https://journal.unnes.ac.id/sju/index.php/rji

p-ISSN 2963-5551

e-ISSN XXXX-XXXX

Adaptive Difficulty in Earthquake Mitigation Game Using Fuzzy Mamdani

Rika Jane Ardiadna¹, Abas Setiawan^{2*}

¹Informatics Engineering Department, Faculty of Computer Sciences, Universitas Dian Nuswantoro, Indonesia ²Computer Science Department, Faculty of Mathematics and Natural Sciences, Universitas Negeri Semarang, Indonesia

Abstract. Earthquake disasters cause a lot of casualties. Therefore, needs to be education on earthquake disaster mitigation to minimize losses. In addition to counseling and teaching in schools, mitigation education can also be through games. Some education games for earthquake disaster mitigation have circulated quite a lot but have disadvantages, namely the difficulty level that hasn't been adaptive. A game requires an adaptive level of difficulty that can adjust between the ability and playing experience of the player with the level of difficulty so that players do not feel bored or frustrated.

Purpose: This study aims to provide earthquake disaster mitigation education and discuss making the level of difficulty in the game adaptive to suit the abilities and experience of the player.

Method: From the research carried out by applying the Mamdani Fuzzy Logic, the game's difficulty level for each player becomes more adaptive or different for each player according to the ability and experience of each player in the previous stage measured from 6 input parameters.

Result: The level of difficulty that is obtained becomes adaptive. It changes according to conditions or is adjusted based on the player's ability. It is from the playtesting experiment conducted on 20 players. The minimum difficulty level's score is five, and the difficulty level's score is 28.36.

Novelty: This paper's purpose is an educational game for earthquake mitigation with the feature of adaptive level based on fuzzy Mamdani.

Keywords: Earthquake Mitigation, Educational Games, Adaptive Difficulty, Fuzzy Logic, Mamdani **Received** February 21, 2023 / **Revised** March 06, 2023 / **Accepted** March 06, 2023

This work is licensed under a <u>Creative Commons Attribution 4.0 International License.</u>



INTRODUCTION

Earthquake is a natural disaster that occurs in many countries. Earthquakes are caused by shifts in the earth's plates and other biological factors where the incident causes casualties and material losses [1]. Indonesia is the one country that is prone to earthquakes. Indonesia is in the confluence of three of the world's tectonic plates that cause the thousand island countries to have high earthquake susceptibility levels and part of the Pacific belt (Ring of fire) [2]. One of the cases was the earthquake in Donggala, Central Sulawesi. The earthquake occurred on Sept. 28, 2018, at 17.02, with a scale of 7.4SR, which claimed as many as 744 people died [1], [3]. Apart from Indonesia, the region in China also experienced an earthquake on Aug. 8, 2017, in the city of Jiuzhaigou. It claimed that 25 people died, and 525 were injured [4]. After the Pacific Belt, the second most active seismic region is the Alpine-Himalayan orogenic belt. The area is located in Iran and along the Himalayan mountains. This region also includes China, India, and Nepal, which often experience earthquakes [5].

Education about disaster mitigation for the community is critical. It can minimize victims and material losses because it has many victims and a high earthquake vulnerability. Therefore, earthquake mitigation education has been carried out through outreach to the community or government [6] or teaching about earthquakes in schools [7]. In addition to outreach and education, earthquake mitigation education can be carried out through a game [8]–[12]. Therefore, video games are an excellent and exciting education medium, especially for children [13]–[15].

²*Corresponding author.

Email addresses: abas.setiawan@mail.unnes.ac.id (Setiawan) DOI: 10.15294/rji.v1i1.66543 Narrative video games are used to provide earthquake disaster mitigation knowledge to children [8]. The game also uses a User Experience evaluation with a User Experience Questionnaire. Another earthquake disaster mitigation game is physical (Snake and Ladder), specifically for children with special needs [9], [10]. The Snake and Ladder game is also applied to normal children by observing their response to natural disasters, especially earthquakes [10]. Serious web-based games are used to prepare children for earthquakes [11]. This Serious game also has a quiz feature to evaluate player knowledge. In addition, games based on Virtual Reality (VR) are used to train earthquake preparedness [12]. This game is intended not only for children but also for adults. From some of the research and games mentioned earlier, there is a drawback: the level of difficulties in games that have not been adaptive.

Game difficulty levels that are still static or not yet adaptive allow players to choose their level of difficulty, which makes it less flexible with abilities and player experience [16] so that, in the end, it can lead to boredom or frustration. It can be illustrated by the theory of flow in games that if the level of the game difficulty is too high compared to the player's ability to enter the frustration zone and vice versa [17], One technology to create an adaptive or dynamic game is fuzzy logic. Several studies related to fuzzy logic for adaptive games have been done before by Pratama et al., which is research to organize the behavior of the bots obtained based on the player's ability [18]. Futhermore, Fuzzy logic is frequently utilized to provide dynamic behavior in-game aspects, such as difficulty setting [16], player involvement assessment [19], and Non-Playable Character (NPC) behavior in player interactions [20]. In addition, fuzzy logic is frequently utilized because it is simple to implement, does not complicate in-game processing, and can result in dynamic, intelligent agent behavior [21].

