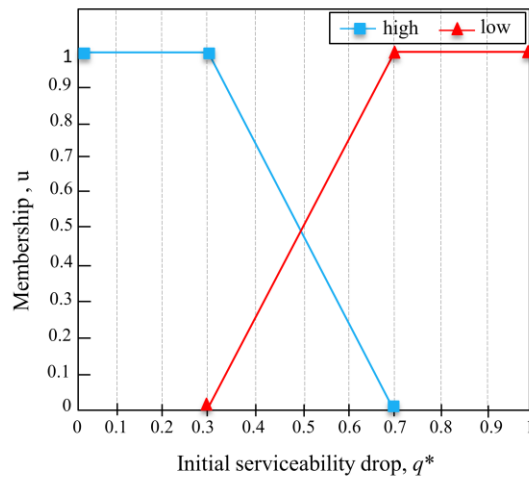


Figure 12. Membership functions for the serviceability variable  $q^*$ .

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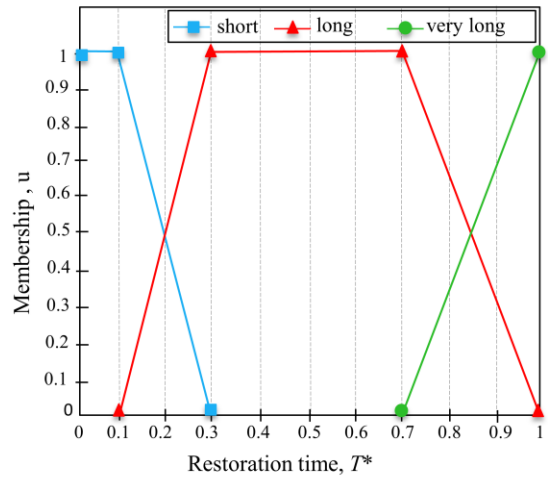


Figure 13. Membership functions for the downtime variable  $T^*$ .

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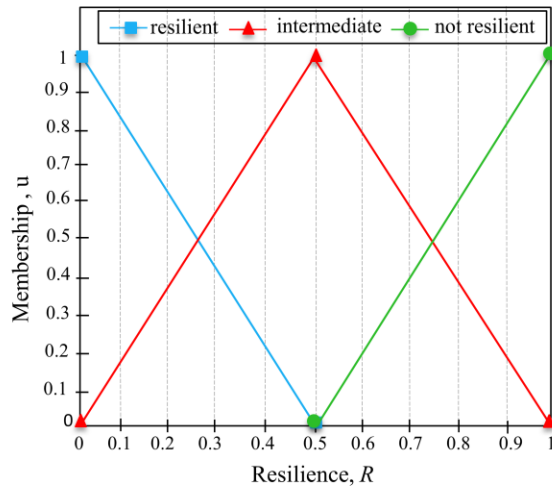


Figure 14. Membership functions for the resilience variable  $R$ .

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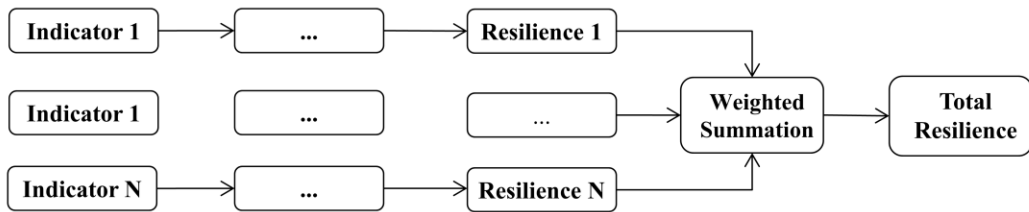


Figure 15. Hierarchical scheme of the fuzzy system with the weighting process.

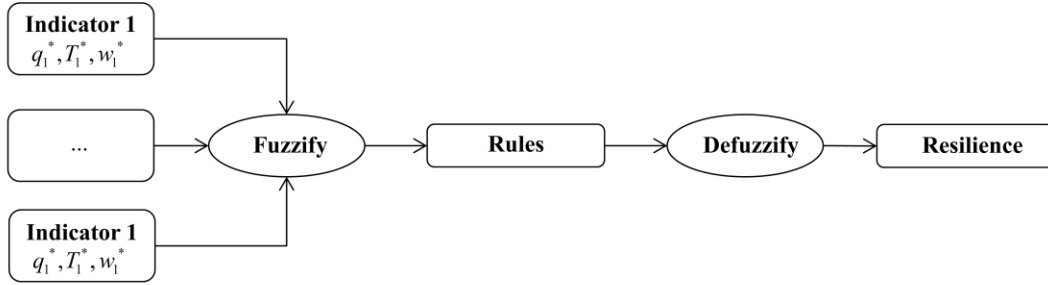


Figure 16. Full PEOPLES approach general hierarchical scheme with the weighting process included in the fuzzy system as a separate variable.

