

# LANDSLIDE SUSCEPTIBILITY MAPPING USING FREQUENCY RATIO-BASED FUZZY LOGIC APPROACH IN LIMA PULUH KOTA REGENCY

Surtani<sup>1</sup>, \*Ahyuni<sup>1</sup>, Hamdi Nur<sup>2</sup>, Alifah Tahsyia<sup>1</sup>, Dina Rahayu Eliza Prisma<sup>1</sup>

<sup>1</sup>Department of Geography – Universitas Negeri Padang, Indonesia

<sup>2</sup> Department of Urban and Regional Planning, Universitas Bung Hatta, Indonesia

Email: ahyuniaziz@fis.unp.ac.id

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**ABSTRACT:** One of the non-linear models for determining landslide-prone areas is the fuzzy logic method. This research aims to determine the landslide prone zoning map in Limapuluh Kota Regency, by using fuzzy gamma operations in GIS, from 11 factors: aspect, curvature, elevation, slope, landform, geology, proximity to stream, landcover, rainfall intensity, NDVI and soil. The data used is 149 landslide events in Limapuluh Kota Regency. The data was processed using the factors that cause landslides with a train and test ratio of 60: 40. The research results show that by using  $\gamma = 0.975$ , an area with high landslide characteristics was obtained covering an area of 28,463 ha (8.79%), with a medium area of 76,544 ha (23.64%) and low area of 218,820 (65.57%). By validating using the ROC curve and test data, an AUC value of 0.746 was obtained, which means the accuracy level of the resulting map is high

**Keywords:** *Landslide, fuzzy gamma, susceptibility map*

## 1. INTRODUCTION

Limapuluh Kota Regency is an area prone to landslides in West Sumatera. This regency connects West Sumatera Province and Riau Province. It was recorded that in 2017, there were 10 cases of landslides, 22 cases in 2018, 11 cases in 2019, 3 cases in 2020, 2 cases in 2021, 8 cases in 2022 and at the end of 2023 there were landslides on the West Sumatra-Riau route [1] (BPBD, 2023). For this reason, it is necessary to estimate the locations prone to landslides in Limapuluh Kota Regency.

Various methods are used to determine landslide susceptibility maps. Starting from heuristic, statistical and deterministic approaches [1](Ahyuni, Susetyo, B.B, Oktari, F., Nur, H., Aziz, 2021). Other researchers put forward 2 techniques, namely direct hazard mapping and indirect hazard mapping [2](Tangestani, 2009). It is directly related to the geomorphologist's determination of his knowledge of slope conditions, while indirectly it uses statistical models which are determined based on the factors that cause landslides and landslides events.

The use of statistical methods has been widely used as stated by [3](Mandal, S., Mondal, 2019). The use of statistical models is related to the data used in an area. Preparation of the data to be used (training data and its quality depends on the nature of the field investigation used. However, the model used will determine the data requirements [4] The

model used can be in linear or non-linear form [5] [6] [7] Examples of comparisons of statistical methods have been carried out by many experts, such as [8] [9] [10] [11].

The use of non-linear models between influencing factors and landslide events can be done with fuzzy membership functions with the help of GIS and raster domains. This has been done by several experts such as [12] [13] [14] [13] [15] [16]. In this study, fuzzy logic was applied to determine the landslide susceptibility zoning map in Limapuluh Kota Regency, Indonesia by determining factors in the occurrence of landslides with values between 0 and 1. Fuzzy gamma was used to determine the landslide susceptibility map.

## 2. METHODS

Preparation for making disaster prone maps was carried out using Gamma operations on fuzzy membership with the help of GIS. The data used includes aspect, curvature, elevation, slope, landform, geology, proximity to stream, landcover, rainfall intensity, NDVI and soil.

One of the methods used in the fuzzy gamma operator is Frequency Ratio. The ratio frequency of a particular class in a causal parameter can be obtained when compared with a landslide inventory.