To make the game exciting and not dull and the players are still in the flow zone, there needs to be an adaptive level of difficulty that can adjust to the player's ability and experience. However, our study not only focuses on the game for earthquake mitigation but also considers the game's flow. So, our study has several contributions:

- 1. We create an educational game for earthquake mitigation for all ages deployed on the Android Platform.
- 2. We design adaptive difficulty inside our game using fuzzy Mamdani to make dynamic interactions based on user experience in every game level applied to the table's movement and the obstacle falling speed.

METHODS

Game Design

This game generally tells of a child trapped in a room when an earthquake occurs. The child must be able to survive by being under the table. Meanwhile, the table used for shelter moved erratically because of the earthquake shaking. Therefore, falling obstacles must not hit him. Within a certain period, there will be a door that can be used to exit, and the player will win. However, you will lose if time or health points run out. Besides, there are also some items to increase health points. There is also a victim character that will appear randomly that the child can help. This game is named "Quake Run." All ages can play this game on the Android platform. The controls in this game are illustrated in Figure 1. Players can control the child character by swiping right and left. Besides that, players can also tap on certain items to get the effects of these items. The difficulties level on this game is environment adaptation. There is table movement during the earthquake and the obstacle falling speed. Therefore, it is challenging for a player because if the table's movements or obstacle falling speed is fast, the player has difficulty controlling the game. In our case, we use Fuzzy logic to make those adaptations.



Figure 1. Game Control Design

Input and Output Definition

In this study, the research method used is Fuzzy Mamdani logic with multilevel Fuzzy, divided into three stages, where the research was conducted with the fuzzification step first. In fuzzification, it needs to determine the input and output parameters and the membership function of each parameter (Fuzzification). The input parameters used are as follows:

1. "HPP," the parameter of the remaining health points owned by the player, has three fuzzy sets: {Weak, Medium, Strong}. Equations 1, 2, and 3 show the membership function for each fuzzy set. (1. $HPP \le 25$

$$HWeak = \begin{cases} 1, HPP \le 23 \\ \frac{50 - HPP}{50 - 25}, 25 < HPP < 50 \\ 0, HPP \ge 50 \\ 1, HPP = 50 \\ \frac{HPP - 25}{50 - 25}, 25 < HPP < 50 \\ \frac{75 - HPP}{75 - 50}, 50 < HPP < 75 \\ 0, HPP \le 25 \text{ or } HPP \ge 75 \\ 1, HPP \ge 75 \\ HStrong = \begin{cases} \frac{HPP - 50}{75 - 50}, 50 < HPP < 75 \\ 0, HPP \le 75 \\ 0, HPP < 50 \end{cases}$$
(2)

0, HPP ≤ 50
2. "Stimer" is the remaining time parameter after the game is over and has three fuzzy sets: {Low, Moderate, High} as seen in Equations 4, 5, and 6, respectively.

$$SLow = \begin{cases} 1, STimer \le 25 \\ \frac{10-STimer}{10-5}, 5 < STimer < 10 \\ 0, STimer \ge 10 \\ 1, STimer = 10 \end{cases}$$
(4)
$$SModerate = \begin{cases} \frac{STimer-5}{10-5}, 5 < STimer < 10 \\ \frac{20-STimer}{20-10}, 10 < STimer < 20 \\ 0, STimer \le 5 \text{ or } STimer \ge 20 \\ 1, STimer \ge 20 \end{cases}$$
(5)
$$SHigh = \begin{cases} \frac{STimer-10}{20-10}, 10 < STimer < 20 \\ 0, STimer \le 10 \\ 0, STimer \le 10 \end{cases}$$
(6)

3. "jObs," the parameter for the number of obstacles hit by players, has three fuzzy sets: {Low, Moderate, High}. The fuzzy set is described in Equations 7, 8, and 9.
 (1, jObs ≤ 5)

$$JLow = \begin{cases} \frac{10 - jObs}{10 - 5}, & 5 < jObs < 10\\ 0, & jObs > 10 \end{cases}$$
(7)

$$JModerate = \begin{cases} 1, \ jObs \ge 10 \\ \frac{jObs-5}{10-5}, \ 5 < jObs < 10 \\ \frac{jObs-5}{10-5}, \ 5 < jObs < 10 \\ \frac{15-jObs}{15-10}, \ 10 < jObs < 15 \\ 0, \ jObs \ge 5 \ or \ jObs \ge 15 \\ 1, \ jObs \ge 15 \\ 15-10 \\ 15-10 \\ 0, \ jObs < 10 \end{cases}$$
(8)
$$JHigh = \begin{cases} \frac{jObs-10}{15-10}, \ 10 < jObs < 15 \\ 0, \ jObs \le 10 \\ 0, \ jObs \le 10 \end{cases}$$
(9)

4. "nItemHP," which is the parameter for the number of health point items, has three fuzzy sets: {Low, Moderate, High} as shown in Equations 10, 11, and 12, respectively.

