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Natural Hazards

Semeru volcano, Indonesia: Measuring hazard, exposure and response of densely populated neighbourhoods facing persistent volcanic threats

--Manuscript Draft--

ABSTRACT

26 We studied Semeru, **East Java** to show the population exposure to volcanic threats from its persistent, daily eruptive activity which endangers at least 50,000 of the 950,000 inhabitants 28 living on the **East and SE slopes and ring plain.** Surveys, mapping and statistical investigation enabled us to assess the extent of exposure of 145 neighbourhoods (termed blocks) and characterize hazards and response to eruptions in 15 rural villages and small towns. Statistical analyses of datasets of 23 variables (11 of exposure, 7 of hazards, and 5 of response) and their attributes involved three operations: 1. Univariate and bivariate analyses enabled us to explore data and characterize the relationships between 11 variables to compute a multi-component exposure index. 2. Polytomous Logistic Regression (PLR) models selected six optimal exposure variables, suggesting that logistic regression can predict the exposure index for blocks outside the survey area and potentially on any active volcano. 3. Multivariate analyses and Hierarchical Agglomerative Clustering (HAC) distinguished four groups of blocks based on attributes of all variables correlated to the exposure index score. To contribute to disaster risk reduction, the distance/time criterion was applied to access ways and response facilities to highlight remote or blocked blocks in danger of imminent eruption including evacuation. Statistical analysis of optimal variables from local scale surveys can help identify neighbourhoods where disaster risk mitigation requires improvement. **Key-words**: human exposure; statistics; neighbourhood, volcano; Semeru; Indonesia.

1. Introduction

1.1. Study background and terminology

 The present study is part of the 'Local Adaptation to Volcanic Risk' research project, and seeks to understand how and why dense rural communities continuously exposed to persistent volcanic threats can thrive on Semeru's slopes and ring plain. Another paper explains how Semeru's communities adjust to, compensate for, and tolerate continuous exposure to persistent volcanic threats (Thouret et al., 2022). Here, based on field surveys, mapping and statistical analyses, we seek to determine the population exposure to persistent volcanic threats by means of a composite index computed at the scale of neighbourhoods (termed blocks) in rural villages and small towns (Fig. 1, ESD Table 1). Using a second group of

 variables on volcanic hazards and a third group on access and response, enabled us to rank and map a range of remote blocks in the case of imminent eruptions.

In this study, we use the international terminology on Disaster Risk Reduction (UNDRR,

58 2017; UNISDR, 2017) to which the reader is referred. Risk is the product of probability \times

losses, where the probability is a function of hazard and the losses depend on both exposure

and external and internal vulnerability (Aspinall and Blong, 2015). Volcanic risk analysis has

61 focused on cost-benefit evaluation (e.g., Woo , 2015) in particular the case of evacuations

preceding imminent eruptions (Jumadi et al., 2018, Lechner and Rouleau, 2019). The concept

- of risk also includes hazard knowledge, risk perception and resilience among communities
- living on active volcanoes (Gaillard, 2008; Gaillard and Dibben, 2008; Paton et al., 2008;

Lavigne et al. 2008; Donovan, 2010; Donovan et al., 2018). A broader risk concept stems

from the appraisal of value systems and beliefs, governance systems and decisions, and

political economies (Bakkour et al., 2015).

Exposure and vulnerability are two principal components of risk. Exposure is the location and

spatial distribution of people, buildings, property, infrastructure, networks, lifelines,

production capacities and other tangible human assets within reach of a hazard event

(UNDRR, 2017). In the context of disaster risk reduction, Wisner et al. (2004) defined

vulnerability as "the characteristics of a person or group to anticipate, cope with, resist or

recover from the impact of a natural hazard". Coping capacity is part of the widely used term

of resilience, which is the ability of a community exposed to hazards to resist, absorb,

accommodate to and recover from the effects of a hazard in a timely and efficient manner

(UNISDR, 2008, 2017). Here we focus on exposure, a concept that has often been included in

- the vulnerability field; but exposure means being potentially affected by volcanic activity in a
- peculiar place, whereas vulnerability is the inability to withstand the effects of a harmful

process. As stated by the First IAVCEI-GVM workshop (2018), "exposure and vulnerability

may be related: exposure quantifies number of people and/or buildings in the area, while

vulnerability is one of the characteristics of the exposed elements that may suffer hazard

impacts".

Studies measuring population exposure to volcanic hazards (Ewert, 2007; Kinvig et al., 2010;

Brown et al., 2015a; Nieto-Torres et al., 2021) are relatively few, compared with works on

- physical exposure of buildings and critical infrastructure (e.g., Lerner-Lam, 2007; Jenkins et
- al., 2014, 2015; Wilson et al. 2014; Jimenez et al., 2019) and profuse investigations on
- 87 vulnerability and risk perception (e.g., Jóhannesdóttir and Gísladóttir, 2010; Zuccaro et al.,
- 2015; Thouret et al., 2014a; Michellier et al., 2020). Yet, exposure, having both spatial and
- temporal patterns, is a fundamental component of risk analysis, and together with
- vulnerability, underpins loss. Exposure compels threatened people to adjust to risk through
- resilience to avoid the dangers of persistent volcanic activity.

1.2. Global and other approaches to exposure

 We use local-scale surveys and statistical methods for defining exposure around Semeru. The main alternative is the Global approach. Exposure of people to volcanic risk has been the focus of the 'priority countries of the Global Facility for Disaster Reduction and Recovery' (GFDRR) report by Aspinall et al. (2011), This pilot study on volcanic risk in 31 countries presented a method for "measuring the Volcano Population Index (the number of people threatened by each volcano: Ewert and Harpel, 2004) combined with the hazard level of each volcano to quantify population risk". The Population Exposure Index (PEI: Aspinall et al., 2011, later summarized by Brown et al. (2015a) is one of the main indices used in assessing 101 volcanic risk in highly populated areas together with Volcanic Hazard Index (VHI, Brown et al., 2015b), Human Development Index (HDI, UNDP, 2020) and Social Vulnerability Index (SoVI, The Hazards and Vulnerability Research Institute, University of Southern Carolina). VHI is an index rating the hazard level of a volcano based on the recurrence of past eruptions, the average and maximum Volcanic Explosivity Index ratings, and the extent of pyroclastic, mud, or lava flows in the eruptions. The Human Development Index (UNDP, 2020) measures levels of social and economic development based on "four criteria: mean and expected years of schooling, [life expectancy](https://www.investopedia.com/terms/l/lifeexpectancy.asp) at birth, and [gross](https://www.investopedia.com/terms/g/gross-national-income-gni.asp) national income per capita". The Social Vulnerability Index to hazards is based on "many socioeconomic, demographic, and built environment variables at the country or district level" (HVRI, University of South Carolina).

- 111 The GFDRR report (2011) estimated the numbers of people living within 10 km, 30 km and 100 km of each volcano, which was weighted according to evidence on historical distributions
- of fatalities within a given distance from the vent. Population numbers were weighted,
- 114 summed, and assigned to one of seven PEI scores in populated areas (Brown et al., 2015b).
- The authors estimated population risk for each volcano by "taking the product of the Hazard
- Level and PEI, and the numerical product was assigned to one of three Population Risk
- Levels". Following this approach, Semeru volcano, one of the prominent Indonesia's active
- volcanoes within the GFDRR category A, was assigned a high PEI level 3 with hazard level 3

 and a low uncertainty level 1. The method relies on a number of parameters having high uncertainty or that are unknown, which restricts its application to volcanoes with good

historical records.

 A physical exposure component was introduced in the USGS re-assessment of US dangerous 123 volcanoes. The updated Volcanic Threat Assessment (e.g., Ewert et al., 2018; Mangan et al., 2018) combines 15 hazard factors and 9 exposure factors that describe an individual volcano's hazard potential and the exposure of people and property to hazards. The exposure factors include population within 30 kilometers of the volcano; visitor numbers if it is located in a national park or monument; population beyond 30 kilometers if a far-traveling lahar is a primary hazard; prior eruption fatalities; prior evacuations; aviation impacts; impacts on power and transportation infrastructure; and major developments such as parks. A physical exposure component is also part of Del Negro et al (2020) quantification of lava flow risk on the flanks of Mt. Etna volcano using a GIS-based approach that integrates exposure of elements at stake within the hazard. The total exposure results from a weighted linear combination of four thematic layers, population, buildings, service networks, and land use, the weights of which were calculated using the Analytic Hierarchy Process. Population exposure relied on the density of population for each community normalized by the value obtained for the more populated municipality. Wood and Soulard (2009) proposed a composite exposure index for communities living in lahar-prone hazard zones around Mt. Rainier. They estimate the exposure index on the amount and percentage of six variables: developed land, residents, employees, public venues, dependent-population facilities, and parcel value, to estimate the spatial variations in community exposure.

141 More recently, Nieto-Torres et al (2021) quantified a multi-component volcanic risk index, which encompasses as many as 41 parameters: 9 parameters for hazard, 9 for exposure, 10 for vulnerability and 13 for resilience. However, the exposure criteria contain only one single human parameter, i.e., the density of population within 5, 10, 30 and 100 km radii from the main crater.

1.3. Semeru volcano, setting and rationale for targeting continuous exposure

Semeru, East Java, is Indonesia's highest and the southernmost volcano of the Tengger

massif, which includes the Tengger caldera with the active Bromo cone (Fig. 2). The proximal

(10 km), medial (30 km) and distal (100 km) areas exposed to the effects of volcanic hazards

- rank Semeru as one of the world's most dangerous volcanoes according to the GFDRR report.
- It hosts one of the most exposed populations worldwide, as the 100-km circle distance
- encompasses Java's second densest populated area (see Freire et al., 2019, their Fig. 9),
- including 10 million people of the metropolitan area of Surabaya, Indonesia's second
- economic centre and second international airport. The province of East Java with an area of c.
- 155 $48,000 \text{ km}^2$ is home to about 40 million people, making it one of the most densely populated,
- 156 largely rural areas on Earth with 830 people per $km²$ on average. East Java presents typical
- 157 low- to middle-income population at the lower range of the country's **HDI** (0.59-0.71; BPS)
- 2017) and mean SoVI score (Siagan et al., 2013). Java hosts 58% of Indonesia's population
- 159 over $\langle 7\%$ of the country size, i.e., 141 million (2015) among which 10 % are exposed to
- 160 disaster risk within 30 km from 22 historically active volcanoes.

 Semeru is a real concern to civil authorities owing to the combination of daily explosive activity and a dense population: at least 950,000 people live within a radius of 35 km from the volcano summit (Thouret et al., 2022). Lahar-related disasters caused >10,000 casualties during the 20th century alone. The 1909 catastrophic event ranks fourth among the ten deadliest eruptions in the world between 1900 and 2010, as the 1909 PDCs and lahars killed at least 5500 people, i.e., 5.5% of all fatalities of the ten world's deadliest eruptions since 1900 (Doocy, 2013). These events created havoc in thriving urban centres located in the 168 eastern ring plain, in particular **Pasirian, Tempeh**, and Lumajang, harboring 123,000 people 169 (as of 2015). Semeru's summit separates the Regencies of Lumajang to the East (1791 km², 170 1,036,000 people in 2015) and Malang to the West $(3531 \text{ km}^2, 2,547,000 \text{ people})$ (Fig. 2; BPS, 2017). Here we focus on the regency of Lumajang where 621,000 people live within 35 172 km of the summit, and the SW flank where 317,000 people represent a quarter of Malang's 173 regency.

