Living with the Merapi Volcano: Risks and Disaster Microinsurance

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June 2014

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ABSTRACT This paper investigates the influence of perception of natural disaster risks on the probability of local people to participate in a hypothetical disaster microinsurance. We use household data for a specific disaster risk of the Mount Merapi in Java. We find that respondent’s perception of natural disaster risk is in line with experts’ risk assessment. Estimation results show that this perception positively influences the interest to participate in disaster microinsurance. We also find that insurance literacy has a strong positive relationship with the respondent’s interest to participate in disaster microinsurance.

Keywords: perception of risks, eruptions, microinsurance, Merapi
JEL Code: Q54, G21, D12

Acknowledgements

Aloysius, Henri, Budy and Wouter dedicate this paper to the memory of Piet Rietveld who sadly passed away on November 1, 2013 after a short illness. The authors would like to thank Danang Darmawan of Gadjah Mada University for coordinating the household survey for this paper and Aryanto Steyn of BPBD DI Yogyakarta for sharing useful information about the current situation of the Merapi volcano. The first author acknowledges a scholarship from the Directorate General of Higher Education Ministry of Education and Culture of the Government of Indonesia.
“Risk and disaster emerge from volcanoes, but livelihood opportunities emerge from volcanoes too.”
(Kelman and Mather, 2008: 195)

I. Introduction

A report by the International Strategy for Disaster Reduction (ISDR) mentions that “a natural hazard is only a disaster because people are in the wrong place at the wrong time, had no choice but to be in the disaster or were caught unaware when it struck” (ISDR, 2004: xi). Meanwhile, the World Bank (2001) argues that natural disasters are a typical covariate risk that can affect everyone in the community at various levels. These statements indicate that certain locations will be more prone to certain types of natural events than others underlining that location is one of the important characteristics of natural disasters, besides their magnitude as natural shocks.

Some studies show that low-income households often face difficulties to replace their destroyed assets and to sustain their welfare following a natural shock (for instance, Carter et al., 2007; Skoufias, 2003). The self-insurance strategies of poor households (for instance seasonal migration, savings and livestock, or informal mutual support mechanisms) are also costly in terms of current income and future welfare (Skoufias, 2003). The consequences of such inequality on wealth is fundamental as it will influence the level of investment in human capital in the short run and the long run; and will also determine how the economy and households adjust to shocks (Galor and Zeira, 1993).

It is argued that disaster microinsurance can play an important role for low-income households. This insurance is a specific type of microinsurance that can be defined as an insurance for low income people to help them to manage risks and vulnerability (Cohen and Sebstad, 2006; Heydel et al., 2009; Morelly et al., 2010; De Bock and Gelade, 2012). Large natural disasters in recent years such as the Aceh tsunami in 2004 have triggered discussions on the possibility to introduce disaster microinsurance in Indonesia but it is still a hardly
explored field (Heydel et al., 2009; World Bank 2011; World Bank and GFDRR, 2012). People in Indonesia are reluctant to buy insurance products (Heydel et al., 2009) although natural disasters are one of the important risks that can affect the households’ welfare (Viverita et al., 2011; Reinhard, 2012).

In the present paper, we focus on a specific natural disaster risk, notably the disaster risk caused by eruptions of the Merapi volcano, one of the most active volcanoes in Indonesia. The Merapi eruption in 2010 has been attributed as the largest eruption in more than a century and historically this volcano erupted every four years (Subandriyo, 2011; Surono et al., 2012; see, Figure 1). Specifically, we will relate household’s perception of natural disaster risks to disaster microinsurance. In addition, we also include some variables that have been observed by several studies on microinsurance and disaster insurance. We use household

Figure 1. The history of Merapi eruptions.
Notes: This figure is taken from Subandriyo (2011).
data collected through a survey that was conducted around 18 months after the eruption at
the time when some respondents were still living in temporary shelters. This survey
focussed on three villages in the Cangkringan subdistrict which are close to the top of the
Merapi volcano (Kepuharjo, Glagaharjo and Umbulharjo) and the Bokoharjo village in the
Prambanan subdistrict as a control area (see, Figure 2).²

The present study can be classified as a study at the first level of research on demand
to microinsurance (see, Cohen and Sebstad, 2006). As Cohen and Sebstad identified, research
at this level concerns key risks facing poor people, the impact of these risks, existing coping
mechanisms, the effectiveness of the coping mechanisms, and the potential role
microinsurance (or other financial services) can play.