$$NLow = \begin{cases} 1, \ nItemHP \le 5\\ \frac{10 - nItemHP}{10 - 5}, \ 5 < nItemHP < 10\\ 0, \ nItemHP \ge 10 \end{cases}$$
(10)

$$NModerate = \begin{cases} 1, \ nItemHP = 10 \\ \frac{nItemHP-5}{10-5}, \ 5 < nItemHP < 10 \\ \frac{15-nItemHP}{15-10}, \ 10 < nItemHP < 15 \\ 0, \ nItemHP \le 5 \ or \ nItemHP \ge 15 \\ 1, \ nItemHP \ge 15 \\ NHigh = \begin{cases} \frac{nItemHP-10}{15-10}, \ 10 < nItemHP < 15 \\ \end{cases}$$
(11) (12)

0.
$$nItemHP < 10$$

5. "nItemTimer," which is the parameter for the number of items to add time, has three fuzzy sets: {Slight, Fair, High} as shown in Equations 13, 14, and 15, respectively.

$$NISlight = \begin{cases} 1, nItemTimer \le 5\\ \frac{10-nItemTimer}{10-5}, 5 < nItemTimer < 10\\ 0, nItemTimer \ge 10\\ 1, nItemTimer = 10\\ \frac{nItemTimer-5}{10-5}, 5 < nItemTimer < 10\\ \frac{15-nItemTimer}{15-10}, 10 < nItemTimer < 15\\ 0, nItemTimer \le 5 \text{ or } nItemTimer \ge 15\\ 1, nItemTimer \ge 15\\ NIHigh = \begin{cases} \frac{nItemTimer-10}{15-10}, 10 < nItemTimer < 15\\ 0, nItemTimer \ge 15\\ 0, nItemTimer \le 10\\ 15-10\\ 0, nItemTimer \le 10 \end{cases}$$
(13)

0, *nItemTimer* ≤ 10
"Safety" is a parameter calculated in Stage 1 using parameters HPP and STimer, showing the player's safety level. It has three fuzzy sets: {Low, Fair, and Good} as described in Equations 16, 17, and 18, respectively.

$$FLow = \begin{cases} 1, Safety \le 20 \\ \frac{40-Safety}{40-20}, 20 < Safety < 40 \\ 0, Safety \ge 40 \end{cases}$$
(16)

$$FFair = \begin{cases} 1, Safety \ge 40 \\ 1, Safety = 40 \\ \frac{Safety-20}{40-20}, 20 < Safety < 40 \\ \frac{60-Safety}{60-40}, 40 < Safety < 60 \\ 0, Safety \le 20 \text{ or } Safety \ge 60 \end{cases}$$
(17)

$$FGood = \begin{cases} 1, Safety \ge 60 \\ \frac{Safety-40}{60-40}, 40 < Safety < 60 \\ 0, Safety \le 60 \end{cases}$$
(18)

7. "Reflex" is a parameter calculated in Stage 2 using the JObs, nItemHP, and nItem Timer parameters. It has three fuzzy sets: {Fast, Normal, Slow} as seen in Equations 19, 20, and 21, respectively.

$$RFast = \begin{cases} 1, Reflex \le 20 \\ \frac{40-Reflex}{40-20}, & 20 < Reflex < 40 \\ 0, Reflex \ge 40 \end{cases}$$
(19)
$$RNormal = \begin{cases} 1, Reflex = 40 \\ \frac{Reflex-20}{40-20}, & 20 < Reflex < 40 \\ \frac{60-Reflex}{60-40}, & 40 < Reflex < 60 \\ 0, Reflex < 20 \text{ or } Reflex \ge 60 \end{cases}$$
(20)

$$RSlow = \begin{cases} 1, Reflex \ge 200 \text{ Reflex} \ge 00 \\ 1, Reflex \ge 60 \\ \frac{Reflex-40}{60-40}, \ 40 < Reflex < 60 \\ 0, \ Reflex \le 60 \end{cases}$$
(21)

8. "nSelamat," the parameter for the number of rescued victims, has three fuzzy sets: {low, medium, high} as described in equations 22, 23, and 24, respectively.

$$NSLow = \begin{cases} 1, \ nSelamat \le 5\\ \frac{10-nSelamat}{10-5}, \ 5 < nSelamat < 10\\ 0, \ nSelamat \ge 10\\ 1, \ nSelamat = 10\\ \end{cases}$$
(22)
$$NSMedium = \begin{cases} \frac{nSelamat-5}{10-5}, \ 5 < nSelamat < 10\\ \frac{15-nSelamat}{15-10}, \ 10 < nSelamat < 15\\ 0, \ nSelamat \le 5 \ or \ nSelamat \ge 15\\ 1, \ nSelamat \ge 15\\ \end{cases}$$
(23)
$$NSHigh = \begin{cases} \frac{nSelamat-10}{15-10}, \ 10 < nSelamat < 15\\ 0, \ nSelamat < 15\\ 0, \ nSelamat < 15\\ 0, \ nSelamat < 15 \end{cases}$$
(24)

The "Level" is the output parameter that will determine the game's table movement speed and obstacle motion. It has five fuzzy sets: {Very Easy, Easy, Medium, Hard, Very Hard}. The graph of the Membership function is depicted in Figure 2.