- The study area hosts a range of areas exposed to deadly or disastrous pyroclastic density
- currents (PDCs), tephra fallout on proximal slopes, lahars along valleys as far as 35 km, and a
- variety of exposed assets, such as crops, road network, lifelines, factories, trade centres,
- religious monuments and touristic facilities in the ring plain. We defined and computed
- exposure of population living in 15 *dusun* of 6 districts (*Kecamatan*) with c. 800
- 179 inhabitants/ km^2 on average, although the size of rural settlements is highly variable between
- hamlets (350 people) and small, mixed rural/urban towns c. 7,500 people (ESD Table 1). The
- outlook of many of the towns is still rural as the growth of urban centres (Pronojiwo, Senduro,
- Candipuro) has not been accompanied by a parallel growth of [industry.](https://www.britannica.com/technology/industry) Urban centres benefit
- 183 from diversity, as they offer a variety of jobs and business such as agricultural trade, wood
- 184 industries, traditional crafts, and local tourism. Towns around Semeru mirror other Javan
- 185 centres in displaying great socio-economic [diversity,](https://www.merriam-webster.com/dictionary/diversity) which underlies a three-tiered social

186 [hierarchy.](https://www.merriam-webster.com/dictionary/hierarchy) Villagers are ethnically Javanese with a significant Madurese minority. Agriculture

- 187 and forestry are the main activities occupying 50% of the land and 40% of the workforce.
- 188 Trade routes and small manufacturing are growing in towns, but roads and railways remain
- 189 poorly developed.

190 **1.4. Semeru's persistent eruptive activity and extensive hazard-prone areas**

- 191 Eruptive activity, which has been recorded since 1884, is usually mild (VEI 1-2), but
- 192 increases every 8 to 11 years (Thouret et al., 2007; Solikhin et al., 2012). The constant activity
- 193 includes four eruptive styles:

194 1. The persistent vulcanian and sometimes strombolian regime consists of short-lived, ash-

195 laden < 0.5 km-high columns several times a day. Small columns usually disperse ash about 4

196 km around the summit, but all villages can be affected by ash dispersed from 4-6 km-high 197 columns as far as 8 to 12 km mostly to East, **SE** and **NE**, while exceptional seasonal winds

198 can blow fine ash as far as $20 \text{ km} \times 20 \text{ km} \times 10^8 \text{ m/s}$ (Fig. 3).

- 199 2. Increased vulcanian/pelean activity every 8 to 11 years produces several kilometer-high 200 eruption columns, ballistic bombs and thick tephra fall around the vent, and dispersed ashfall 201 40 km downwind. Collapses from crater- and dome-fed, steep lava flows produce block-and-202 ash flows that travel toward the SE as far as 12 km from the summit, e.g., 2002 and 2020-21.
- 203 3. Flank 'aa' lava flows 5 km long erupted on the lower SE and E flanks in 1895 and in 1941–
- 204 1942, while the crater termed Jenggring-Seloko (Fig. 3) has regularly produced stubby lava
- 205 flows reaching 1 to 3.3 km along the steep scar open to the SSE over the past 40 years.
- 206 4. Highly explosive eruptions are not unknown. Nakada et al. (2019) refer to deposits from
- 207 relatively large $(\geq$ VEI 3) eruptions from the 3rd to the 11th Centuries as well as thick scoria
- 208 falls and PDCs that destroyed the temple of the Majapahit Kingdom (13-16th Century) at
- 209 Candi Jawar (5.5 km SW of the present summit), around the 16th century. The East and South
- 210 flanks of Semeru together with the East and **SE** ring plain are the most prone to volcanic
- 211 hazards where 50,000 to 100,000 people are exposed to the effects of eruptive activity.
- 212 The hazard-zone map $(Fig. 3)$ displays three areas prone to volcanic threats as follows
- (Thouret et al., 2007, 2014b):

214 1) The high and steep summit cone within a circle of 5 km radius is affected by tephra-fall, 215 lava flows and PDCs on the SE flank on daily to annual basis. No one lives there, but scores of tourists frequently climb and visit the summit area (5 victims were reported in 2000). New tephra fallout covers the summit cone on a daily basis, while dome-collapse rock avalanches 218 and PDCs propagate annually as far as 4 km through the scar open to the SE , but they do not

- reach villages during the usual volcano "unrest" level 2 on a scale of 4.
- 220 2) The extensive South, SE and ESE flanks affected by PDCs as far as 12 km and lahars as far

as 9 to 18 km from the vent, along which more than 50,000 people now live within 0.5 km of

the active rivers (e.g., K. Koboan, K. Bang and K. Kembar, Fig. 3). On 4 December 2021,

223 both margins of the Koboan valley were affected by exceptionally long runout PDCs and hot

lahars flowing 16 km down valley and causing more than 50 fatalities (GVP, 2021).

225 3) The principal valleys draining the South (K. Glidik, K. Bang and K. Kembar), $SE(K)$.

Koboan), and East slopes (K. Tengah or Besuk Sat) convey many annual lahars across the

- ring plain at least 35 km down valley to the Indian Ocean (Thouret et al., 2014b). Post-
- eruptive lahars are the most frequent hazardous flows that propagate every rainy season along
- 229 the principal valleys to the South, SE and East. Voluminous lahars, exceeding 5 million m^3 ,

230 killed hundreds of people at least five times since 1884; they swept the SE and E ring plain in

231 1976, 1981, and devastated the city of Lumajang 35 km East in 1909 (Fig. 1).

- In sum, the most exposed population live along the valleys within 8 to 25 km from the
- summit. Recorded fatalities due to lahars and PDCs are higher along the principal valley
- reaches between 9 and 12 km distance (75%). Fatalities due to lahars occurred between 12

and 25 km (22%), the remainder (3%) being located along the > 25 km distal valleys in 1909,

1976 and 1981 (Thouret et al., 2007).

1.5. Location choice and local scale of observation: *Dusun* **and neighbourhood**

We studied dense, mostly rural communities located in the exposed areas between 8 and 20

km from the summit of Semeru noting past eruptions and reported fatalities (Fig. 3; Thouret et

al., 2007, their Table 1). The primary unit of study was neigbourhood (block) in sub-villages

- (termed *dusun*) located on the valley margins, terraces and interfluves within 0.5 km from the
- active channel that used to be affected by lahars and PDCs in the recent past. Locations
- 243 included the South (K. Kali= river) Bang, K. Kembar, e.g., 2002-2003), SSE (K. Koboan and
- 244 C. Lengkong, 1994-1995, 2020-2021), East (K. Tengah/ Besuk Smut, 1981), and the **SE** and
- East ring plain (K. Rejali, Mujur) towards the city of Lumajang, affected in 1909 (Fig. 3). We
- also studied villages farther away from active rivers and the town of Senduro outside the
- 247 affected valleys to avoid any bias that would stem from the exposed settlements in affected
- hazard zones (Thouret et al., 2022). The remaining *dusun* within the reach of light ashfall
- were selected for the purpose of comparison as they extend beyond active valleys (e.g.,
- Pasrujambe, Fig. 2) and farther away from the volcano. We examined small rural
- communities at higher altitude on the west flank that are exposed to frequent tephra-fall
- within 9 km of the crater (Fig. 2). We also considered mixed rural/urban communities in two
- towns: Pronojiwo in the vicinity of the active K. Bang and Kembar valleys, and Senduro
- farther away (17 km) from the volcano and any active river.
- We selected 15 *dusun* that belong to six *Desa* on the SW, S and E slopes of Semeru inside
- four *Kecamatan* in the Regency of Lumajang and one in Malang Regency (Table 1, Figs. 2-3; ESD Fig. 1) based on the following rationale:
- 1. Both *dusun* Blubuk and Karangsuko (*Desa* Tamansatryan**)**, located between 900 and 1250 259 m asl on the WSW flank 8-9 km from the vent, are exposed to frequent, light tephra-fall from the volcano. Both mid-altitude hamlets exemplify chronic exposure of a population living 261 close to the crater.
- 2. The small town of Pronojiwo (*Desa* and *Kecamatan* Pronojiwo**)** is the largest settlement (c. 7,600 people) on the South flank of Semeru between 600 and 700 m asl and at a 11-12 km distance of the summit. The rural suburb to the East of the town near two large rivers (K. 265 Bang and K. Kembar) has regularly been affected by PDCs and annual lahars. On the **ENE** side of Pronojiwo, the *dusun* Supit and Supit Timur 9 to 11 km from the vent and between 720 to 820 m asl are most exposed to volcanic threats (PDC, lahars, and tephra-fall) along the West margin of the K. Bang, with a few casualties due to overbank PDCs reported in Supit Timur in 2002. *Dusun* Rowobaung is located on a volcaniclastic fan between two active rivers at 11-12 km from the summit. These *dusun* host exposed dwellers living in relatively high magnitude and frequency PDC- and lahar-prone zones.
- 3. On the SE flank of Semeru, a group of four *dusun* belong to *Desa* Supit Urang. Oro-oro Ombo, away from active valleys, is less exposed than Sumbersari and Gumuk Mas, sitting on 274 a high terrace to the South of the K. Koboan 9 to 11 km from the summit. Curah Koboan is
- situated 9-10 km from the summit on a high terrace of Curah Lengkong, a tributary to K.
- Koboan. The population of these *dusun* live in high magnitude / frequency PDC-prone areas,
- with fatalities in 1994, 1995, 2021, and paid a heavy toll in 2020-2021.

4. Kajar Kuning (*Desa* Sumberwuluh; *Kecamatan* Candipuro) is located 0.5 km North of

Curah Koboan at the foot of hills on which the Volcano Observatory sits at Gunung Sawur.

We chose this hamlet to observe how the location away from the Lengkong River would

influence the response of less exposed blocks.

5. Both *dusun* Jabon and Tulungrejo (*Desa* and *Kecamatan* Pasrujambe**)** are located 10-13 km

away on the Semeru East slope on the North side of the K. Tengah valley. The *dusun*

Tulungrejo spreads out on the low and middle terraces of the active river, while Jabon is less

exposed on ridges 0.8 km north of K. Tengah. We targeted these *dusun* because the Tengah

- 286 valley was the site of many PDC- and lahar-related fatalities of the 1981 VEI > 3 eruption.
- 6. Farther East on the ring plain, three *dusun* Sumbermulyo, Juranglangak and Rekesan (*Desa* and *Kecamatan* Senduro**)** are located in the SE area of the town of Senduro (population c. 7,500) 18-19 km from the summit. Senduro, only 20 km away from *Kabupaten* Lumajang, is also a touristic hub, hosting historic Hindu temples, and a gateway to the Semeru-Tengger massif. These *dusun* host a mixed urban/rural population well away from the volcano and active rivers, hence much less exposed, in a low-frequency fallout-prone zone.