We find that respondent’s perception of natural disaster risk is in line with experts’
risk assessment that is produced based on the Merapi Hazard Map. Logit estimation shows
that respondent’s perception of natural disaster risks strongly influences respondent’s
interest to participate in disaster microinsurance. Consequently local people’s perception of
natural disaster risks should be taken into account in designing and proposing a specific
insurance for people living in vulnerable areas to natural disasters. Insurance literacy also
has a strong relationship with the respondent’s interest to participate in disaster
microinsurance. Therefore, it is important to develop insurance literacy. Other important
variables are access to house relief, membership of local associations and health risk.

The next section discusses the link between location, perception of risks and disaster
insurance. This is followed by setting the scene of our study (Section III). We then discuss
the framework of the present study (Section IV). In Section V we present estimation results
and discussion as well as robustness tests. Section VI provides main conclusions.
II. Location, Perception of Risks and Disaster Insurance

A considerable response is expected when introducing microinsurance especially in developing countries experiencing negative shocks frequently. However, studies find that the rate of participation in microinsurance in developing countries is low (for a review, see De Bock and Gelade, 2012). Low participation is also found in countries with a well-established insurance system, such as in Japan (Naoi et al., 2010; Naoi et al., 2012) and the US (see, Akter et al., 2008; Longwell, 2013). A number of factors determine participation in microinsurance, but De Bock and Gelade (2012) conclude that the evidence on the low demand for microinsurance is still far from decisive, on the basis of a review of studies on this issue in the past 10 years.

Some studies show the role of locational factors as well as experiences of past shocks on the people’s perception of risk exposure to natural disasters (for instance, Brody et al., 2008; Brody et al. 2004; Siegrist and Gutscher, 2006). These perceptions could in turn affect households’ participation in microinsurance (De Bock and Gelade, 2012). Fiers and Carson (2009) find a statistically significant relationship between natural disasters and the demand for life insurance across states in the US confirming that natural disasters not only affect property insurance, but also life insurance. Naoi et al. (2010) find that many households in Japan do not buy earthquake insurance since it is too expensive and this insurance does not reflect regional differences in earthquake risks. Longwell (2013) suggests that lack of equity and a certain bias in the perceived earthquake risk among Californians tend to lower the participation rate in earthquake insurance. In a case study on the risk of flooding and cyclones in Bangladesh, Akter et al. (2008) find that the return period of natural disasters and the distance people live from the river significantly explain the household decision to participate in insurance. Wang et al. (2012), however, emphasise that disaster insurance participation is not directly influenced by this perceived risk of hazards when people expect the government to provide support.
Siegrist and Gutscher (2006) note that differences between expert and lay people judgments of risks have been found in several studies for various domains but they also underline that recent studies suggest that the lay people’s risk assessments are less veridical than the expert’s risk assessments. In the context of natural disaster, some studies present interesting results. Comparing the perceived risks and frequency of disaster from the survey results and the historical records of disasters, Wang et al. (2012) argue that Chinese people in general have a correct perception of the hazards in the areas they live in. It indicates that there is no substantial gap between experts and lay people in assessing natural disaster risk. Brilly and Polic (2005) show that experience with floods influences the perceived threat and concern related to them, based on their study of a flood-prone area in Slovenia. They also find that people are aware of the importance of insurance against floods resulting in an increase in the rate of insurance participation. Siegrist and Gutscher (2006) confirm that lay people’ risk perceptions and experts’ risk assessment of flooding risks in Switzerland are correlated, but the strength of this relationship differs across regions. For instance, they observe that respondents in the German-speaking urban area underestimated the risk they faced, while the opposite is found for the urban area in the French-speaking part.

We may argue that disaster microinsurance is an option for vulnerable people to balance between risks and opportunities from natural disasters. As noted by Kelman and Mather (2008: 190), one of the options for dealing with environmental hazards is to live with the hazards and risks implying that livelihoods are integrated with environmental threats and opportunities.

III. Setting the Scene: the Merapi Eruption

According to an official report, at least 79 districts and cities in Indonesia are highly vulnerable to volcanic eruptions (Bappenas and BNPB, 2010). Relating this information with the population data of the Population Census 2010 (BPS, 2012) and taking a focus on Java—
the most populous island in Indonesia—produces an estimate that more than 56 million inhabitants face eruption risks.

One of the most active volcanoes in Indonesia is the Mount Merapi located 25–30 km north of the city of Yogyakarta in Java. It is not surprising that Indonesian and international teams have extensively studied this volcano (Surono et al., 2012). The eruption in 2010 affected four districts, killed 386 people and caused losses and damages in the order of US$ 403 million (Bappenas and BNPB, 2011). An amount of US$ 150 million—mainly from the central government—was used to pay for the post eruption rehabilitation and reconstruction in two provinces (US$ 1= Rp 9,000). Thousand houses in the Cangkringan subdistrict in the Yogyakarta province were destroyed by the eruption, especially in Kepuharjo village.