Frequency Ratio

### *Information:*

$N_{\text{cell}}(\text{Li})$  = the number of avalanche cells in a particular category

N cell (Ci) = the total number of avalanche cells in a particular category

N cell (L) = total number of avalanche cells

N cell (C) = number of cells

$$FR = \frac{X_{FR} - Min_{FR}}{Max_{FR} - Min_{FR}}$$

## **Max FR - Information**

$X_{FB}$       ≡    Frequency ratio value

$\Delta_{FR}$  = Frequency ratio value  
 $Min_{FR}$  = Minimum value of Frequency ratio

**Max<sub>FR</sub>** = Maximum value of Frequency ratio

Operator Fuzzy

### 2.1. Fuzzy Algebraic Product

Where.

$\mu_i$  = Fuzzy membership function for map  $i$  ( $i = 1, 2, \dots, n$  maps to be combined)

## 2.2. Fuzzy Algebraic Sum

Where,

$\mu_i$  = Fuzzy membership function for map  $i$  ( $i = 1, 2, \dots, n$  maps to be combined)

### 2.3. Fuzzy Gamma Operator

The fuzzy gamma operator is a fuzzy operator for combining membership functions. The fuzzy gamma operator formula, ie:

Where,

The exponent  $\gamma$ , which is a number from  $<0, 1>$  interval, allows optimization of the membership combination. Setting it to the extremes of the interval give either fuzzy algebraic sum ( $\gamma = 1$ ) or fuzzy algebraic product ( $\gamma = 0$ )

Analysis of landslide hazard levels using Fuzzy overlay. The tools provided by GIS for fuzzy overlay consist of 5 operators, And, Or, Sum, Product and Gamma. In this study, the Gamma tools were used. In Fuzzy Gamma it is the result of Fuzzy Sum multiplied by Fuzzy Product. The study area and landslide events can be seen in figures 2 and 3.