2. Methods

2.1. Data acquisition and survey procedure

295 The **SE, East and NE** slopes of Semeru were the target of two field campaigns in September 2018 and 2019. Four researchers from the IPB University, Bogor, the University Gadjah Mada, Yogyakarta (Indonesia), and the Laboratoire Magmas et Volcans, Université Clermont- Auvergne (France) conducted field observations on the population and home exposure to the impacts from PDCs, lahars and tephra fall. A Trimble TC 1000 GIS mapper device, Google Earth maps and topographic maps (scale 1:25,000 as of 2000) have been used in QGIS to locate the surveyed sites and collate structural observations on edifices (homes, offices, schools, health centres, mosques, and markets: Fig. 4; ESD Tables 1, 2). Using a satellite 303 image SPOT5 (pixel: 2.5 m) as of 2014 and Google Earth maps, we outlined the boundaries of *dusun* and built or un-built blocks in the vicinity of principal valleys that convey lahars or PDCs.

- The campaigns provided a representative number of observations, as we collected data from
- 279 households and we mapped 145 blocks in 13 rural *dusun* and 2 small towns. A *dusun*
- usually entails 4 to 5 *RukunWarga* (RW, a set of 50-75 buildings, mostly homes with a few
- offices, mosques or schools), which in turn includes 3 to 5 *RukunTetanga* (RT, 20-25 homes).
- 310 We collected data from 2 to 4 households per RT , the smallest administrative unit at which
- 311 respondents can be identified. A *dusun* has an average area of 2.07 km² and density of 806
- inhabitants/km², while a built block area is 0.28 km² with a density of 1,397 inhabitants/km²
- on average (ESD Table 1). We categorized the quality of construction and roof, and building
- orientation, as these parameters play a role on dwellers' exposure to tephra fall and volcanic
- flows (ESD Table 2). To complete the 2017 BPS census at the smallest scale possible, we also
- collected data on population, economic activity (mostly agriculture, husbandry and
- agroforestry) and emergency facilities from eight governmental offices in: *Desa* (sub-district)
- Tamansatryan (WSW flank), *Desa* and *Kecamatan* (district) Pronojiwo (South), *Desa* Oro-oro
- Ombo and Supit Urang (SSE), *Kecamatan* Candipuro and *Desa* Sumberwuluh (SE), *Desa*
- Pasrujambe, and *Desa* Senduro (East) (Figs. 2, 3; ESD Table 1).
- Few limitations encumbered the field survey on multi-component exposure of people and
- dwellings, which was conducted in Indonesian and/or Javanese thanks to our partners. We
- interviewed respondents available in the household, either male or female, through a random
- 324 door-to-door approach and with the interpreter's help. One limitation may be
- representativeness. Among 950,000 people around Semeru, 50,000 live along active rivers,
- and up to 100,000 people in areas where eruptive activity exceeds VEI 3. Surveys involving
- about 279 households and 300 buildings represent approximately 0.6% to 2.6% of the
- 328 respective cohorts. However, mapping the exposure index involved 145 blocks, hence 6% of
- 329 the exposed population.

2.2. Statistical analyses

- Statistical analyses explored a set of 23 variables collected through surveys and organized in
- three fields: exposure, hazard, and access/response (Fig. 1, Tables 2, 3). The first group of 11
- variables and their 43 attributes helped compute the population exposure index of 145 blocks
- (see 2.2.1 below). The second group of seven variables determined the hazard level of blocks
- both inside and outside the hazard-prone zones around Semeru (Fig. 3; Thouret et al., 2007).
- The third group of five variables defined the access to blocks and response facilities for
- imminent evacuation.

 Table 1 summarizes the statistical methods and techniques together with their purposes for coding, quantifying and validating three groups of variables and their attributes. Statistical investigation conducted with the R software has involved four operations.

 First, univariate and bivariate analyses were used to extract relevant attributes from variable observations and detect their relationships.

 Second, Polytomous Logistic Regression models helped select optimal exposure variables and predict the exposure index of populated neighbourhoods (EIPN). Logistic regression is a technique commonly applied to highlight dependences between one variable that must be explained (dependent, endogenous, here the Exposure Index of a block) and several variables that explain relationships (independent, exogenous, quantitative and qualitative, ordinal or nominal). Logistic regression allows a model to be developed while the selection between models is done though a set of criteria such as reduction of the Akaike Information Criterion (AIC; Akaike, 1987) and confusion matrix. AIC was designed as the divergence between the true model (that actually generated the data) and a proposed statistic approximation of this model. Polytomous logistical regression (PLR) was performed using Cumulative Links Models to quantify the strength of association between each active variable and the response

variable (EIPN score) to be explained.

Third, multiple correspondence analyses (MCA) of all variables cross-referenced with the

EIPN score were used to characterize the blocks. Hierarchical Ascending Classification

(HAC) was used to discriminate clusters amongst blocks.

 Fourth, we used the distance/timing criterion combined with the third group of access and response variables to rank blocks with respect to relief operations for evacuation.

2.2.1. Methods to compute the population exposure index in neighbourhoods

 To compute the Exposure Index of Populated Neighbourhoods, we used five techniques (Fig. 1, Table 1):

1. A normalization technique was used to avoid biases arising from the fact that all qualitative

- variables do not have the same number of attributes. For each attribute, x being the initial
- value, we subtract the minimum value and we divide by the maximum interval, as follows:

 $x_{normalized} = \frac{x - x_{min}}{x - x}$ 366 $x_{normalized} = \frac{x - x_{min}}{x_{max} - x_{min}} \in [0,1]$ Thus, all values of qualitative variables are equidistributed

between 0 and 1.

- 368 2. A Chi² test was performed to determine whether the variables are dependent or independent (ESD Table 3). In other words, to find out whether the difference between observed and expected data is random or due to a relationship between the variables under study.
- 3. The non-parametric Wilcoxon test [\(Wilcoxon,](https://fr.wikipedia.org/wiki/Frank_Wilcoxon) 1945) was used because some pairs of
- 372 variables were identified as independent according to the Chi² test. This test allowed us to test
- whether two variables show similar distributions.
- 374 Based on results of the Chi² and Wilcoxon tests, we determined that all variables were
- dependent and followed similar distributions. The sum of the normalized values of 11
- variables and their 43 attributes for each of the 145 blocks provided the EIPN score.
- 4. The Shapiro-Wilks test (Shapiro and Wilks, 1965) can be used to determine whether a particular dataset follows a normal distribution, which is a common assumption used in many statistical tests. The Shapiro-Wilks test is based on the null hypothesis according to which the EIPN score follows a normal distribution. We discarded the null hypothesis when the p-value 381 was lower that the α risk fixed to 0.5 (5%).
- 5. Discretization was based on Jenks optimization method or natural breaks classification method (Jenks, 1967) because the distribution of values does not follow a normal distribution and the number of attributes varies between variables. The EIPN score parametric distance between classes is calculated using the Jenks technique. As it minimizes the intraclass variance while maximizing the interclass variance, we fix the number of breaks, i.e., number of discretized classes. The Jenks-type discretization, as opposed to other methods, allowed us to fit data to the shape of the statistical distribution, while it provided four homogenous classes.

2.2.2. Statistical analyses to characterize block exposure, hazards, and access/response

- 1. Univariate analysis (UA, Table 1) is the first step for exploring and preparing a dataset for further analysis. UA summarizes descriptive statistics and provides graphical representations for their univariate distribution (e.g., Chambers et al., 2018).
- 2. Bivariate analysis (BA, Table 1) involves the analysis of two variables with the aim to test
- simple hypotheses of association or any relationship between two variables and attributes. The
- Chi-square test aims to compare observed results with the expected results (see the test
- purpose in section 2.2.1 above; ESD Table 3). Factorial Correspondence Analysis (FCA) is
- used to study links between two qualitative variables and explore correlations and oppositions of categorical variables in a table of frequencies or contingency (ESD Table 4). From these
- correlations and oppositions, FCA thus allows us to state hypotheses to identify typical blocks in the Semeru *dusun*.

3. Multiple Correspondence Analysis (MCA) is run when a set of observations includes

multiple qualitative variables. MCA is a data analysis technique used to detect and represent

underlying structures in a large categorical dataset (Abdi and Valentin, 2007). This technique

represents data as points, illustrated by biplot graphs. MCA consists of the following steps

(Table 1; ESD Fig. 3):

(1) Eigenvalues helped define the number of retained dimensions having the highest inertia

rate, hence the most relevant information from 81 attributes of 23 variables. Benzécri

correction (Benzécri, 1979) applied to the number of dimensions (inertia) helps estimate how

much information is included in each dimension and select the smallest number of dimensions

411 which contain the maximum information available (ESD Fig. 2). We retained the first three

dimensions, which contained almost 91% of the information.

(2) Biplots show the relationships between attributes of all variables, while taking into

 account the quality of representation quantified by square cosines, contribution and coordinates of each attribute.

(3) Scatter plots establish relationships between the attributes among the same qualitative

variable. Confidence ellipses highlight proximities (closeness) and oppositions between the

attributes of one variable in the projection of blocks along a given axis in scatterplots.

MCA procedure includes two tests: Fisher test to highlight the correlation between a given

variable and the factorial axis and Student test to verify which attributes have singular

coordinates along the axis (dimension). ESD Table 5 displays the coordinates, squared

cosines and contributions of the attributes with respect to dimensions. These contributions

help select the contributory variables retaining more information and their attributes. Squared

cosines show the quality of representation and help select variables with sufficient quality

within the most contributory ones. Coordinates of this table indicate the position of attributes,

the most distant from the barycentre of each biplot contributing most.

427 4. Hierarchical Agglomerative Clustering (HAC; see Table 1) was performed on the MCA

outputs to confirm previous MCA groups and/or construct groups of blocks sharing similar

statistical characteristics. HAC leads to a factor map identifying four groups of blocks (ESD

Fig. 4) based on results from frequencies of attributes.

3. Statistical results

The statistical analysis using UA, BA and MCA techniques of 11 exposure variables and their

attributes (Table 5, ESD Table 3) aimed to predict the variations in the index of population

exposure between blocks.