Figure 2. Number of houses damaged in the Cangkringan subdistrict due to the Merapi eruptions in 2010.

Notes: Map is taken from Surono et al. (2012: 122); data on houses damaged is taken from BNPB (2010), data per 19 November 2010. Villages with (*) are the focus area of survey.
Glagahharjo village, Umbulharjo village and Wukirsari village (see, Figure 2). The first three villages are the focus area of the survey from which the results are used in the present paper.

In response to the large 2010 eruption, the Center of Volcanology and Geological Hazard Mitigation (CVGHM) updated the Merapi Hazard Map (Figure 3) by employing information on geomorphology, geology, historical records of eruptions in the last 100 years (for instance, Figure 1) as well as the spatial distribution of its impacts (for example, Figure 2), studies and field surveys. The first Merapi Hazard Map was made in 1978 and it then was revised in 2002. The map shows three hazard zones: Hazard Zone I (HZ I), Hazard Zone II (HZ II), and Hazard Zone III (HZ III). The most dangerous zone is HZ III. There are two sub-areas in this zone: the directly affected area (DA) and the indirectly affected area (IA) by the eruptions. Subvillages in HZ III were then identified by the government of Sleman District (Bappenas and BNPB, 2011: 73). These subvillages were also classified into two categories with two classes: totally affected and partly affected.

Based on the hazard zone classification, DA is clearly the most dangerous area. The development policies of DA are as follows: (1) this area is not recommended for human settlements; (2) this area is highly recommended for forestry, conservation area and eco-tourism; (3) infrastructures in this area are provided to support conservation area, eco-tourism and disaster mitigation. The zero growth policy also applies to Zone IA. Zone II is planned to become an area for limited expansion with a strict land use control. Zone I is safer than Zone II, but the risk of lava flood cannot be excluded.
This Merapi Hazard Map basically represents risk assessments by the experts on the Merapi volcano. We can match the list of subvillages in the HZ III with the place of residence of respondents before the eruption to find out whether respondents’ location are in HZ III. Some studies note that the villagers did not follow up the relocation policy after an eruption exposed their village. This response may be explained by locational advantages of the Merapi area (for instance advantages in crop cultivation, livestock rearing and sand mining) along with the traditional belief system (Dove, 2008; Donovan, 2010; Bappenas and BNPB, 2011).

The “matching procedure” used here is quite simple. For instance, the survey reports that the pre-eruption place of residence of the respondent A was in subvillage X. If the subvillage X appeared in the list of subvillages in HZ III provided by the government, we then treat the pre-eruption location of this respondent as in HZ III. The same procedure applies to DA, the most dangerous area in HZ III. We can attribute the label experts’ risk assessment to these matched locations since it is based on the Merapi Hazard Map produced by the CVGHM, March 2011, taken from Bappenas and BNPB (2011: 62); both directly and indirectly areas are in Hazard Zone III.4
by experts. We then also identify the respondents’ perception on disaster risk by using their response on the question: “Was your place of residence in disaster-prone area (for any natural disasters)?”. It is important to note that this question assesses the former place of residence or the place of residence before the 2010 eruption. Therefore, the response to this question may indicate the respondents’ disaster risk perception over their former place of residence.

### Table 1. Pre-eruption place of residence: experts’ assessment and respondents’ perception

<table>
<thead>
<tr>
<th>Assessment of Risk by using the Hazard Map (Disaster-prone area)</th>
<th>Respondents’ Perception of Risk (Disaster-prone area)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>No</td>
</tr>
</tbody>
</table>

#### Panel A: Cangkringan and Prambanan (n=271)

**Hazard Zone III:**

- **No**  
  - 58  
  - 9  
  - Total 67
- **Yes**  
  - 38  
  - 166  
  - Total 204

- **Matching rate (%)**  
  - 82.66

**Directly Affected Area:**

- **No**  
  - 67  
  - 34  
  - Total 101
- **Yes**  
  - 29  
  - 141  
  - Total 170

- **Matching rate (%)**  
  - 76.75

#### Panel B: Cangkringan (n=211)

**Hazard Zone III:**

- **No**  
  - 0  
  - 7  
  - Total 7
- **Yes**  
  - 38  
  - 166  
  - Total 204

- **Matching rate (%)**  
  - 78.67

**Directly Affected Area:**

- **No**  
  - 9  
  - 32  
  - Total 41
- **Yes**  
  - 29  
  - 141  
  - Total 170

- **Matching rate (%)**  
  - 71.09

**Notes:** This table is produced based on the household survey 2012. The matching rate is the percentage of total matched responses (in shadowed cells) per total respondents for the respective zone.