Figure 2. Graph of "Level" Membership Function

Rules Definition

Because it uses multilevel fuzzy, the rules are made according to the stage, such as Stage 1 is for Safety, Stage 2 is for Reflex, and Stage Level. Therefore, those three stages have fuzzy rules, as shown in Tables 1, 2, and 3.

Table 1. Rules for Safety							
	ID		THEN				
		HPP STimer		Safety			
	1	HWeek	SLow	FLow			
	2	HWeek	SModerate	FLow			
	3	HWeek	SHigh	FFair			
	4	HMedium	SLow	FLow			
	5	HMedium	SModerate	FFair			
	6	HMedium	SHigh	FGood			
	7	HStrong	SLow	FFair			
	8	HStrong	SModerate	FGood			
	9	HStrong	SHigh	FGood			

ID		IF		THEN
	nItemHP	nItemTimer	jObs	Reflex
1	NLow	NISlight	JLow	RFast
2	NLow	NISlight	JModerate	RFast
3	NLow	NISlight	JHigh	RNormal
4	NLow	NIFair	JLow	RFast
5	NLow	NIFair	JModerate	RNormal
6	NLow	NIFair	JHigh	RNormal
7	NLow	NIHigh	JLow	RNormal
8	NLow	NIHigh	JModerate	RNormal
9	NLow	NIHigh	JHigh	RSlow

JLow

JModerate

NISlight

NISlight

10

11

NModerate

NModerate

Table 2. Rules for Relex

RFast

RFast

	IF		THEN
nItemHP	nItemTimer	jObs	Reflex
NModerate	NISlight	JHigh	RNormal
NModerate	NIFair	JLow	RFast
NModerate	NIFair	JModerate	RNormal
NModerate	NIFair	JHigh	RSlow
NModerate	NIHigh	JLow	RNormal
NModerate	NIHigh	JModerate	RSlow
NModerate	NIHigh	JHigh	RSlow
NHigh	NISlight	JLow	RNormal
NHigh	NISlight	JModerate	RNormal
NHigh	NISlight	JHigh	RSlow
NHigh	NIFair	JLow	RNormal
NHigh	NIFair	JModerate	RSlow
NHigh	NIFair	JHigh	RSlow
NHigh	NIHigh	JLow	RSlow
NHigh	NIHigh	JModerate	RSlow
NHigh	NIHigh	JHigh	RSlow
	nltemHP NModerate NModerate NModerate NModerate NModerate NModerate NHigh NHigh NHigh NHigh NHigh NHigh NHigh NHigh NHigh NHigh NHigh NHigh NHigh NHigh	IF nltemHP nltemTimer NModerate NISlight NModerate NIFair NModerate NIFair NModerate NIFair NModerate NIHigh NModerate NIHigh NModerate NIHigh NHigh NISlight NHigh NISlight NHigh NISlight NHigh NISlight NHigh NIFair NHigh NIFair NHigh NIFair NHigh NIFair NHigh NIFair NHigh NIFair NHigh NIFair NHigh NIHigh NHigh NIHigh NHigh NIHigh	IF nltemHP nltemTimer jObs NModerate NISlight JHigh NModerate NIFair JLow NModerate NIFair JModerate NModerate NIFair JHigh NModerate NIHigh JLow NModerate NIHigh JModerate NModerate NIHigh JHigh NHigh NISlight JLow NHigh NISlight JHigh NHigh NISlight JHigh NHigh NIFair JLow NHigh NIFair JLow NHigh NIFair JModerate NHigh NIFair JHigh NHigh NIFair JHigh NHigh NIFair JHigh NHigh NIFair JHigh NHigh NIFair JHigh NHigh NIFair JHigh NHigh NIHigh JLow NHigh NIHigh JLow NHigh NIHigh JModerate NHigh NIHigh JModerate NHigh NIHigh JModerate NHigh NIHigh JHigh

ID		IF		THEN
	Safety	Reflex	nSelamat	Level
1	FLow	RSlow	NSLow	VeryEasy
2	FLow	RSlow	NSMedium	Easy
3	FLow	RSlow	NSHIgh	Easy
4	FLow	RNormal	NSLow	Easy
5	FLow	RNormal	NSMedium	Easy
6	FLow	RNormal	NSHIgh	Medium
7	FLow	RFast	NSLow	Easy
8	FLow	RFast	NSMedium	Medium
9	FLow	RFast	NSHIgh	Hard
10	FFair	RSlow	NSLow	Easy
11	FFair	RSlow	NSMedium	Easy
12	FFair	RSlow	NSHIgh	Medium
13	FFair	RNormal	NSLow	Easy
14	FFair	RNormal	NSMedium	Medium
15	FFair	RNormal	NSHIgh	Hard
16	FFair	RFast	NSLow	Medium
17	FFair	RFast	NSMedium	Hard
18	FFair	RFast	NSHIgh	Hard
19	FGood	RSlow	NSLow	Easy
20	FGood	RSlow	NSMedium	Medium
21	FGood	RSlow	NSHIgh	Hard
22	FGood	RNormal	NSLow	Medium
23	FGood	RNormal	NSMedium	Hard
24	FGood	RNormal	NSHIgh	Hard
25	FGood	RFast	NSLow	Hard
26	FGood	RFast	NSMedium	Hard
27	FGood	RFast	NSHIgh	VeryHard

RESULT AND DISCUSSION

Game Result

This game has a background of conditions at the time of the earthquake. A player will be in the room where later he will be under the table for shelter during an earthquake. The table will move randomly, and players must stay under it as much as possible so that obstacles don't hit it, which can reduce the player's health points. Players can move characters by clicking and sliding to the right or left. Then obstacles will fall due to earthquake vibrations, and players must avoid them by staying under the table. Also, a quest collects mitigation items that fall until the game quest is complete so that the door exits to the next stage appears. In addition, there are also useful items that can help players stay in the game. Figure 3 depicts the in-game interface result.