3.1. Defining variables for the multi-component exposure index

 To compute the EIPN at the block scale, we used 11 variables and their attributes: 9 variables increase the exposure to volcanic threats, to which we added two defining the exposure of

dwellings (Fig. 1; Table 2; ESD Table 3). Each variable shows attributes divided by

440 thresholds, amounting to 43 attributes (Table 2). A number from 0 to 5 is attributed to each

attribute, although variable V8 exceptionally has 10 attributes (Table 2). One variable V4 only

was attributed a weighting coefficient (see the method involving four steps: Table 3). The

rationale for each variable and its attributes relies on risk-related exposure works (e.g., Ewert,

2007; Ewert et al., 2018; Loughlin et al., 2015; Nieto-Torres et al., 2021), as follows:

V1: Three circle radii around the vent (Fig. 2) are: a maximum 9 km distance at which

ballistics and heavy (lapilli-sized) tephra-fall occur and may impact roofs commonly made of

tiles supported by light timber; a maximum 12 km distance at which the longest PDCs

reached settlements in recent history (e.g., in 2002 with VEI 3), and an average 18 km

distance for lahars, exceptionally 35 km to the sea for the most voluminous ones (e.g., in

450 2021). The energy line concept (Malin and Sheridan, 1982) yields maximum runout distances

between 12 and 18 km for Semeru's PDCs while the 30-35 km range is commonly reported

for voluminous lahars (e.g., Ewert et al., 2005). Each of the circle radii were assigned the

following values based on the distribution of *dusun*: 9-12 km= 3, as 71% of *dusun* are located

 and most of prior fatalities occurred within this range; 12 to 18 km= 2 (22% of *dusun*); and < 9 km= 1 (7% of *dusun*).

V2: We retrieved population density at the *dusun* scale from the 2017 census (BPS, 2017).

V3: Population density within the built blocks was estimated from house counting (4 persons

on average per household) on Google maps and field observations with an error range of c.

10%. We ranked densities in increasing order from the lowest to the highest threshold (using

quartile interval) to reflect the increasing effect on people exposure in both *dusun* and blocks

(Table 2).

- **V4**: Table 3 shows the four steps of the methods used to compute the coefficients for V4. The
- inferred number of fatalities from the recorded eruption dataset (Table 2 *in* Thouret et al.,
- 2007), total 1,115 victims from 18 events reported since 1884. As about 950,000 people live
- 465 in the 35 km-distance circle with an area of ca. 1380 km^2 around the volcano, the ratio yields
- c. 1175 victims per million people, c. 8.5 victims per year, and c. 62 fatalities per event on
- average. These figures are likely under-estimated, while numbers of injured or displaced
- people are possibly large.
- 469 Using the hazard zones affected by the historical eruptive activity and lahars (Fig. 3) and the
- number and approximate location of fatalities, we chose the circle distances at 9, 12, 18 and
- 35 km. Population numbers in villages per circle area are 2,864 with radius distances of 9 km,
- 59,236 with 12 km; 85,303 with 18 km, and 842,597 people with 35 km radius, based on our
- calculation of the number of houses in the circles 1 to 3, and on the BPS 2017 census beyond
- 18 km. The weighted population (fatalities and distances) according to Brown et al.'s (2015a)
- method (Table 3) is 103 in circle of 9 km radius; 4,858 in circle of 12 km; 15,739 in circle of
- 18 km; and 587,800 in circle of 35 km radius. As 1,115 fatalities are reported for a population
- of 26,432 living in the 15 surveyed *dusun* based on 9 km, 12 and 18 km circle distances (Fig.
- 2), the final weighting coefficient for **V4** is 0.036 for the 9-km circle, 0.082 for the 12-km
- circle, 0.185 for the 18-km circle, and 0.698 for the 35-km circle (Table 3).
- **V5**: Location of fatalities with respect to valleys that transmit flows across three sites with
- values ranked from 3 to 0 according to hazard-type effects: 3 is given to valley margins
- reached by pyroclastic surges and overbank lahar flows; 2 is assigned to valley channels
- conveying confined pyroclastic flows and lahars, and terraces or interfluves swept by ash-
- cloud surges up to 500 m from the river; 0 indicates 'no reported death' along valleys (Table
- 2).
- **V6**: Lahar fatalities is a binary variable (Yes 1/No 0) and both figures 1,0 have been
- distributed along the valley channels, terraces and margins.
- **V7**: PDC fatalities is also binary (Yes 1/No 0), and spatially distributed, valley-confined flows
- being distinct from un-confined surges on valley margins and adjacent interfluves.
- **V8**: Location of mapped blocks on terraces is based on a range of distances from, and
- elevations above the river channel. Ten cases based on both criteria involved distance
- 492 thresholds from 15-50 m to > 120 m, and elevation thresholds from < 2 m to > 30 m. We
- attributed marks 10 to 1 based on decreasing exposure to lahars and PDCs, from the
- maximum 10 for channels and the minimum 1 for high and distant terraces.
- **V9**: Timely location of people living and/or working on Semeru's flanks encompass three situations in decreasing exposure duration order: permanent, daily work and home at night;
- temporary, home at night only, and; temporary, daily work only and night elsewhere.
- To the 9 variables that define human exposure we added two variables that usually are criteria
- defining the physical vulnerability of buildings. Here we consider them within the exposure
- field, as they determine the extent to which residents are exposed to volcanic threats owing to
- tephra fallout load on roofs and pressure impacts from PDCs and lahars on homes.
- **V10**: Orientation of house, i.e., perpendicular, oblique or parallel to PDC and lahars.

 V11: The quality and type of roof and construction are ranked as poor (wood), regular (zinc) with light timber, and fair when the roof is concrete and/or covered by tiles supported by sturdy timber. Almost all roofs, being pitched and many being covered by tiles or zinc, could probably withstand about 20 cm-thick tephra-fall, but concomitant, frequent rainfall would diminish the resistance threshold, in particular for roofs of light timber.

3.2. Computing the exposure index of populated neighbourhoods (EIPN)

509 The table in Figure 5 and ESD Table 6 display the EIPN score (from 1.17 to 8.56, median 4.17) obtained for each block after discretization. ESD Table 6 shows columns as variables and their thresholded attributes, while rows are the 145 surveyed *dusun* blocks. As a result, four colour-coded EIPN scores levels, from 'residual' (beige) to 'very high' (red), have been 513 attributed to each of the 145 blocks $(Fig. 5 A-F)$. The relatively narrow range of EIPN (standard deviation c. 1.05) reflects a relatively low exposure of all *dusun* to the potential effects of the Semeru activity except those which are located inside the proximal hazard zones and along valley channels depicted in Figure 3. The EIPN score displays three groups of blocks (Figs. 4): c. 39% of blocks show a moderate to high EIPN, and almost 25%, located in the vicinity of valley channels, show a very high EIPN. On the other end, almost 37% of the blocks show a very low or residual EIPN score, which correspond to older neighbourhoods that used to settle away from the active valleys. At this stage, we need to explore the relationships between the attributes of exposure variables to highlight the most contributory ones.

3.2.1. Relationships between attributes of exposure variables

 MCA scatter plots represent attributes using elliptic envelopes of confidence around the barycentre of groups (Fig. 6). Ellipses of 36 attributes that belong to the 11 exposure variables were plotted in the dimensions 1 and 2 scatter plots (Fig. 6). A number of opposed attributes (i.e. away from the barycentre) lead to the following inferences:

 1) The 9-12 km circle distance stands out among other distances from the vent, coinciding with the majority of exposed *dusun* within this circle. 2) In V2 and V3 scatter plots as well as in the V9 plot (timely location), the '0 density' stands out, which identifies temporary people location and un-built blocks. 3) The attributes valley channel and valley margins, although they are less distinct from other 'fatalities locations', coincide with river channels and low- altitude terraces that are more exposed to lahars and PDCs. 4) In V6 and V7 scatter plots, the presence/absence of prior fatalities due to PDCs and lahars also suggest the extent to which blocks are exposed. 5) The V8 plot shows that the relationship distance/elevation to the

- channel is not discriminant enough, except for the river channel. 6) The V10 and V11 scatter
- plots also suggest that building parameters (quality of roof, construction, and house
- orientation) may be significant to dwellers' exposure.
- As a result, the following eight contributory attributes of nine exposure variables may
- characterize the block exposure: the 9-12 km circle distance (V1), the *dusun* and block density
- (V2, V3) and temporary location (V9), the channel and low terrace location for prior fatalities
- (V8), the presence of PDCs and lahars (V6, V7), and both building parameters (V10, V11).
- For the sake of research efficiency, we reduced the number of variables to be collected in the
- field for predicting the EIPN.

3.3. Predictive capacity of Polytomous Logistic Regression to obtain EIPN scores

- Polytomous Logistic Regression (PLR) is one of the logistic regressions adapted to the study
- of categorical and/or ordinal variables with more than two attributes (Kleinbaum and Klein,
- 2010). As the EIPN score, variable of interest, is qualitative, PLR is the most adequate type of
- regression, involving a succession of techniques (Table 1):

3.3.1. Selection of a small number of active variables, and model significativity

- The initial PLR model included 11 variables and their 43 attributes (Table 4A). In fact, the
- selection procedure reduced 11 variables down to 8 to avoid collinearity (a linear relationship
- between two explanatory variables); for example, V1 distance to vent and V4 'Brown's PEI'
- parameter are dependent and collinear (i.e., retaining the same information and same
- distribution) with V2 and V3 (population density), hence V4 was not retained (Table 4A). We
- performed backward and forward model selection procedures, which consisted in sequentially
- adding or removing variables and investigating the resulting model performances. During
- backward or forward selection processes, AIC offers a compromise between parsimony and
- error reduction by penalizing a high number of parameters. The model having the smallest
- AIC possible was chosen to select optimal variables.
- Table 4A shows the initial PLR model with estimated coefficients of the selected attributes. 'Significativity', indicating that the underlying model coefficients are significantly different from 0, is the most important criterion for prediction. Based on the significativity, '3 stars' are attributed to attributes with < 0.1% error in the model, '2 stars' to attributes with < 1% error 565 and '1 star' to those with $<$ 5% error (Table 4B). According to the significativity results, a rigorous model would only include eleven 3-stars attributes with high probability. The selected PLR model (Table 4B) then reduced the non-collinear 8 variables to 6 only: V1 distance to vent; V2 density of *dusun*, V3 density of blocks, V5 Location of prior fatalities, V6 Lahar fatalities, and V7 PDC fatalities. The model also reduces the 43 attributes down to 15, as shown in Table 4B.
- The selected model comprises both positive and negative coefficients (Table 4B). Positive coefficients mean that when attributes increase, the EIPN score increases as well, while negative coefficients diminish the EIPN score. Thus, the attributes of the following variables: V1 distance to vent; V2 density of *dusun*, V3 density of blocks and V5 Location of prior fatalities will increase the chance that the blocks score a high EIPN, i.e., are more exposed. Conversely, blocks that possess low thresholds of both variables, V6 Lahar fatalities, and V7 PDC fatalities, will have a weak EIPN score: both hazards threatening these blocks are frequent, but they do not imply a large number of fatalities.