The results for both perspectives are presented in Table 1 for all respondents (Panel A) and for the Cangkringan subdistrict only (Panel B). Panel A shows that the expert’s risk assessments based on the Merapi Hazard Map are in line with respondent’s risk perception. The matching rate of risk perceptions is 83 per cent and 77 per cent for HZ III and DA, respectively. We also provide a cross-tabulation for respondents from the Cangkringan
subdistrict only since this is close to the source of volcanic risk. The results are presented in Panel B. As we can see, there are no significant differences in the matching rates between Panel B and Panel A.

Taking all together, the figures suggest that respondent’s perceptions of risks are in line with the expert’s perceptions of risk. Explaining the determinants of a respondents’ perception of risk is not our main interest. However, we expect that this perception will positively influence respondents’ decision to participate in disaster microinsurance. We assume that people who face natural disaster risks may be aware of important strategies of which disaster microinsurance is a promising option. Disaster microinsurance has several benefits: reducing vulnerability, a better position to cope with risk, protecting living standards and complementing any social security system (Heydel et al., 2009).

IV. Empirical Framework

It is important to underline that disaster microinsurance has not existed yet in our study area. One may argue that it is not possible to conduct a study under this circumstance because there is no insurance product meaning that there is no information about details of insurance (design, coverage, rate of premium, and so forth). The modalities of the insurance offered are likely to have an impact on the participation rate. But since the aim of this study is to find out some early evidence on this prospective counter-measure of natural disaster risks, we argue that it is valuable to investigate a hypothetical disaster microinsurance. In their study on Bangladesh, Akter et al. (2008) also used this approach. In their study on producers’ demand for a public-private crop insurance program in the Netherlands, Van Asseldonk et al. (2002) also applied a hypothetical insurance scheme.

To note, the section of the disaster microinsurance in the questionnaire has been designed to deal with the above problems. This section begins with a description about insurance in general, followed by a question with four possible answers to assess
respondents’ knowledge about insurance. A description of disaster microinsurance is then presented. These steps are implemented in the survey in order to minimise measurement errors. The surveyors explained what insurance is and reexplained it for respondents who still did not understand the explanation. The meaning of disaster microinsurance is also explained to respondents before the surveyors offered further questions. Since it is microinsurance for natural disaster risks, the survey also assessed the main concern of respondents based on their experiences with the 2010 Merapi eruption by choosing one of six possible options of the impacts of this eruption (house damaged, crop damaged, cattle killed or loss, health problems, unemployment problems and children’s education). This strategy is used to link questions on insurance to the context of disaster risks.

The survey then delivered a question whether respondents would be interested in participating in disaster microinsurance in view of their experiences with the eruption of Merapi volcano in 2010 as well as their financial situation. This question is preceded by a question whether respondents need insurance as a protection for natural disaster risks. The intuition of using this step is that only those respondents who need insurance might be interested in participating in disaster microinsurance. In other word, the first question provides a general framework for the second question.

In an ideal situation (for instance a perfect insurance market) one may assume that questions about the need for insurance and the interest to participate in disaster microinsurance can only be asked to respondents who have a perfect insurance knowledge. But it is difficult to expect that respondents have a perfect insurance knowledge, even in developed countries. For instance, McCormack et al. (2002) find that a sizeable proportion of Medicare beneficiaries in Kansas City are unaware about some basic and critical aspects related to cost, coverage, and supplemental insurance options although they are already informed about some elements of the Medicare program. Therefore, they argue that development of meaningful information for beneficiaries is important.
Logit Model

The following logit model will be used:

$$\text{Logit}(P_I) = \beta_0 + \beta_1 R_i + \sum \beta_k X_{ki} + \varepsilon_i$$  \hspace{1cm} (1)

where $P_I$ is the probability of a respondent to be interested in participating in a (hypothetical) disaster microinsurance, $R$ represents respondent’s perception of eruption risk (a dummy variable), $X_k$ represents a range of $k$ other relevant variables and $\varepsilon$ is the error term.