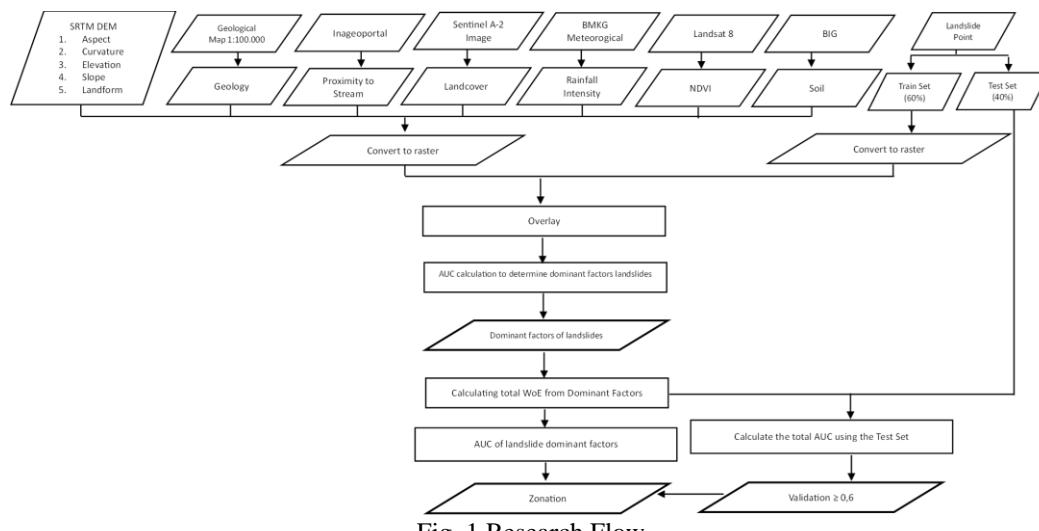


Fig. 1 Research Flow

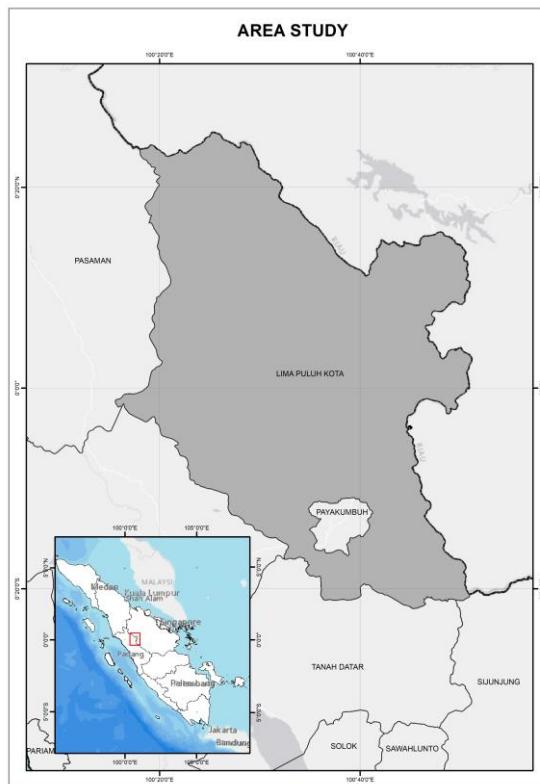


Fig. 2 Study Area

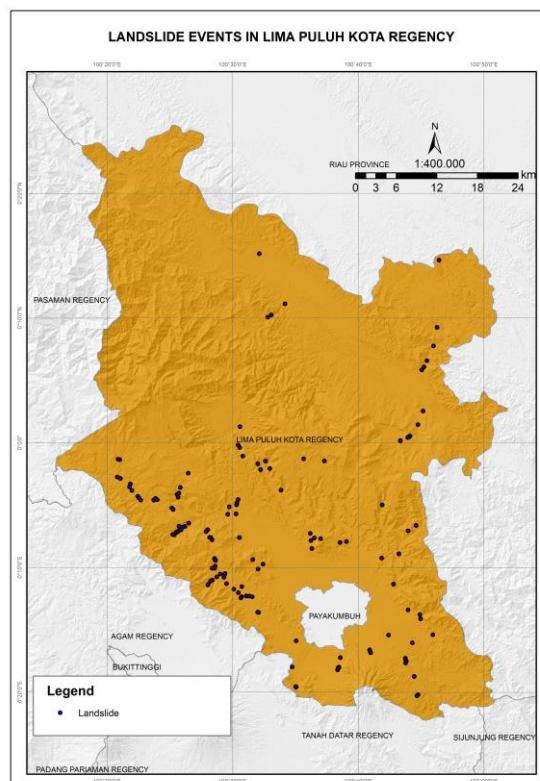


Fig. 3 Landslide Events in Lima Puluh Kota Regency

Source: Results of data analysis, 2023

### 3. RESULT AND DISCUSSION

The results of calculations based on the fuzzy logic method show the minimum and maximum values, namely 0 to 1, where the closer the value is to 0, the lower the resulting level of danger and vice versa. In the aspect factor, northeast and east are the classes with the highest number of landslides, each with 15 landslide points. In the rainfall intensity factor, 2000-2500 mm has the highest number of landslides with 77 landslide points. The curvature factor with the highest landslide occurrence is in the class -0.006 - 0.003, namely 25 landslide points. The elevation factor for the highest class of landslide occurrence is at an altitude of 581 – 728 m, namely 37 landslide points. In terms of geological factors, Miocene Sandstone is a type of rock where many landslides occur 21 points. The proximity to stream where the most landslides occur is in the 2000 – 3000 m class, namely 22 landslide points. Plantation land cover is the class with the most landslides, namely 34 points. For landforms, the Tectonic hills class is the highest, with 21 points. The highest NDVI class is 0.392 - 0.421, 21 landslide points. The slope factor that has a high incidence of landslides is in the class 18.453 - 24.180, that is, there are 37 landslide points. And for the last parameter soil, the highest incidence of landslides occurred in the Haplic Podzol class 42 points.