3.3.2. Probability of obtaining a block exposure index and quality of the model

PLR allows us to compute the probability of getting EIPN for any given block. We first

assessed the quality of the PLR model (Table 4): the calculated MSE is 0.179, while the

efficient ranking rate is as high as 82.07%. We used a Table termed confusion matrix (ESD

Table 7), which crosses 'real' observed EIPN scores with predicted ones, to test the quality of

- the estimated model. For a 'perfect' model, the confusion matrix would show a diagonal
- matrix in which non-diagonal entries are all equal to zero. The model is efficient if the
- number of non-diagonal, predicted values is below 40% of the total number of observed
- values (Kleinbaum and Klein, 2010). Among 145 observed blocks, we cannot predict 26,
- yielding a relatively low error rate of c. 18%, which ensures relative efficiency to the model.
- Two blocks located in the town of Pronojiwo and in the *dusun* Rowobaung (Fig. 5A) are given in the table below as examples based on specific attributes of six variables: R3, away from both
- 591 active rivers (K. Kembar, K. Bang), has been attributed a residual EIPN score, whereas P1,
- 592 located near the K. Bang valley channel, has been attributed a very high EIPN score in maps of
- 593 exposed blocks (Fig. 5A). We compute the EIPN probability of both blocks in order to test how
- 594 well PLR predicts the score:

595

596 Using the function 'predict', the probabilistic results for P1 block are as follows:

EIPN score prediction for block P1

597

598 Thus, the P1 block has 99.64% probability of obtaining a very high EIPN score. The result

599 shows that the selected model is working well, as we calculated a very high EIPN score for P1

600 (Fig. 5A). Using the same function, the probabilistic results for the R3 block yields:

 $E = \frac{1}{2}$

601 Thus, the R3 block has 82.41% probability of obtaining a residual EIPN score. The result 602 shows again that the selected model is valid, as we calculated a residual EIPN score for R3 603 (Fig. 5A).

604 **4. Mapping the composite exposure index at the scale of** *dusun* **neighbourhoods**

605 Maps display levels of the colour-coded EIPN score of 145 blocks (Fig. 5 A-F, ESD Table 6).

606 The most exposed, active valleys convey lahars and PDCs, in particular the large valley

607 channels (K. Koboan to the SE, K. Bang and Kembar to the S, and K. Tengah to the E). These

608 wide channels host no permanent homes, but only temporary shelters and small makeshift

609 shops, located on low terraces close to the river. Workers extracting material are highly

610 exposed to lahar and PDC impacts.

611 1. Most exposed *dusun* blocks to the current eruptive activity of Semeru are located along the 612 principal valleys that convey lahars and/or that have been affected by PDCs over the past 140

- years. These are Supit Timur near K. Bang, North and West blocks of Rowobaung between K.
- Bang and K. Kembar (Fig. 5A), affected in 2002. Highly exposed are the North blocks of
- Sumbersari and Curah Koboan on both margins of K. Koboan (heavily damaged in December
- 2020 and January 2021), and on the North bank of Curah Lengkong (Fig. 5B). The blocks of
- Tulungrejo (Pasrujambe) on the low and middle banks of K. Tengah are also highly exposed
- (Fig. 5C), while dykes built in the years 1970-1980 along the North bank of K. Tengah may
- provide a false sense of security.
- 2. Increasing eruptive activity producing the VEI 3 events every 8 to 11 years. At such times
- additional blocks would suffer heavy tephra-fall, lahars, and ash-cloud surges from PDCs
- along river valleys. Such *dusun* blocks encompass Supit along K. Bang, the East blocks of
- Pronojiwo near the bridge across K. Bang, most of Rowobaung between K. Bang and K.
- Kembar (Fig. 5A), most of Sumbersari and part of Gumuk Mas and Curah Koboan near K.
- Koboan (Fig. 5B), and most of Tulungrejo in the Desa Pasrujambe (Fig. 5C).
- 3. Other *dusun* blocks located down valley on low terraces can be affected by large lahars,
- e.g., K. Koboan down valley as far as Sumberwuluh (ESD Fig. 1D), as happened in January
- 2021, or by ash-cloud surges as far as Tawonsongo upstream of K. Tengah (Fig. 5C). Heavy
- 629 tephra-fall can affect the *dusun* of Blubuk and Karangusko in large VEI \geq 3 eruptions (Fig.
- 5D).
- 4. Blocks with very low to residual EIPN characterize the town of Senduro 17-19 km away
- from Semeru vent and away from active rivers (Fig. 5E). Senduro's dwellers are the least
- 633 exposed, except for tephra-fall dispersed from Semeru in large (VEI \geq 4) but un-frequent
- eruptions, and from Bromo (Tengger caldera) in 2009-2011, and 2020.
- 5. Contrasting EIPN scores of blocks situated in the same *dusun* are due to spatial changes
- and specific attributes at a local scale, such as density, hazard frequency, home quality and
- orientation, and dirt roads instead of paved roads (e.g., Supit and Pronojiwo, Fig. 5A).
- Decreasing EIPN to moderate level across *dusun* is due to decreasing densities and increasing
- elevation and/or distance from the channel (e.g., North and East blocks of *dusun* Curah
- Koboan and Sumbersari down valley, Fig. 5B, upper area of Tulungrejo and Tawonsongo,
- Fig. 5C). Different EIPN scores are also due to the distance to health centres and locations
- with limited lahar-related fatalities (e.g., Blubuk and Karangsuko, Fig. 5D).
- EIPN maps reflect the 2018-2019 situation when Semeru eruptive activity was mild (VEI 2)
- and the alert status was 2 on a scale of 4. As Semeru activity escalated to VEI 3 in December
- 2020 and January 2021, triggering dome-collapse PDCs and tephra-fall from > 6 km-high

 columns, together with hot lahars (GVP, 2022), every EIPN score would increase to the next higher colour-coded level (see EIPN Table in Fig. 5A-C).

5. Characterizing blocks (hazard, access/response) based on statistical analyses

- We used UA, BA and multivariate analyses to define the characteristics of blocks based on
- hazard and access/response variables, compared with the EIPN score. We defined the
- variables of the second and third groups as follows:

5.1. Second group of variables: volcanic hazards

- The second group of variables (Table 2) determines the hazard level of neighbourhoods inside
- and outside the hazard-prone zones outlined around Semeru (Fig. 3; see Thouret et al., 2007
- for the type, occurrence and frequency of lahars, PDCs, and tephra fallout.)
- **V12**: Volcanic Explosive Index is ranked according to current, daily (since 1967) and chronic,
- episodes of increased eruptive activity: VEI 1-2 unrest and mild activity, VEI 3 every 8 to 11
- 658 vears on average, and VEI $>$ 4 every 25 years on average.
- **V13**: Two types of lahars are debris flows (DF) and hyperconcentrated flows (HCF), ranked
- in three attributes: DF and/or HCF, HCF alone, and absent. Frequent lahars occur every rain season, and shortly after eruption.
- **V14**: Lahar frequency is in decreasing interval order from high, week or months, to low: > 25 years.
- **V15**: Two PDC types are valley-confined pyroclastic flows affecting valley channels and
- banks, and un-confined, pyroclastic surges affecting valley terraces and sometimes adjacent
- interfluves. Because valley margins are populated, we ranked surge first, then confined PFs,
- and last, interfluves exceptionally affected > 0.5 km from valley channels.
- **V16**: The PDC frequency range follows the same time interval order from high (1-8 years) to 669 $\text{low} > 25 \text{ years}.$
- **V17**: Tephra fallout types are twofold: ballistics and lapilli within 8-9 km from the vent and 1 for tephra-fall beyond the 8-9 km distance.
- **V18**: Tephra-fall frequency is ranked in decreasing interval order from < 1 year to 5-25 years.
- **5.2. Third group of variables: access and response**
- The third group defines the access to blocks and the response to imminent eruption including
- evacuation (Table 2), ranked on distance and quality criteria. We attributed the maximum
- mark to the poorest quality or to the absent attribute of five variables, as the ultimate goal was
- to rank a range of blocks based on exposure and remoteness in case of crisis.
- **V19**: Three types of access ways according to the decreasing quality of the network, from paved road, dirt road to trail.
- **V20**: Evacuation roads, shelters and storage facilities include distance (5 km-threshold, i.e.,
- 20 minutes driving a 4x4 vehicle or c. 1 hour walking) and absence from the *dusun*.
- **V21**: Civil Protection works, such as dykes and check dams, are present or absent.
- **V22**: Early warning system and local offices of the Indonesian board for [natural disaster](https://en.wikipedia.org/wiki/Natural_disaster)
- management (*Badan Nasional Penanggulangan Bencana* BNPB) are ranked according to
- distance (2.5 km-threshold, i.e., within walking distance) or absence.
- **V23**: Health centres: four hospitals around Semeru are located in Dampit, Pronojiwo, Pasirian
- and Lumajang, but at least one small hospital or dispensary is present in all *Desa*. Distance
- (10 km- and 5 km-threshold) and presence or absence helped rank the health centres.

5.3. Attributes of representative blocks based on univariate analysis

- Univariate analysis (UA, Fig. 7) was used to compute the frequency of attributes of all
- variables from the dataset including 145 *dusun* blocks.
- First, the most frequent attributes allow us to elaborate on representative blocks (Fig. 7):
- (1) As many as 84% of blocks are built and permanently occupied, 75% are located between 9
- 694 and 12 km from the summit, and most of them are located on high $(> 10 \text{ m})$ terraces at mid-
- distance (> 120 m) from the valley channel. (2) About 62% have recorded 0 fatalities, but
- most of them can be affected by light ashfall within 1 to 3 years interval; almost 27% of
- blocks can be damaged by lahars in valleys nearby, and 10% to 19% recorded PDC- and
- lahar-related fatalities, respectively. (3) The construction quality of 75% of homes is regular.
- (4) About 61% of blocks are close to a shelter or storage facility and c. 77% of blocks lie
- 700 within 10 km of hospitals. However, this means that c. 23% of blocks lie ≥ 10 km from
- hospitals, while 51% remain poorly connected to other response facilities.
- Second, UA results highlight the most relevant variable attributes among the blocks (Fig. 7):
- (a) the population density of *dusun* and built blocks is discriminant, but higher densities of
- built blocks increase human exposure to volcanic threats; (b) the location of c. 24% of blocks
- on low to middle terraces and within short to middle distance to river channel may be more
- discriminant than the location of prior fatalities (as c. 62% of blocks record no death) and
- timely location of people (as homes in 83.50% of blocks are permanent).

 Third, strong contrasts in exposure of inhabitants are due to the block location with respect to rivers due to lahar and PDC type, frequency, extent, and impacts. Blocks without reported fatalities prevail (c. 62%) in *dusun,* which long settled away from the active valley channels. Among PDCs, the predominant, un-confined surges reflect the correlation between fatalities and valley terraces on which few *dusun* have recently spread out. The impact of the daily tephra-fall hazard on blocks is binary: very few *dusun* lie within the 8-9 km radius from the vent affected by ballistics and lapilli from the recurrent, low-altitude columns at the summit, whereas decreasing tephra-fall frequency beyond 9 km has little impacts on *dusun* blocks if 716 VEI remains low (< 3) .