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### TABLES

Table 1. Serviceability parameters of the indicators within the Physical Infrastructure dimension for the city of San Francisco after the Loma Prieta earthquake.

4- Physical infrastructure									
Component/indicator	Measure	w	Nat	$q_{ou}$	SV	$q_0$	$q_l$	$q_r$	$T_r$ (days)
<b>4.1 Facilities</b>									
4.1.1 <i>Sturdy (robust) housing types</i>	% housing units that are not manufactured homes	3	D	1	1	1	0.599	0.998	120
4.1.2 <i>Temporary housing availability</i>	% vacant units that are for rent	3	D	2.68	5	0.536	0.050	0.536	620
4.1.3 <i>Housing stock construction quality</i>	100-% housing units built prior to 1970	3	D	0.241	1	0.241	0.145	0.241	700

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4.1.4	Community services	%Area of community services (recreational facilities, parks, historic sites, libraries, museums) total area ÷ SV	2	D	0.16	0.2	0.800	0.480	0.800	430
4.1.5	Economic infrastructure exposure	% commercial establishments outside of high hazard zones ÷ total commercial establishment	2	S	0.85	1	0.850	-	-	-
4.1.6	Distribution commercial facilities	%Commercial infrastructure area per area ÷ SV	3	D	0.13	0.15	0.867	0.520	0.867	160
4.1.7	Hotels and accommodations	Number of hotels per total area ÷ SV	3	D	102	128	0.797	0.478	0.797	130
4.1.8	Schools	Schools area (primary and secondary education) per population ÷ SV	3	D	134	140	0.957	0.574	0.957	90

#### 4.2 Lifelines

4.2.1	Telecommunication	Average number of Internet, television, radio, telephone, and telecommunications broadcasters per household ÷ SV	3	D	5	6	0.833	0.500	0.833	90
4.2.2	Mental health support	number of beds per 100 000 population ÷ SV	2	D	69	75	0.920	0.644	0.920	35
4.2.3	Physician access	Number of physicians per population ÷ SV	2	S	2.5	3	0.833	-	-	-
4.2.4	Medical care capacity	Number of available hospital beds per 100000 population ÷ SV	3	D	544	600	0.907	0.635	0.907	35
4.2.5	Evacuation routes	Major road egress points per building ÷ SV	2	S	0.67	1	0.670	-	-	-
4.2.6	Industrial re-supply potential	Rail miles per total area ÷ SV	3	D	5412	6000	0.902	0.631	0.902	45
4.2.7	High-speed internet infrastructure	% population with access to broadband internet service	3	D	0.9	1	0.900	0.450	0.900	300
4.2.8	Efficient energy use	Ratio of Megawatt power production to demand	3	D	0.8	1	0.800	0.160	0.800	25
4.2.9	Efficient Water Use	Ratio of water available to water demand	3	D	1	1	1.000	0.240	1.000	60
4.2.10	Gas	Ratio of gas production to gas demand	3	D	0.1	1	0.100	0.050	0.100	70
4.2.11	Access and evacuation	Principal arterial miles per total area ÷ SV	3	D	172138	200000	0.861	0.602	0.861	45
4.2.12	Transportation	Number of rail miles per area ÷ SV	3	D	5412	6000	0.902	0.631	0.902	72
4.2.13	Waste water treatment	Number of WWT units per population ÷ SV	3	D	3	4	0.750	0.300	0.750	65

- Note:  $q_{0i}$  = initial serviceability;  $SV$  = standard value;  $q_0$  = initial normalized serviceability;  $q_1$  = post disaster serviceability;  $q_r$  = recovered serviceability;  $T_r$  = restoration time.

- Source: City Data, Census Data, This Study, City Assessor's Data, Dept of Numbers, SF Indicator Project, Data World Bank, Dot Ca, SF Bos, Arcadis, SF Water, Energy Ca.

Table 2. Fuzzy rule base for resilience.

$T^*$	$q^*$	$R$
short	high	resilient
Long	high	resilient
Very long	high	intermediate
Short	low	intermediate
long	low	not resilient
Very long	low	not resilient

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