In the aspect factor, west and north are the classes with the lowest number of landslides with 5 landslide points each. In the rainfall intensity factor, the class that does not have landslides is the 3,500 –

4,000 mm class. The curvature factor that does not have landslide events is class -0.232 - -0.060. The elevation factors for classes that do not have landslides are at heights 1038 - 1210 m, class 1210 - 1417 m, class 1417 - 1701 m and class 1701 - 2269 m. In terms of geological factors, the classes that do not have landslides are Kuantan Formation, Rhyoandesite Volcanic, Carboniferous Carbon, Airbalam Formation, Amas Volcanic Formation, Bohorok Formation, Kota Alam Volcanic Formation, Pematang Formation, Telisa Formation, Telukkido Formation, Granite, Minor Granite Intrus, Miocene Polymictic and Mangani Porphyry. Proximity to streams that do not have landslides are in the 100 - 200 m class. Landcover factors that do not have landslides are lakes, fishponds, rivers and swamps. For the landform factor, classes that do not have landslides are Intermontane plateau, Old Volcanic Plain, Escarpment, Flow path, Volcanic foot, Upper slope, Peneplain, Karst hills, Tectonic Plate Ridge and Body of water. The NDVI factor that does not have landslides is 0.018 – 0.111. For slope factors that do not have landslides, they are class 34.997 - 40.405, class 40.405 - 46.768, class 46.768 - 55.04 and class 55.04 - 81.128. And for the last parameter soil, the classes that do not have landslides are Umbrik Andosol, Escarpment, Eutric Gleysol, Kandic Nitisol and Body of water. These results can be seen in more detail in table 1.

The results of the fuzzy sum and fuzzy product operations can be seen in the figures 4 and 5. By using  $\gamma = 0.975$ , results were obtained with an AUC value of 0.746 in model testing.

Table 1 Fuzzy Membership Calculation based on Frequency Ratio

No	Parameter	Class	Pixel (Ci)	Class of pixel (Ci) (%)	Landslide Pixel (Li)	Landslide Pixel (Li) (%)	Frequency Ratio	Fuzzy Membership (FR)
1	Aspect	North	344103	6,57	9	10,11	1,54	1,00
		Northeast	732151	13,98	15	16,85	1,21	0,69
		East	700050	13,37	15	16,85	1,26	0,74
		Southeast	647773	12,37	11	12,36	1,00	0,49
		South	615405	11,75	8	8,99	0,76	0,27
		Southwest	641977	12,26	7	7,87	0,64	0,16
		West	617756	11,80	5	5,62	0,48	0,00
		Northwest	614125	11,73	14	15,73	1,34	0,81
2	Rainfall Intensity	North	324001	6,19	5	5,62	0,91	0,41
		2000 – 2500	2655162	50,71	77	86,52	1,71	0,81
		2500 – 3000	2395306	45,75	6	6,74	0,15	0,07
		3000 – 3500	167286	3,19	6	6,74	2,11	1,00
3	Curvature	3500 – 4000	18304	0,35	0	0,00	0,00	0,00
		-0,232 - -0,060	7067	0,13	0	0,00	0,00	0,00
		-0,060 - -0,039	44136	0,84	1	1,12	1,33	0,85