- Fourth, hazard impacts depend on perpendicular or oblique orientation of buildings, which
- renders almost 63% of homes exposed to flows, while the poor (5.5%) and regular (c. 76%)
- 719 quality of roofs would not withstand thick (\geq 25 cm) and wet tephra-fall deposit (Fig. 4).
- About 51% of the blocks show poor access ways and most of them (c. 85%) lack civil
- protection works (dykes, dams) against volcanic flows, while check dams across active
- valleys remain filled up or damaged. In contrast, three variables of the third group may counterbalance eruption impacts, therefore reducing block exposure (Fig. 7): (a) Almost half of the ways are paved roads, favouring access to *dusun*; (b) evacuation shelters, roads signs and storage facilities do exist in c. 61% of blocks, and; (c) early warning system and BNPB offices are close enough (< 2.5 km) to c. 81% of blocks, while c. 77% of *dusun* lie within 10
- km of any hospital.

5.4. Block definition based on BA and MCA of attributes

 MCA was used to find statistical relationships between the exposure index score and other attributes that belong to hazards, access and response.

First, contributory variables of MCA were tested by means of Chi-square test (Table 5) to find

out how dependent they are with other contributory variables to whom they are correlated.

- 733 The Chi² test shows a strong statistical link (95% confidence) between variables indicated as
- dependent in Table 5 (bold figures) and ESD Table 3. There are six exposure variables: V2
- density of *dusun* and V3 density of built block population, V8 Terrace elevation/distance
- relationship, V9 Timely location of people, V10 House orientation, and V11 Roof and
- construction quality. Table 5 also include variables outside exposure that are dependent with
- the majority of variables: V13 Lahars, V15 PDCs, V16 PDC frequency, V18 tephra-fall
- frequency, V19 access roads, and V22 Early warning system and BNPB office.
- Second, colour-coded MCA biplots show how the attributes contribute to the definition of
- blocks according to the Dimensions 1 and 2 (Fig. 8A) and Dimensions 2 and 3 (Fig. 8B).
- These graphs display proximities and oppositions between all attributes: the darker the colour,
- the more contributory the attribute is; conversely, light colour points to less contribution.
- MCA biplots (Figs. 8A, B) distinguish four groups of attributes based on contributions and
- squared cosines (ESD Table 6).

5.4.1. Four groups of attributes in the Dimensions 1 & 2 and 2 &3 biplots

 1. The right-hand side group along Dimension 1 include the following attributes: channel and low terrace/short distance (V8), both attributes DF or HCF (V13), 0 inhabitant density (V2, V3) together with the 'temporary location' attribute of V9. Such attributes point to un-built blocks located along the valley channels where shelters are temporary and people used to work daily and occasionally. These un-built blocks or temporary settlements, being not constantly exposed, show a residual to moderate EIPN. They recorded lahars and PDCs fatalities as they can be hit by both confined pyroclastic flows and unconfined surges, while their access by trail is challenging.

- 2. Two groups are positively and negatively correlated along Dimension 2 (Fig. 8A). The
- second group of attributes, positively correlated, define blocks with residual EIPN due to long
- distance and high elevation from the rivers (V8), absent lahars and PDCs (V13, V15),
- together with low frequency of tephra fall (V18). The relatively densely populated blocks are
- located within 12 and 20 km from vent, but they are devoid of evacuation shelter, early
- warning system, and civil protection office.
- The third group is observed below the graph barycentre (Fig. 8A), but the quality of
- contribution is weaker than that of the previous group. Several attributes include: 9 to 12 km
- (V1), high block density (V3), PDC surge (V15) every 8 to 25 years (V16), ashfall every 1 to
- 3 years (V18), and existing civil protection works (V21). These blocks, located in the most
- affected 9-12 km circle distance from the volcano summit, may be damaged by tephra fall and
- sometimes by lahars and PDCs, which characterize their moderate to very high EIPN.
- Figure 8B shows the contribution of variable attributes in the Dimensions 2 and 3 biplot,
- although the contribution is relatively weak. The small fourth group of correlated,
- contributory attributes appears in the upper right corner of the biplot: distance < 9 km (V1)*,*
- Lapilli and ash within 9 km from vent (V15), < 1 year (V16), and no health centre (V23)*.*
- 771 Such attributes point to blocks located near the volcano summit ($> 1,000$ masl) liable to be
- affected by tephra fall and far from available emergency facilities.
-

6. Statistical grouping of blocks; Application to relief operation in case of evacuation

6.1. Four groups of blocks result from MCA and HAC

 Attributes extracted from MCA (Fig. 8) and HAC (Table 6, ESD Fig. 4) point to four groups 777 of blocks, ranked according to the most discriminant variables (see the Chi² test conducted on HAC variables, ESD Table 8) and the EIPN score. These groups are similar to those defined using MCA biplots.

 Group 1 looks like the first group described in subsection 5.4.1: un-built blocks located along active valley channels and their margins are temporarily occupied by laborers or workers. The location explains why they are affected by all types of and frequent lahars and PDCs, inducing fatalities or injuries. Trails and dirt roads hinder access to these blocks, which remain blocked from any health centre. Examples are located on the margins of K. Bang and Kembar (*Dusun* Supit Timor), K. Koboan (*Dusun* Sumbersari), and K. Tengah (*Dusun* Tulungrejo) (Fig. 5A-C).

 Group 2 is similar to the second group (see 5.4.1). Built blocks with residual EIPN are located farther away (12 to 20 km) from the volcano, on low slopes and in the ring plain. These blocks are occasionally affected by distal lahars along the active valleys and light, relatively un-frequent tephra-fall associated with long-reach PDCs from VEI≥ 3 eruptions. Most of them are located away from the valley channels, but low-frequency surges may reach them during large, un-frequent eruptions. What makes the exposure residual is the average density of both blocks and *dusun*, and the lack of response facilities. Examples are Kajar Kuning, Gumuk Mas, Oro-Oro Ombo, and Jabon (Fig. 5B, and C).

 Group 3 looks like the third group (see 5.4.1). Densely populated built blocks within highly populated *dusun*, located between 9 and 12 km distance, have long settled on high terraces away from the valley channels. Un-affected by lahars and valley-confined PDCs, these blocks can be impacted by relatively frequent tephra-fall and low-frequency surges. Such blocks

- exhibit moderate EIPN, when response facilities are close and they are protected by civil
- protection works (e.g., Oro-Oro Ombo and West of Sumbersari, Fig. 5B). Other blocks
- exhibit high EIPN if they remain un-protected and remote from response facilities (Supit
- Timor, Rowobaung, Fig. 5A; Curah Koboan, Fig. 5B).

 Group 4 resembles the fourth group (see 5.4.1). Blocks within 9 km from Semeru's vent, frequently affected by light tephra-fall, are located higher up on the volcano's flanks, far from the main roads, response facilities such as hospitals. These blocks with poor house construction quality and small dispensaries exhibit very low or low EIPN, such as Blubuk and Karangsuko (Fig. 5D), because the people density is low and they are affected by light ashfall only.

6.2. Application to relief operations based on access to blocks and response facilities

 Our method to characterize blocks around Semeru was applied to rank them to highlight potential challenges in evacuation procedure. First, the distance/time criterion is computed to evaluate evacuation based on access to the blocks and means of transport. This criterion was related to distances to the shelters and/or storage facilities and health centres, as shown by Table 7 and ESD Table 9, ESD Figure 5. Second, we used the results of analysis of attribute frequencies (Table 7) to distinguish a range of blocks based on access to response facilities. 816 Factor map (ESD Figure 5) helps distinguish the number of clusters. Maps in Figure 9 display a range of four block clusters based on access versus remoteness in case of evacuation. Blocks accessible only by dirt roads (e.g., Sumbersari, Curah Koboan along K. Koboan, Fig. 9A; Tulungrejo along K. Tenggah, Fig. 9B; Supit and Supit Timor along K. Bang, Fig. 9C) are challenging for relief operations in case of evacuation. This was reflected by the cumbersome evacuation of injured and affected people following the 4 December 2021 eruption on both mid reaches of the Koboan valley. *Dusun* like Rowobaung (Fig. 9C) linked to Pronojiwo by one bridge would remain cut off by lahars propagating along two rivers surrounding the blocks. *Dusun* located high on the Semeru' slopes with narrow dirt roads (e.g., Blubuk and Karangsuko, Fig. 9D) include upstream blocks at high elevation (> 1,000 m asl) connected only by trails that may become blocked by heavy tephra fall. The town of Senduro, far from active rivers and well connected by paved roads to Lumajang, can be evacuated, but the narrow street network in densely populated *dusun* may be challenging (Fig. 9E).

7. DISCUSSION

- Table 8 compares the number of exposure parameters from the literature with our local-scale, multi-component EIPN method. Specific characteristics of exposure arise from this comparison.
- Globally, two criteria measure human exposure index using the distribution of population potentially affected by a volcanic eruption: either the (log) number or the density per area (e.g., Wild et al., 2021; Nieto-Torres et al., 2021). The majority of publications, following Yokoyama et al. (1984) and Ewert (2007), consider the number or density of population within circle distance or radii from the main crater. Recent studies correlated the circle- distance thresholds with fatalities (Auker et al., 2013; Brown et al., 2017) and population density (Freire et al., 2019). A limited number of studies encompass several parameters of human exposure (e.g., five among 10 parameters describing exposure on the Nisyros Island; Kinvig et al., 2016). The most complete study to date in ranking volcanic risk (Nieto-Torres et al., 2021) includes nine exposure parameters within a total of 41 covering hazard, exposure, vulnerability and resilience. However, their single human exposure parameter is the density of
- population within 5, 10, 30 and 100 km radii.

 In contrast, the multi-component EIPN is based on six optimal variables using PLR models 846 from a set of 11 initial exposure variables. Instead of a simple statistical technique to weigh and sum exposure parameters, statistical operations enabled us to convert the qualitative variables in semi-quantitative criteria and to elaborate on normalization of attribute values and discretization of EIPN scores. Then polytomous logistic regression models quantified the link between the probability of a EIPN score and the variables of exposure. For research efficiency, future surveys should collect as few optimal variables as possible to obtain the EIPN score for blocks potentially around any populated, active volcano.