		160		
Slider It	em 6.00			
SM	SS			
	à.	8	- -	
	Nyawa	Waktu	Mempenambat	

Figure 3. In-game Interface

Fuzzy Mamdani Scenario Test

The research conducted produces a game with the theme of mitigating earthquakes whose level of difficulty is set using Fuzzy Mamdani logic to match with player abilities. The following is an example of manual calculation fuzzy testing using sample input remaining player stats "HPP" of 35, "STimer" of 15, "nItemHP" is 6, "nItemTimer" is 7, "JObs" is 8, and "nSelamat" is 5. In this study, we will use multilevel fuzzy so that the first will calculate game stage 1, proceed to game stage 2, and then calculate stage levels. The calculation is as follows:

Stage 1

The first stage is calculating Mamdani fuzzy logic to find the value Safety using HPP, STimer parameters. The steps are:

A. Fuzzification

The first step is to calculate the membership degree of each input based on the membership function that has been created. In this case, the memberships degrees of HPP weak is $\mu(HPP_{HWeak}(35)) = 0.6$ and medium $\mu(HPP_{HMedium}(35)) = 0.4$. Thus, the memberships degrees of STimer moderate is $\mu(STimer_{SModerate}(15)) = 0.5$, and STimer high is $\mu(STimer_{SHiah}(15)) = 0.5$.

B. Inference

The inference process uses Min-Max operations with the table rules that have been made. There are two processes: the implication process using the MIN operator and the composition process using the MAX operator. There are several rules whose results of inference are 0, so there is no need to calculate them. The fulfillable regulations are 2, 3, 5, and 6 rules id. The detailed inference is described below:

- 1. IF HPP **WEAK** AND STIMER **MODERATE** THEN SAFETY **LOW** FLOW = min(0.6;0.5) = 0.5
- 2. IF HPP **WEAK** AND STIMER **HIGH** THEN SAFETY **FAIR** FFAIR = min(0.6;0.5) = 0.5
- 3. IF HPP **MEDIUM** AND STIMER **MODERATE** THEN SAFETY **FAIR** FFAIR = min(0.4;0.5) = 0.4
- 4. IF HPP **MEDIUM** AND STIMER **HIGH** THEN SAFETY **GOOD** FGOOD = min(0.4;0.5) = 0.4

After evaluating each inference, then it will receive the output of $\mu(Safety_{FLow}) = 0.5$, $\mu(Safety_{FGood}) = 0.4$, and $\mu(Safety_{FFair}) = max(0.5; 0.4) = 0.5$.

C. Defuzzification

In this process, calculations are carried out to change the output results, namely the value of the conclusions resulting from the inference engine to a crips value. The crips value z^* generated using the centroid method is as in Equation 25. The z_j is the function of specific (output) parameters concerning fuzzy sets *j*.

$$z^* = \frac{\sum_{j=1}^{n} z_j \mu(z_j)}{\sum_{j=1}^{n} \mu(z_j)}$$
(25)

After getting the inference results, then calculate the defuzzification using Equation 25, then the result $Z * = \frac{(((5+10+15)\times0.5)+((30+35+50)\times0.5)+(55+65+70)\times0.4)}{(0.5\times3)+(0.5\times3)+(0.4\times3)} = 35.36.$

Stage 2

The second stage is calculating Mamdani fuzzy logic to find the value Reflex using nItemHP, nItemTimer, and JObs parameters. The steps are:

A. Fuzzification

The first step is to calculate the membership degree of each input based on the membership function that has been created. In this case, the memberships degrees of nItemHP low is $\mu(\text{nItemHP}_{NLow}(6)) = 0.8$ and moderate $\mu(\text{nItemHP}_{NModerate}(6)) = 0.2$. The memberships degrees of nItemTimer slight is $\mu(\text{nItemTimer}_{NISlight}(7)) = 0.6$, and nItemTimer fair is $\mu(STimer_{NIFair}(7)) = 0.4$. The memberships degrees of jObs low is $\mu(\text{JObs}_{JLow}(8)) = 0.4$, and jObs moderate is $\mu(\text{JObs}_{IModerate}(8)) = 0.6$.