No	Parameter	Class	Pixel (Ci)	Class of pixel (Ci) (%)	Landslide Pixel (Li)	Landslide Pixel (Li) (%)	Frequency Ratio	Fuzzy Membership (FR)
4	Elevasi	-0,039 - -0,025	159807	3,05	1	1,12	0,37	0,24
		-0,025 - -0,015	426780	8,15	10	11,24	1,38	0,88
		-0,015 - -0,006	874546	16,70	21	23,60	1,41	0,90
		-0,006 - 0,003	1616004	30,86	25	28,09	0,91	0,58
		0,003 - 0,011	1176621	22,47	13	14,61	0,65	0,42
		0,011 - 0,021	640131	12,22	13	14,61	1,20	0,77
		0,021 - 0,039	254563	4,86	4	4,49	0,92	0,59
		0,039 - 0,206	37686	0,72	1	1,12	1,56	1,00
		65 – 237	1025804	19,59	5	5,62	0,29	0,09
		237 – 409	761290	14,54	7	7,87	0,54	0,16
5	Geology	409 – 581	1003682	19,16	14	15,73	0,82	0,24
		581 – 728	649868	12,41	37	41,57	3,35	1,00
		728 – 883	613551	11,72	24	26,97	2,30	0,69
		883 – 1038	516381	9,86	2	2,25	0,23	0,07
		1038 – 1210	348748	6,66	0	0,00	0,00	0,00
		1210 – 1417	172873	3,30	0	0,00	0,00	0,00
		1417 – 1701	94259	1,80	0	0,00	0,00	0,00
		1701 – 2269	50814	0,97	0	0,00	0,00	0,00
		Alluvium	57272	1,09	1	1,12	1,03	0,05
		Andesite to Basalt	210834	4,03	8	8,99	2,23	0,11
6	Soil	Andesite of Gunung Malintang	257524	4,92	6	6,74	1,37	0,07
		Kuantan Formation	73274	1,40	0	0,00	0,00	0,00
		Rhyoandesite Volcanic	2032	0,04	0	0,00	0,00	0,00
		Carboniferous Carbon	13775	0,26	0	0,00	0,00	0,00
		Carboniferous Metamorphic	187797	3,59	7	7,87	2,19	0,11
		Miocene Sandstone	285968	5,46	21	23,60	4,32	0,21
		Airbalm Formation	64233	1,23	0	0,00	0,00	0,00
		Amas Volcanic Formation	26973	0,52	0	0,00	0,00	0,00
		Bohorok Formation	15257	0,29	0	0,00	0,00	0,00
		Brani Formation	217904	4,16	8	8,99	2,16	0,11
		Kota Alam Volcanic Formaton	32630	0,62	0	0,00	0,00	0,00
		Kuantan Formation	1848717	35,31	20	22,47	0,64	0,03
		Pematang Formation	58581	1,12	0	0,00	0,00	0,00
		Sangkarewang Formation	184117	3,52	2	2,25	0,64	0,03
		Sihapas Formation	833812	15,93	1	1,12	0,07	0,00
		Telisa Formation	462199	8,83	0	0,00	0,00	0,00
		Telukkido Formation	36214	0,69	0	0,00	0,00	0,00
		Totolan Formation	283049	5,41	10	11,24	2,08	0,10
		Granite	55057	1,05	0	0,00	0,00	0,00
		Minor Granite Intrus	3752	0,07	0	0,00	0,00	0,00
		Miocene Polymictic	163	0,00	0	0,00	0,00	0,00
		Mangani Porphyry	9838	0,19	0	0,00	0,00	0,00
		Andesite or Dacite Porfiry	14514	0,28	5	5,62	20,27	1,00