853 In the literature, the global-scale 'PEI' underestimates local factors that induce spatial and temporal exposure patterns. In contrast, our work highlights local factors that combine on a multi-component human exposure. Local factors that govern the extent to which settlements 856 are exposed to specific threats involve: (1) topography and geomorphological location near 857 valleys that convey most of the flows; (2) fatalities location related to hazard type and 858 frequency, which affect human habitat and livelihoods, and; (3) home and roof quality that 859 may increase exposure of residents. We consider exposure related to volcanic flows (PDCs, lahars) inasmuch these hazards induce impacts on settlements located near active river 861 channels and adjacent low banks. We thus accounted for shallow and sinuous valley channels,

- which favour overbank pyroclastic surges, overbank lahars and associated floods along the
- most active rivers (Fig. 3). We do not discard outside hazard-prone areas affected in the recent
- past by lahars and associated floods produced by overbank and avulsion into secondary
- drainage. On the other hand, we also considered local factors that may contribute to decrease
- 866 exposure, e.g., easy access on paved roads to the *dusun*, short distance/time (\leq 5 km and \leq 25
- minutes by car depending on the quality of roads) to health centres, emergency facilities, and
- finally the existence of civil protection works.
- Measurements of exposure depends on the time frame at which assessments are conducted
- (Auker et al., 2015). We examined exposure not only in primary residence (Wild et al., 2021),
- but also in working areas: schools, shops, farmland and valleys, as many men extract
- construction material from lahar and PDC deposits in active river channels. Here, the
- 'counting record' of events and impacts held by the Dutch and Indonesian Volcanological
- Survey since 1884 ensures completeness of the Semeru daily explosive activity and its
- multiple chronic spurts. Monitoring from the Semeru Observatory at Gunung Sawur, 12 km
- SE of the summit (Fig. 3) contributed to the eruption record as early as 1953.
- Few studies address exposure of highly populated communities on a persistently active
- volcano. Instead of sporadic (e.g., Mayon) or chronic eruptive activity (e.g., Merapi),
- 879 Semeru's acute and constant, daily explosive activity since at least 1967–the longest daily
- 880 explosive unrest worldwide with the exception of Sakurajima-exerts a heavy, constant toll on
- 881 human life, thus on perception and adaptation of communities exposed to volcanic risk.

8. CONCLUSION

- The local-scale method, including field data collection, mapping and a range of statistical
- techniques, helped compute four levels of a multi-component index of exposure applied to
- 145 neighbourhoods mapped in 13 *dusun* and two small towns.
- Polytomous logistic regression models allowed us to select six optimal variables and predict
- the EIPN score of blocks These optimal variables are reproducible parameters to assess
- human exposure on active, populated volcanoes.
- Computing and mapping human exposure at the block scale may be more adapted to: 1) the
- characteristics of the population, (2) the diversity of hazards and timely change in exposure to
- a persistently explosive activity, and (3) the mixed rural-small urban communities with a
- variety of resources, which support livelihoods and sustain the community coping capacity.
- Ranking blocks using distance-timing and access to response facilities is a useful tool to point
- blocks that need relief operations to be implemented. Civil authorities may provide advice and
- funds to retrofit home construction, relocate exposed homes that encroached on low terraces,
- clean up check dams, pave access ways to, and implement dispensaries in remote villages.
- Results should help disaster risk management staff to improve their participation at the scale
- of neighbourhoods on active volcanoes.

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Declarations

- *Authors' contribution statements*. All authors contributed to the study conception and design.
- Material preparation, data collection and analysis were performed by JC Thouret, M
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TABLE CAPTIONS

 Table 1. Methods, techniques and objectives of the statistical analyses (UA, BA, FCA, MCA and HAC) conducted on variables and attributes of the Exposure Index of Populated Neighbourhoods (EIPN). Observations are neighbourhoods termed blocks. Polytomous Logistic regression (PLR) is a model to extract most pertinent variables for predicting EIPN scores of blocks, which can be applied outside the survey area and on any populated, active volcano. HAC is a clustering technique to distinguish groups of variable attributes that describe a range of exposed blocks.

 Table 2. Variables and their attributes (with thresholds) used to compute the EIPN at the local *dusun* (sub-village) and block (neighbourhood) scale. **A**. Eleven variables define what can increase population exposure. **B**. Seven variables linked to hazard types, occurrence and frequency. **C**. Five variables describe accessibility to blocks and the existence/distance of response facilities in case of imminent eruption and evacuation. Ordinal numbers (without weighting) were attributed as thresholds of attributes within each variable, but different values for each attribute were normalized for computing the EIPN, while the index score intervals were discretized using the Jenks method. We calculated weighting coefficients associated to the variable no.4 using the method of intersection of probabilities adapted from Brown et al. (2015), see Table 3.

 Table 3. Four steps describing the method for computing the exposure index adapted from the 1112 People Exposure Index (PEI: Brown et al., 2015a) according to four circle distances (9, 12, 18, and 35 km) around Semeru, taking into account the recorded volcanic events, village population, and reported fatalities since 1884.

 Table 4. **A**. PLR initial model with 8 variables and their attributes with coefficients of explanatory variables, significativity and thresholds of EIPN coefficients. **B**. PLR selected model showing the 13 attributes of 6 variables with their estimated coefficients and 'significativity'. We selected the most significant variables (3 stars), meaning < 0.1% error in probabilities, although a less strict selection would also retain 2-stars attributes, i.e., with < 1% error in probabilities.

 Table 5. Results of the Chi-square test conducted on the three groups of variables. The dashed black line indicates the results for the 11 exposure variables. Bold P-values indicate variables that are independent (no statistical relationship) because p-value exceeds 5% based on the fact

- that a few sample cohorts do not exceed 5 observations (ESD Table 3). P-value, a probability
- number between 0 and 1, is defined as the probability of getting a result that is either the same
- or more extreme than the actual observations. Almost all p-values are significant for a
- threshold of 5%. Hence, we pose two hypotheses: two variables are independent if p-value
- 1128 exceeds 5% (i.e., the null hypothesis H_0), against the alternative H_1 , i.e., two variables are
- 1129 dependent if p-value is $\lt 5\%$. While taking account of the size of the table of contingency
- (number of freedom degree), the Chi-square test computes the deviation between observed
- and theoretical counts (i.e., those expected if two variables were independent). Upon
- assessing this deviation, the hypothesis of independence is accepted or rejected.
- **Table 6**. Results from Hierarchical Analytical Clustering (HAC) showing four clusters or
- groups of blocks based on the frequency of attributes, as shown by the outputs of MCA. See
- 1135 Factor map, ESD Figure 4 that helps support the distinction between four clusters of blocks.
- **Table 7**. Four clusters of at-risk blocks in *dusun* based on the distance/timing criterion, access
- to blocks and response facilities. See Factor map, ESD Figure 5 and ESD Table 9 that help
- support the distinction between four clusters of blocks in case of evacuation.
- **Table 8**. Comparison of parameters used in the Exposure Index of Populated Neighbourhoods at Semeru with previous methods and parameters defining exposure within risk studies from the literature.

FIGURE CAPTIONS

- **Figure 1.** Flow chart showing four research steps and how we defined, computed and mapped an Exposure Index of Populated Neighbourhoods (EIPN) at the local scale of sub-village (*dusun*) blocks (neighbourhoods). The fourth research step is an application of the local scale method to identify remote and/or blocked blocks for imminent eruption including evacuation.
- **Figure 2**. **A**. Map of Semeru volcano and ring plain in the middle of the regencies of
- Lumajang to the east and Malang to the west. Distance circles 9, 12, 18 and 35 km (dashed
- white lines) with respect to the persistently active vent. Principal cities and towns are
- indicated. Red circle in map **B** points to the Semeru-Tengger massif, East Java. M= Malang,
- 1151 $S = S$ urabaya.
- **Figure 3**. Hazard-zone map of the Semeru's slopes and ring plain depicting the extent of volcanic phenomena in the case of a medium-sized (VEI 3) eruption, based on the 1994, 1995
- and 2002 eruptions and post-eruption lahars (Thouret et al., 2007). The map also shows the
- extent of lahars and floods in the case of a catastrophic eruption (VEI > 3) along valleys
- through the distal south and east ring plain, based on the 1909 and 1981 events and the
- disaster-prone areas map (Bronto et al., VSI, 1996). Initials indicate the surveyed 13 *dusun*:
- B-K Blubuk, Karangsuko, SU Supit, RB Rowobaung, SB Sumbersari, GM Gumuk Mas, CK
- Curah Koboan, KK KajarKuning, TU Tulungrejo, JA Jabon, SM Sumbermulyo, JL
- Jaranglangak, RE Rekesan, and two towns of Pronojiwo and Senduro.
- **Figure 4.** Results of univariate analysis (given in percentage) conducted on 43 attributes of 11 EIPN variables applied to all studied *dusun* blocks. The grayscale block attributes, from the lowest exposure index (light grey) to the highest exposure index (dark grey), are similar for all plots.
- **Figure 5**. Maps showing the EIPN score at the scale of each *dusun* block. **A**. Town of
- Pronojiwo, *Dusun* Supit-Supit Timur, and Rowobaung. **B**. *Dusun* Sumbersari, Gumuk Mas,
- Curah Lengkong, and Kajar Kuning on both sides of K. Koboan. **C**. *Dusun* Tulungrejo,
- Tawonsongo and Jabon in Desa Pasarjumbe. **D**. *Dusun* Blubuk and Karangsuko*.* **E.** *Dusun*
- Sumbermulyo, Juranglangak, and Rekesan in *Desa* Senduro. The four colour-coded final
- scores of the EIPN (obtained from ESD Table 6) show how one of the exposure index score
- levels were assigned to every block.
- **Figure 6**. Scatter plots derived from MCA results showing the most opposed attributes (i.e.,
- away from the plot barycentre) of 11 contributory variables for exposure along the
- Dimensions1 and 2 that convey 76% of the information.
- **Figure 7.** Results of univariate analysis (given in percentage) conducted on all attributes of
- the second group and third group of variables applied to all studied *dusun* blocks. The seven
- hazard variables are V12 to V18. The five variables V19 to V23 describe access to blocks and
- the existence and distance of response facilities.
- **Figure 8.** Results from Multiple Component Analysis (MCA) on variables and their attributes
- correlated to the EIPN score. **A**. Plot showing contribution of variables along Dimensions 1
- and 2 (76% of the information). **B**. Plot showing contribution of variables along Dimensions 2
- and 3 (23% of the information). Percentages of information are shown in ESD Figure 3.
- **Figure 9**. Maps of four clusters of blocks in *dusun* shown in Figure 5 based on
- distance/timing criterion (including the quality of access and means of transport) to the
- response facilities (shelters, storage facility, early warning system, BNPB offices and health
- centres). The four block clusters stem from the high/low frequency analysis (Table 7) and
- HAC Factor map (ESD Figure 5). **A.** Greeen: Easy access to blocks (10-25 minutes) and
- short distance response facilities (2.5–5 km). **B.** Yellow: Slow access to blocks (25 to 45
- minutes) and short- and mid-distance response facilities (2.5–10 km). **C.** Orange:
- Challenging access to blocks (45 to 170 minutes) and long-distance (> 10 km) response
- facilities. **D.** Red: Delayed access to blocked blocks (80 to 170 minutes) and remote or absent
- response facilities.