B. Inference

There are two processes: the implication process using the MIN operator and the composition process using the MAX operator. There are several rules whose results of inference are 0, so there is no need to calculate them. The fulfillable regulations are 1, 2, 4, 5, 10, 11, 13, and 14 rules. The detailed inference is described below:

- 1. IF nITEMHP LOW AND nITEMTIMER SLIGHT AND JOBS LOW THEN REFLEX FAST RFast = min(0.8;0.6;0.4) = 0.4
- IF nITEMHP LOW AND nITEMTIMER SLIGHT AND JOBS MODERATE THEN REFLEX FAST RFast = min(0.8;0.6;0.6) = 0.6
- 3. IF nITEMHP LOW AND nITEMTIMER FAIR AND JOBS LOW THEN REFLEX FAST RFast = min(0.8;0.4;0.4) = 0.4
- 4. IF nITEMHP LOW AND nITEMTIMER FAIR AND JOBS MODERATE THEN REFLEX NORMAL RNormal = min(0.8;0.4;0.6) = 0.4
- 5. IF nITEMHP MODERATE AND NITEMTIMER SLIGHT AND JOBS LOW THEN REFLEX FAST RFast = min(0.2;0.6;0.4) = 0.2
- 6. IF nITEMHP MODERATE AND nITEMTIMER SLIGHT AND JOBS MODERATE THEN REFLEX FAST RFast = min(0.2;0.6;0.6) = 0.2
- 7. IF nITEMHP MODERATE AND nITEMTIMER FAIR AND JOBS LOW THEN REFLEX FAST RFast = min(0.2;0.4;0.4) = 0.2
- IF nITEMHP MODERATE AND NITEMTIMER FAIR AND JOBS MODERATE THEN REFLEX NORMAL RNormal = min(0.2;0.4;0.6) = 0.2

After evaluating each inference, then it will receive the output of $\mu(Reflex_{RNORMAL}) = max(0.4; 0.2) = 0.4$ and $\mu(Reflex_{RFast}) = max(0.4; 0.6; 0.4; 0.2; 0.2; 0.2) = 0.6$.

C. Defuzzification

After getting the inference results, then calculate the defuzzification using Equation 25, $Z * = \frac{(((25+30+45)\times0.4)+(50+70+75)\times0.6)}{(0.4\times3)+(0.6\times3)} = 52.33.$

Stage Level

The last stage is calculating Mamdani fuzzy logic to find the value of stage level using Safety, Reflex, and nSelamat parameters. The steps are:

A. Fuzzification

The first step is to calculate the membership degree of each input based on the membership function that has been created. In this case, the memberships degrees of Safety low is $\mu(Safety_{FLow}(35.36)) = 0.23$ and fair $\mu(Safety_{FFair}(35.36)) = 0.77$. The memberships degrees of Reflex normal is $\mu(Reflex_{RNormal}(52.33)) = 0.38$, and fast is $\mu(Reflex_{RFast}(52.33)) = 0.61$. The memberships degrees of nSelamat are based on the number of rescued victims $\mu(nSelamat_{NSMedium}(5)) = 1$.

B. Inference

There are two processes: the implication process using the MIN operator and the composition process using the MAX operator. There are several rules whose results of inference are 0, so there is no need to calculate them. The fulfillable regulations are 5, 8, 14, and 17 rules id. The detailed inference is described below:

IF SAFETY LOW AND REFLEX NORMAL AND nSELAMAT MEDIUM THEN EASY Easy= min(0.23;0.38;1) = 0.23 IF SAFETY LOW AND REFLEX FAST AND nSELAMAT NSLOW THEN MEDIUM Medium = min(0.23;0.61;1) = 0.23 IF SAFETY FAIR AND REFLEX NORMAL AND nSELAMAT MEDIUM THEN MEDIUM Medium = min(0.77;0.38;1) = 0.38 4. IF SAFETY FAIR AND REFLEX FAST AND nSELAMAT MEDIUM THEN HARD Hard = Min(0.77;0.61;1) = 0.61 After evaluating each inference, then it will receive the output of $\mu(Level_{Easy}) = 0.23$, $\mu(Level_{medium}) = max(0.23; 0.38) = 0.38$, and $\mu(Level_{Hard}) = 0.61$.

C. Defuzzification After getting the inference results, then calculate the defuzzification using Equation 25, $Z * = \frac{(((9+12+17)\times0.23)+((14+16+19)\times0.38)+(19+21+24)\times0.61)}{(0.23\times3)+(0.38\times3)+(0.61\times3)} = 18.14.$

Based on the case example above with the condition that the player's remaining health points (HPP) are 35, the remaining timer (STimer) is 15, and the number of health point boosting items (nItemHP) taken is 6. The number of times increasing items (nItemTimer) taken is 7, and the obstacle that hits there are eight players (JObs). The number of victims saved (nSelamat) is 5. It results in a difficulty level of 18.14 which will later become table speed and obstacles.

Playtesting Experiment

Subsequent tests were carried out by conducting experiments on 20 players aged 8 to 43 years old. After the players have finished playing the game, each player will be evaluated by noting the parameters used for

the fuzzy logic system and the difficulty level they get. The results of the 20 players are shown in Table 4 for only the last best speed.