No	Parameter	Class	Pixel (Ci)	Class of pixel (Ci) (%)	Landslide Pixel (Li)	Landslide Pixel (Li) (%)	Frequency Ratio	Fuzzy Membership (FR)
6	Proximity To Stream	0 – 100	151304	2,89	8	8,99	3,11	1,00
		100 – 200	119709	2,29	0	0,00	0,00	0,00
		200 – 300	113412	2,17	1	1,12	0,52	0,17
		300 – 400	109305	2,09	1	1,12	0,54	0,17
		400 – 500	106594	2,04	3	3,37	1,66	0,53
		500 – 1000	511765	9,77	13	14,61	1,49	0,48
		1000 – 2000	908070	17,34	19	21,35	1,23	0,40
		2000 – 3000	757523	14,46	22	24,72	1,71	0,55
		3000 – 4000	629997	12,03	4	4,49	0,37	0,12
		>5000	1829499	34,93	18	20,22	0,58	0,19
7	Landcover	Lake	13781	0,26	0	0,00	0,00	0,00
		Fishpond	157	0,00	0	0,00	0,00	0,00
		Forest	1853101	35,40	6	6,74	0,19	0,09
		Shrubs	377507	7,21	8	8,99	1,25	0,57
		Moor	972372	18,57	24	26,97	1,45	0,66
		River	19627	0,37	0	0,00	0,00	0,00
		Settlement	73971	1,41	2	2,25	1,59	0,72
		Paddyfield	400244	7,64	15	16,85	2,20	1,00
		Plantation	1522509	29,08	34	38,20	1,31	0,60
		Swamp	2192	0,04	0	0,00	0,00	0,00
8	Landform	Alluvial plain	189062	3,61	1	1,12	0,31	0,03
		Intermontane plateau	28169	0,54	0	0,00	0,00	0,00
		Kolovial plain	23372	0,45	2	2,25	5,03	0,48
		Tectonic plain	443270	8,47	3	3,37	0,40	0,04
		Volcanic plain	179450	3,43	11	12,36	3,61	0,34
		Old Volcanic Plain	65252	1,25	0	0,00	0,00	0,00
		Escarpmment	8139	0,16	0	0,00	0,00	0,00
		Tectonic Escarpment Plateau	67950	1,30	1	1,12	0,87	0,08
		Volcanic intrusion	16814	0,32	3	3,37	10,50	1,00
		Flow path	1526	0,03	0	0,00	0,00	0,00
		Volcanic foot	30222	0,58	0	0,00	0,00	0,00
		Upper slope	58020	1,11	0	0,00	0,00	0,00
		Lower slope	97873	1,87	6	6,74	3,61	0,34
		Middle slope	77625	1,48	2	2,25	1,52	0,14
		Tectonic mountains	2134005	40,76	16	17,98	0,44	0,04
		Old Volcanic Mountains	465136	8,88	14	15,73	1,77	0,17
		Peneplain	28327	0,54	0	0,00	0,00	0,00
		Karst hills	46365	0,89	0	0,00	0,00	0,00
		Tectonic hills	1083309	20,69	21	23,60	1,14	0,11
		Old Volcanic Hills	88538	1,69	9	10,11	5,98	0,57
		Tectonic Plate Ridge	92892	1,77	0	0,00	0,00	0,00
		Body of water	10153	0,19	0	0,00	0,00	0,00
9	NDVI	0,018 - 0,111	58704	1,12	0	0,00	0,00	0,00

No	Parameter	Class	Pixel (Ci)	Class of pixel (Ci) (%)	Landslide Pixel (Li)	Landslide Pixel (Li) (%)	Frequency Ratio	Fuzzy Membership (FR)
10	Slope	0,111 - 0,183	77777	1,48	2	2,25	1,51	0,78
		0,183 - 0,235	157980	3,02	5	5,62	1,86	0,96
		0,235 - 0,278	272953	5,21	9	10,11	1,94	1,00
		0,278 - 0,314	479636	9,16	13	14,61	1,60	0,82
		0,314 - 0,342	697317	13,31	13	14,61	1,10	0,57
		0,342 - 0,368	986361	18,83	14	15,73	0,84	0,43
		0,368 - 0,392	1098358	20,97	6	6,74	0,32	0,17
		0,392 - 0,421	987348	18,85	21	23,60	1,25	0,65
		0,421 - 0,628	421847	8,05	6	6,74	0,84	0,43
		0 - 6,363	1025804	19,59	5	5,62	0,29	0,09
11	Soil	6,363 - 12,408	761290	14,54	7	7,87	0,54	0,16
		12,408 - 18,453	1003682	19,16	14	15,73	0,82	0,24
		18,453 - 24,180	649868	12,41	37	41,57	3,35	1,00
		24,180 - 29,588	613551	11,72	24	26,97	2,30	0,69
		29,588 - 34,997	516381	9,86	2	2,25	0,23	0,07
		34,997 - 40,405	348748	6,66	0	0,00	0,00	0,00
		40,405 - 46,768	172873	3,30	0	0,00	0,00	0,00
		46,768 - 55,04	94259	1,80	0	0,00	0,00	0,00
		55,04 - 81,128	50814	0,97	0	0,00	0,00	0,00
		Umbrik Andosol	34098	0,65	0	0,00	0,00	0,00
		Escarment	8132	0,16	0	0,00	0,00	0,00
12	Aspect	District Gleysol	257040	4,91	4	4,49	0,92	0,33
		Eutric Gleysol	44398	0,85	0	0,00	0,00	0,00
		District Cambisol	292858	5,59	2	2,25	0,40	0,15
		Lithic Cambisol	59060	1,13	2	2,25	1,99	0,72
		Oxic Cambisol	234474	4,48	7	7,87	1,76	0,64
		Gleyic Luvisol	128073	2,45	6	6,74	2,76	1,00
		Oxic Latosol	101574	1,94	2	2,25	1,16	0,42
		Kandic Nitisol	15064	0,29	0	0,00	0,00	0,00
		Haplic Podzol	2119343	40,48	42	47,19	1,17	0,42
		Kandic Podzol	1863237	35,59	23	25,84	0,73	0,26
13	Elevation	ROC	67944	1,30	1	1,12	0,87	0,31
		Body of water	10167	0,19	0	0,00	0,00	0,00