Electronic Supplement Data ESD

- ESD Figure 1. Maps showing the setting of the 15 *dusun* together with the blocks in which we
- conducted field survey and statistical analyses on exposure parameters. A. Karangsuko and
- Blubuk (Desa Tamansatryan), west flank. B. Town of Pronojiwo, Supit-Supit Timur and
- Rowobaung (Desa Pronojiwo), South flank. C. Oro-Oro Ombo (Desa and *dusun*), Sumbersari,
- Gumuk Mas, Curah Lengkong (Desa Supit Urang), SSE and SE flank. D. Kajar Kuning (Desa
- Sumberwuluh) and Desa Candipuro, SSE flank. E. Tulungrejo and Jabon (Desa Pasrujambe),
- ESE flank, and F. Sumbermulyo, Juranglangak, and Rekesan (Desa Senduro), East flank.
- ESD Figure 2. Scree model with distribution of information according to dimensions.
- ESD Figure 3. Factor map obtained from HAC showing four groups of blocks based on
- attribute frequencies: see Table 6 for the list of high and low attribute frequencies.
- ESD Figure 4. Factor map obtained from distance/timing criteria and HAC (Table 7) and showing four clusters of blocks according to access and response facilities.
- ESD Table 1. Setting of surveys carried out in *dusun* (sub-villages): administrative units,
- location, surface area, people density, and number of surveys in each *dusun*. Symbol meaning:
- * data from BPS reports, *Kecamatan Dalam Angka* 2019, and 2018 for Tamansatryan,
- Sumberwuluh, Candipuro. **A *dusun* usually includes 4 to 5 *RukunWarga* (RW, a
- neighbourhood with 50 to 75 houses). A RW includes usually 3 to 9 *RukunTetanga* (RT, a
- block with 20 to 25 houses). Field survey was carried out at the scale of RWs, including more
- than one observation per RT.
- ESD Table 2. Coordinates of buildings, economic status of respondents, and geographical
- exposure with respect to active valleys.

ESD Table 3. Chi-square test on the set of 23 variables to determine whether two variables are

independent or dependent. In this case, a variable is independent if the p-value exceeds 5%

(see Table 5). As a result, a statistical link exists (95% confidence) between variables

indicated as dependent with corresponding variables listed in the first column. Dark grey

indicates variables of exposure, grey variables of hazards, and white variables of access and

response.

 ESD Table 4. Burt Table of contingency (all attributes are considered) showing statistical links between attributes of two variables at a time.

 ESD Table 5. Coordinates, squared cosine, and contribution of attributes used in MCA biplots.

ESD Table 6. Master Table of computed EIPN per *dusun* blocks, totalling 145 (horizontal

rows; two initials indicate the *dusun* name) according to all exposure variables and their

attributes (vertical rows, see Table 3). A. The colour-coded final scores of the EIPN are

displayed at the end of the Table as well as in Figure 5. All blocks delineated in Figure 5 A-C

and D-F were attributed one of the colour-coded Exposure Index score levels.

ESD Table 7. Confusion matrix of the selected PLR model. This Table crosses 'real' observed

EIPN scores with predicted ones when we applied the model to the initial (observed) data

(145 blocks). Grey boxes show well predicted EIPN values in contrast to yellow boxes

indicating poorly predicted EIPN values.

1234 ESD Table 8. Chi² test on discriminant variables that support HAC clusters.

ESD Table 9. Chi2 test on variables of timing, access and response that support block clusters

for relief operation in case of imminent evacuation.

Clermont, September 24, 2022

Dr. James Goff, Editor in Chief,

Natural Hazards Journal

We would like to submit the revised article entitled: "*Semeru volcano, Indonesia: Measuring hazard, exposure and response of densely populated neighbourhoods facing persistent volcanic threats*" to Natural Hazards.

In response to your request, we apologize and we thank you for having drawn our attention to the short paragraphs that may be considered as plagiarism and needed to be reworked before peer review. We have revised the third paragraph in sections 1.1, the first and third paragraphs in section 1.2, the second paragraph in section 1.3, and finally the first paragraph in section 1.5. We have shortened and reworked the mentioned lines, indicated their sources and added citations where appropriate. In the case of sections 1.3 and 1.5, we have referred to the publication from which the lines have been borrowed. In fact, we borrowed the content of these lines from our publication "Thouret et al., 2022" (International Journal of Disaster Risk Reduction). We hope that the revision would meet your expectation before peer review.

The research presented in this article, which was initiated at Laboratoire Magmas et Volcans (LMV), Université Clermont-Auvergne (UCA) in France, has been conducted in collaboration with the Laboratoire de Mathématiques (also UCA), and with Indonesian partners of the IPB University, Bogor in the framework of the "RiskAdapt" research program funded by the French National Agency for Research (ANR).

The article shows the population exposure to Semeru's volcanic threats from its persistent, daily eruptive activity, which endangers at least 50,000 of the 950,000 inhabitants living on its slopes and ring plain. Surveys, mapping and statistical investigation enabled us to assess the extent of exposure of 145 neighbourhoods (termed blocks) and characterize hazards and response to eruptions in 15 rural villages and small towns. Univariate, bivariate and multivariate analyses were used to explore data and characterize the relationships between 11 variables to compute a multi-component exposure index. Logistic Regression models allowed to select six optimal exposure variables. Multivariate analyses and Hierarchical Agglomerative Clustering distinguished four groups of blocks based on attributes of all variables correlated to the exposure index score.

To contribute to disaster risk reduction, ranking blocks using distance-timing and access to response facilities is a useful tool to point remote or blocked blocks during imminent eruptions including evacuation. Results should help disaster risk management staff to improve their participation at the scale of neighbourhoods on active volcanoes.

We hope this article meets the scope and high standards of Natural Hazards. Thank you very much for your consideration.

Sincerely yours,

Jean-Claude Thouret, First and corresponding author, Professor, Université Clermont Auvergne (UCA), OPGC, CNRS and IRD, Laboratoire Magmas et Volcans, F-63000 Clermont-Ferrand, France. Email: j-claude.thouret@uca.fr; Phone +33 4 73 34 67 73; cell phone 33 6 25 19 41 17

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Measuring exposure, hazard and response of densely populated neighbourhoods facing Semeru's persistent volcanic threats

Field surveys Data acquisition

STEP 1. Computing the Index score to rank exposure of *dusun** blocks (neighbourhoods)

Comparison between **exposure parameters** , and **scale** (global, local) **of exposure analysis**

STEP 2. Polytomous Logistic Regression models Initial model: 8 variables Selected model: 6 variables

STEP 3. Mapping blocks according to the Exposure Index score (EIPN)

Uni-, Bivariate analyses, Factorial and Multiple Correspondence Analyses, Hierarchical Agglomerative Clustering, Polytomous Logistic Regression

Uni-, Bivariate analyses, Factorial and Multiple Correspondence Analyses,
Hierarchical Agglomerative Clustering, Polytomous Logistic Regression

STATISTICAL METHOD APPLIED TO THE SURVEY DATASET

STATISTICAL METHOD APPLIED TO THE SURVEY DATASET

STEP 4. Ranking and mapping at-risk blocks in case of imminent eruption

OBSERVATIONS (23 variables, 145 blocks) **RESULTS**

 Eleven variables that may increase exposu re: V1. Village distance to vent V2. Population densi ty per village (or *dusun**) V4. Prior f atalities/popul ation/ci rcle distan ce V8. Location of neighbourhoods from river channel V9. Timely loc ation of people V10. Building exposu re: orientation V6. Lahar fatalities V3. Population density per built block area V7. PDC fatalities V5. Fatalities location V11. Building exposure: construction quality

> **Six optimal variables from the set of 11** 1 **exposure variables:** V1, V2, V3, V5, V6, V7

 V12. Volcanic Explosivi ty Ind e x (VEI) V13. Lahar occurrence V14. Lahar frequency V15. PDC occurrence **HAZARD Level defined by 7 variables:**

V16. PDC frequency V17. Tephra-fall occurrence

V18. Tephra-fall frequency **ACCESS** to neighbourhoods and

V19. Road access │ V20. Evacuation shelter, road sign, and storage facility V21. Civil defense work V23. Health centres V22. Early warning system and Civil protection office

Distance/timing criteria based on access ways and means of transport, and access to response facilities

Statiscal distribution and dependence between 11 variables

Exposure index (EIPN) score: four levels, from residual to very high

Gaps indentified in literature on exposure definitions and parametrization

Modeling EIPN **Prediction capacity** outside the study area

Definition of blocks

based on correlation of hazard level, access, and response facilities with EIPN score

RESPONSE FACILITIES 5 variables: characterizing four groups t h e among 145 blocks

> **Application to relief operations** in case of evacuation

Areas frequently affected by tephra fallout on annual basis (VEI 2) and ballistics within 5 km

Areas affected by PDCs and companion ashfall due to large VEI > 3 events every 11 to 25 years on average

Areas likely to be affected by tephra-fall associated with PDCs or by fallout in case of large eruptions

Preferential path (scar of Jenggring-Seloko) guiding dome-fed rock **and study of Selong Preferential path (sub-village**) and small towns under study
avalanches, lava flows and pyroclastic flows on annual basis

- \bullet
-

Possible overbank and avulsion in case of large-volume lahars Alluvial plains affected by lahars in case of large-scale eruptions (VEI \geqslant 3) and/or heavy rainstorm

Areas likely to be mantled by annual ashfall associated to PDCs and dispersed in case of large eruption: towards East & SE (rain season), West & SW (dry season)

Valleys swept by large-volume lahars (>1 million m³), e.g., 1909, 1976 and 1981

Dykes and check dams (Sabo) along active valley channels

Fig. 4

Figure 5 ABC

[Click here to access/download;Figure;Figure 5 DE. Map Exposure Index dusun](https://www.editorialmanager.com/nhaz/download.aspx?id=501919&guid=5556153a-224d-4646-9338-13ac0ad27cf6&scheme=1) blocks.jpg

 \equiv

Figure 6 [Click here to access/download;Figure;Figure 6. MCA Scatterplot axes 1-2 MT](https://www.editorialmanager.com/nhaz/download.aspx?id=501920&guid=194d1cb2-1e5c-41cb-a0f0-b4191003e38d&scheme=1) (27.04.22).pdf

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Table 2 [Click here to access/download;Table;Table 2. Three groups variables EW_05-08-22.pdf](https://www.editorialmanager.com/nhaz/download.aspx?id=501925&guid=4a8e451b-5342-4166-bc7e-7dd3e006f4cd&scheme=1) \pm

Table 2.1. Eleven variables (V) increase exposure.

Table 2.2. Second group of variables: volcanic hazards posed to the surveyed neighbourhoods.

Table 2.3. Third group of variables: accessibility of neighbourhoods and response facilities.

Table 4 [Click here to access/download;Table;Table 4 A B. Initial and Selected PLR models.pdf](https://www.editorialmanager.com/nhaz/download.aspx?id=501927&guid=62a33bf7-7b82-41c0-83e2-6a7a686e3d23&scheme=1) \pm

Table 4B. Selected model for polytomous logistic regression

Table 4A. Initial model for polytomous logistic regression.

Distance/timing criterion according to access ways and means of transport. The distance/time criterion displays three situations: the darker grey is, the longer it takes to access any given remote block.

attachment to manuscript ESD Tables 1 to 8

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attachment to manuscript ESD Figures 1 to 4

Click here to access/download [Electronic Supplementary Material](https://www.editorialmanager.com/nhaz/download.aspx?id=501933&guid=bfefe7db-00d3-4d7a-baec-56e946d473d8&scheme=1) ESD Figures 1 to 4.pdf