Table 4. Playtesting experiment										
No	Time	Remaining HP (%)	obstacle	Item HP	Item Timer	Num. Victims	safety	reflex	speed	Stage Room
	0	100	0	0	0	0	0	0	5	1
1	4	100	0	0	3	7	31,37	79,23	21,3	2
	3	44.13	7	2	12	11	21.61	61.12	22.55	3
•	0	100	0	0	0	0	0	0	7	1
2	8	80	2	0	0	11	68,33	80,77	24,97	2
	0	100	0	0	0	0	0	0	6	1
3	19	90	3	6	7	8	55,37	96,35	24,8	2
	0	100	0	0	0	0	0	0	5	1
4	12	80	2	0	0	2	74,95	70,59	21,53	2
~	0	100	0	0	0	0	0	0	5	1
5	9	80	2	0	0	11	24,64	56,8	21,29	2
~	0	100	0	0	0	0	0	0	5	1
6	18	100	0	0	0	3	73,76	68,91	22,82	2
7	0	100	0	0	0	0	0	0	6	1
/	17	90	1	0	0	4	65,26	86,94	21,2	2
0	0	100	0	0	0	0	0	0	5	1
ð	1	90	3	4	1	4	40,13	72,65	20,15	2
	0	100	0	0	0	0	0	0	7	1
9	14	100	0	0	0	1	65,06	65,82	18,39	2
	20	80	4	3	22	12	65,81	31,52	21,57	3
10	0	100	0	0	0	0	0	0	6	1
10	25	79,27	7	6	6	16	72,04	63,36	25,09	2
11	0	100	0	0	0	0	0	0	6	1
11	23	95,76	5	6	6	8	62,97	62,97	25,96	2
12	0	100	0	0	0	0	0	0	6	1
12	8	80	2	0	0	7	52,69	61,95	23,51	2
13	0	100	0	0	0	0	0	0	7	1
15	9	80	2	0	0	14	67,61	71,29	28,36	2
14	0	100	0	0	0	0	0	0	5	1
	16	100	0	0	0	7	80,34	51,88	24,6	2
15	0	100	0	0	0	0	0	0	5	1
	1	80	2	0	0	3	39,4	67,27	16,81	2
	19	46,09	7	3	14	13	61,99	37,27	21,75	3
16		100	0	0	0	0	0	0	7	1
	16	67,39	4	1	9	7	72,69	59,36	25,19	2
	18	30,93	20	20	23	24	40,21	13,17	14,75	3
17	0	100	0	0	0	0	0	0	5	1
	16	100	0	0	0	6	85,9	85,9	22,22	2
18	0	100	0	0	0	0	0	0	5	1
	14	60	4	0	0	6	60,84	74,6	21,71	2
10	23	50,8	10	9	19	20	71,57	15,17	22,36	3
19	0	100	0	0	0	0	0	0	5	1
•	9	80	2	0	0	16	71,42	48,38	25,02	2
20	0	100	0	0	0	0	0	0	1	1
	9	100	0	0	U	5	74,92	56,46	20,83	2

From the results of the tests carried out 20 times with the data shown in Table 4, it can be concluded that each player's difficulty level is almost different for each stage. Therefore, the difficulty level obtained becomes adaptive. It changes according to conditions or is adjusted based on the player's abilities, which can be seen from the value of each parameter of the fuzzy logic system obtained when players play games. Also, it is according to the rule table that has been made. In the test data table, the minimum difficulty level is five and the maximum difficulty level is 28.36 (indicated as speed).

CONCLUSION

A game to provide knowledge on earthquake mitigation has been successfully made under the name Quake Run. This game is an endless game equipped with an adaptive difficulty level feature. Using the Mamdani fuzzy logic system, the adaptive difficulty level was successfully applied to the table's movement and the obstacle falling speed. It is measured from the input parameters in the game. The result can be seen in the playtesting conducted by 20 players, where each stage has a different speed. In the playtesting, the minimum difficulty level obtained is five, and the maximum adaptive difficulty level obtained is 28.36.