Source : Data Processing, 2023

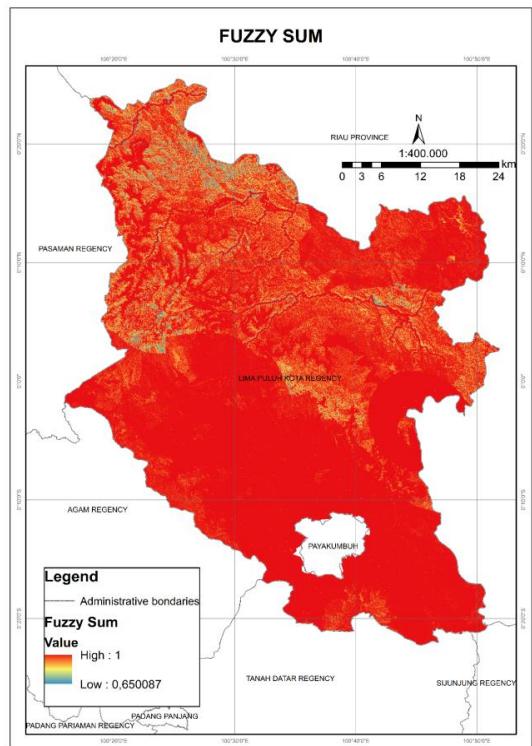


Fig. 4 Fuzzy Sum

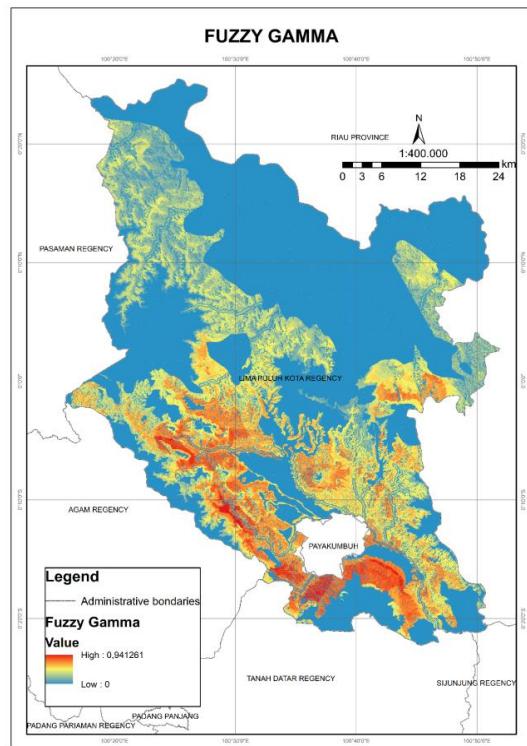


Fig. 6 Fuzzy Gamma

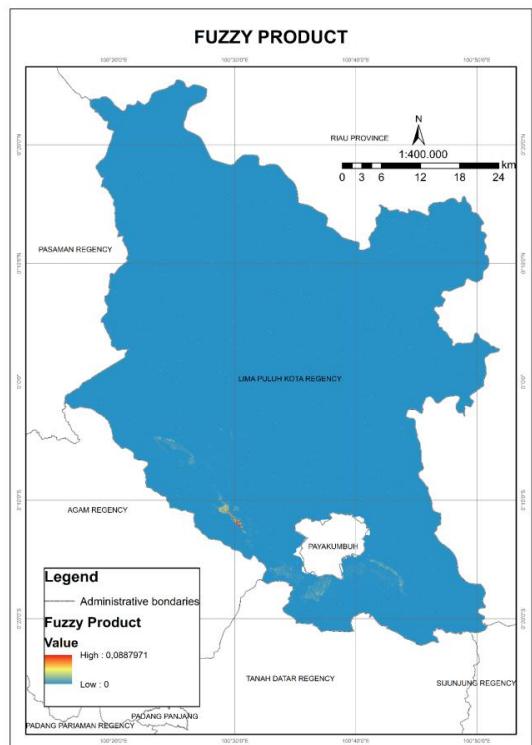


Fig. 5 Fuzzy Product

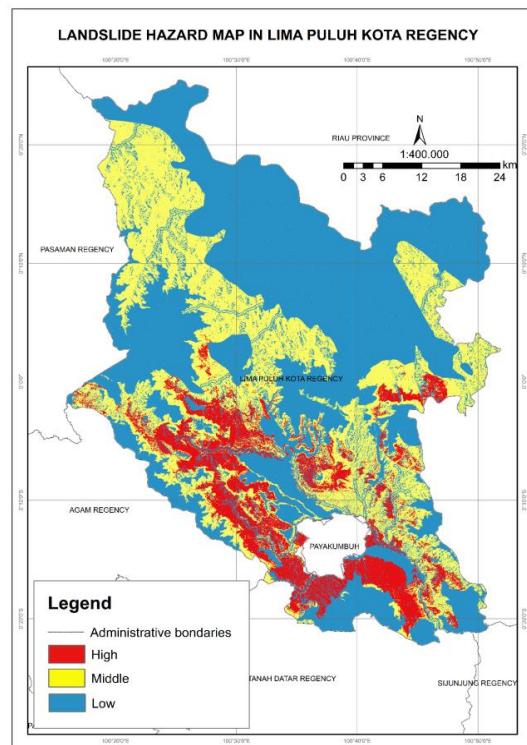


Fig. 7 Landslide Hazard Map Lima Puluh Kota Regency

Table 2 Landslide Hazard Results using the Fuzzy Gamma Operator

No	Class	Area (ha)	Percent (%)
1	High	28.463	8,79
2	Middle	76.544	23,64
3	Low	218.820	67,57

Source: data processing, 2023

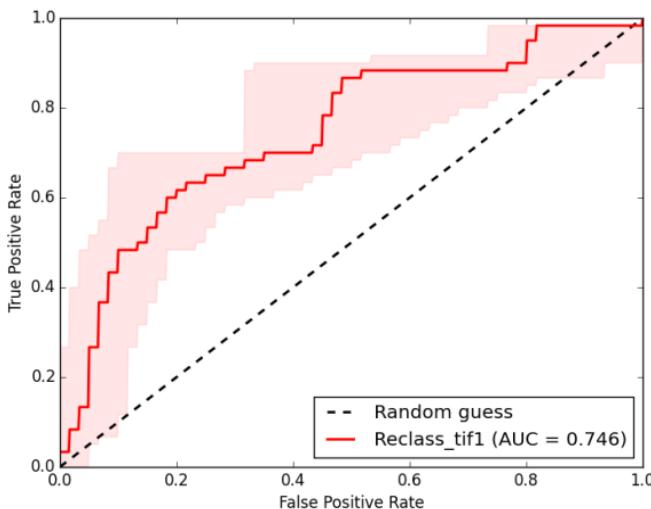


Fig. 8 ROC and AUC

#### 4. CONCLUSION

The landslide hazard map based on the Fuzzy Gamma operator based on the Frequency Ratio membership function has varying levels of danger. The susceptibility map results show that the most suitable one with  $\gamma = 0.975$  is an AUC value of 0.746, which means the level of accuracy of the resulting map is high. The model produces high landslide susceptibility classes covering an area of 28,463 ha (8.79%), medium landslide susceptibility covering an area of 76,544 ha (23.64%) and low landslide susceptibility covering an area of 218,820 ha (67.57%). Thus, awareness of the danger of landslides needs to be paid attention to in the western and central regions of Limapuluh Kota Regency because that is where the concentration of landslide vulnerability is high and medium.

#### 5. ACKNOWLEDGEMENTS

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