REFERENCES

- B. Setyogroho, D. Muslim, M. S. Sadewo, G. O. Muslim, S. Burhanuddin, and H. Hendarmawan, "Correlation between Building Damages and Losses with the Microzonation Map of Mataram— Case Study: Lombok Earthquake 2018, Indonesia," *Sustainability*, vol. 14, no. 4, p. 2028, Feb. 2022, doi: 10.3390/su14042028.
- [2] T. Yanuarto, S. Pinuji, A. C. Utomo, and I. T. Satrio, *Buku Saku Tanggap Tangkas Tangguh Menghadapi Bencana*. Pusat Data Informasi dan Humas BNPB, 2019.
- [3] A. V. H. Simanjuntak and K. Ansari, "Seismicity clustering of sequence phenomena in the active tectonic system of backthrust Lombok preceding the sequence 2018 earthquakes," *Arabian Journal* of Geosciences, vol. 15, no. 23, p. 1730, Dec. 2022, doi: 10.1007/s12517-022-10973-y.
- H. Lei, X. Wang, H. Hou, L. Su, D. Yu, and H. Wang, "The earthquake in Jiuzhaigou County of Northern Sichuan, China on Aug. 8, 2017," *Natural Hazards*, vol. 90, no. 2, pp. 1021–1030, Jan. 2018, doi: 10.1007/s11069-017-3064-3.
- [5] G. P. Hayes, G. M. Smoczyk, A. H. Villaseñor, K. P. Furlong, and H. M. Benz, "Seismicity of the Earth 1900–2018 Scientific Investigations Map 3446," 2020. Accessed: Feb. 04, 2023. [Online]. Available: http://pubs.er.usgs.gov/publication/sim3446
- [6] Y. Zhang, J. F. Fung, K. J. Johnson, and S. Sattar, "Review of Seismic Risk Mitigation Policies in Earthquake-Prone Countries: Lessons for Earthquake Resilience in the United States," *Journal of Earthquake Engineering*, vol. 26, no. 12, pp. 6208–6235, Sep. 2022, doi: 10.1080/13632469.2021.1911889.
- [7] S. Mohadjer *et al.*, "Earthquake emergency education in Dushanbe, Tajikistan," *Journal of Geoscience Education*, vol. 58, no. 2, pp. 86–94, 2010, doi: 10.5408/1.3534854.
- [8] N. A. Belinda, Linawati, and K. O. Saputra, "UI/UX Design of Educational Game for Earthquake Mitigation," in 2020 IEEE International Women in Engineering (WIE) Conference on Electrical and Computer Engineering (WIECON-ECE), Dec. 2020, pp. 70–73. doi: 10.1109/WIECON-ECE52138.2020.9398019.
- [9] H. M. Ihsan, D. P. Putri, H. M. Insani, A. Redo, and A. S. Bratanegara, "Utilization of Fun Games as Earthquake Disaster Mitigation Efforts for Inclusive Children," *Tunas Geografi*, vol. 11, no. 2, p. 89, Dec. 2022, doi: 10.24114/tgeo.v11i2.39739.
- [10] B. Nirmala, A. Agusniatih, and H. Annuar, "Development of snakes and ladders game (disaster response) as earthquake mitigation for children," *Journal of Early Childhood Care and Education*, vol. 3, no. 2, p. 97, Jan. 2021, doi: 10.26555/jecce.v3i2.3111.
- [11] D. N. G. Botelho, "Treme-Treme 2.0-A serious game to teach children earthquake preparedness," Tecnico Lisboa, 2019.
- [12] Z. Feng *et al.*, "Towards a customizable immersive virtual reality serious game for earthquake emergency training," *Advanced Engineering Informatics*, vol. 46, p. 101134, Oct. 2020, doi: 10.1016/j.aei.2020.101134.
- [13] J. Yu, A. R. Denham, and E. Searight, "A systematic review of augmented reality game-based Learning in STEM education," *Educational technology research and development*, vol. 70, no. 4, pp. 1169–1194, Aug. 2022, doi: 10.1007/s11423-022-10122-y.
- [14] D. Vergara, Á. Antón-Sancho, and P. Fernández-Arias, "Player profiles for game-based applications in engineering education," *Computer Applications in Engineering Education*, vol. 31, no. 1, pp. 154–175, Jan. 2023, doi: 10.1002/cae.22576.
- [15] Y. Hu, T. Gallagher, P. Wouters, M. van der Schaaf, and L. Kester, "Game-based learning has good chemistry with chemistry education: A three-level meta-analysis," *J Res Sci Teach*, vol. 59, no. 9, pp. 1499–1543, Nov. 2022, doi: 10.1002/tea.21765.
- [16] M. Zohaib, "Dynamic Difficulty Adjustment (DDA) in Computer Games: A Review," Advances in Human-Computer Interaction, vol. 2018, pp. 1–12, Nov. 2018, doi: 10.1155/2018/5681652.

- [17] X. Fang, J. Zhang, and S. S. Chan, "Development of an Instrument for Studying Flow in Computer Game Play," *Int J Hum Comput Interact*, vol. 29, no. 7, pp. 456–470, Jul. 2013, doi: 10.1080/10447318.2012.715991.
- [18] N. P. H. Pratama, S. M. S. Nugroho, and E. M. Yuniarno, "Fuzzy controller based AI for dynamic difficulty adjustment for defense of the Ancient 2 (DotA2)," in 2016 International Seminar on Intelligent Technology and Its Applications (ISITIA), Jul. 2016, pp. 95–100. doi: 10.1109/ISITIA.2016.7828640.
- [19] T. Killedar, G. Suriya, P. Sharma, M. Rathor, and A. Gupta, "Fuzzy Logic for Video Game Engagement Analysis using Facial Emotion Recognition," in 2021 8th International Conference on Signal Processing and Integrated Networks (SPIN), Aug. 2021, pp. 481–485. doi: 10.1109/SPIN52536.2021.9566124.
- [20] A. Hubble, J. Moorin, and A. S. Khuman, "Artificial Intelligence in FPS Games: NPC Difficulty Effects on Gameplay," in *Fuzzy Logic, Cham: Springer International Publishing*, 2021, pp. 165– 190. doi: 10.1007/978-3-030-66474-9_11.
- [21] A. L. Sánchez and A. G. Lara, "A Serious Game prototype for strengthen Mathematical Logical Reasoning with implementation of Fuzzy Logic System," in *Proceedings of the 7th Mexican Conference on Human-Computer Interaction*, Oct. 2018, pp. 1–4. doi: 10.1145/3293578.3298780.