The first relevant variable is insurance literacy (see, for instance, Wang et al., 2012). We construct a simple index representing respondent’s general knowledge about insurance, abbreviated as $IL$, based on three elements. These elements are (1) whether a respondent was able to choose the correct answer to a question about the basic description of insurance, (2) whether a respondent was able to mention at least one name of an insurance company, and (3) whether a respondent had an experience with insurance products. For the first element, the correct answer is: “A policy holder will have to pay a certain amount of money regularly to insurance in order to anticipate unexpected events in the future.” These three elements are in binary values and the index is a mean average of them. We expect that this insurance literacy index will have a positive influence on the household decision to participate in disaster microinsurance.

Regarding other explanatory variables, first, we include four variables representing individual household risks. The first variable is a ratio of number of household members under the age of 18 to household size ($CR$). Households with children are usually assumed to be more vulnerable to risks and therefore they may have a higher interest to participate in disaster microinsurance (see, Alam et al., 2011). The next variable ($LA$) is a dummy variable for liquid assets (cash, savings or deposit in bank accounts) that may play a role as a self-insurance (Carter et al., 2007; Skoufias, 2003; Heydel et al., 2009). For households with liquid assets, participating in disaster insurance is less urgent. Access to ex-post disaster relief is
another indicator of individual risk as indicated by Akter et al. (2008), Wang et al. (2012) and Longwell (2013). Van Asseldonk et al. (2002) also concludes that the crop producer’s belief in disaster relief lowers the potential degree of participation in the insurance scheme. In this study we use a dummy variable indicating whether a household had access to disaster relief for housing ($HR$). Access to disaster relief is expected to have a negative influence on the participation in disaster microinsurance.

We also employ smoking behaviour as a proxy for health status. This variable is the ratio of the length of the smoking consumption period to the age of the head of the household ($SR$). A lower health (higher health risk) may have a negative effect on household expenditure by increasing health costs due to illness and reducing household capacity to earn income when the household head as the main income earner gets health problems. Therefore, a higher smoking indicator index indicates a higher health risk, which might positively influence household participation in disaster microinsurance.

Social networks have been argued to influence participation in insurance. For instance, respondents who have a close relationship with their relatives will have an opportunity to get informal support. Hence, social networks may reduce the propensity of a household to participate in disaster insurance. But, with regard to access to ex-post disaster relief, we may expect that respondents who have relatives who are known as village elites (like members of the village government), will have a better access to this relief and will negatively respond to disaster microinsurance. In his study on reconstruction funds in Fiji, Takasaki (2011) finds that traditional kin elites who have power, such as the chief’s clans, receive benefits earlier than others in recipient villages. Regarding the role of local associations in the specific context of our study, we use a dummy variable of membership of two types of associations: farmer associations ($FA$) and breeder associations ($BA$) since agriculture and livestock are important activities for villagers. Membership in these local
associations may reduce the likelihood of a household to participate in disaster insurance (Gine et al., 2011).

Finally, we include an income variable to detect the influence of households’ ability to pay their monthly fee for disaster microinsurance (for instance, Akter et al. 2008). But it is also possible that higher income negatively affects participation in this insurance when higher income reflects the capacity to self insure in a similar way as liquid assets. The income variable ($PYM$) is measured by using monthly income per person (total household monthly income divided by household size).\(^5\)

Model (1) can be rewritten by plugging in all potential explanatory variables as already discussed above. Then our final logit model is as follows:

$$
\text{Logit}(PI_i) = \beta_0 + \beta_1 R_i + \beta_2 IL_i + \beta_3 CR_i + \beta_4 LA_i + \beta_5 SR_i + \beta_6 RO_i + \\
\beta_7 FA_i + \beta_8 BA_i + \beta_9 \log PYM_i + \epsilon_i
$$

Model (1) can be rewritten by plugging in all potential explanatory variables as already discussed above. Then our final logit model is as follows:

$$
\text{Logit}(PI_i) = \beta_0 + \beta_1 R_i + \beta_2 IL_i + \beta_3 CR_i + \beta_4 LA_i + \beta_5 SR_i + \beta_6 RO_i + \\
\beta_7 FA_i + \beta_8 BA_i + \beta_9 \log PYM_i + \epsilon_i
$$

\textbf{Table 2. Summary statistics}

<table>
<thead>
<tr>
<th>Notation</th>
<th>Variable Description</th>
<th>Mean</th>
<th>Std. Dev.</th>
<th>Min</th>
<th>Max</th>
</tr>
</thead>
<tbody>
<tr>
<td>$PI$</td>
<td>Interested to participate in disaster microinsurance (D)</td>
<td>0.28</td>
<td>0.45</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>$R$</td>
<td>Place of residence was in disaster-prone area (D)</td>
<td>0.65</td>
<td>0.48</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>$IL$</td>
<td>Mean average of three dummy variables for insurance knowledge (see text)</td>
<td>0.25</td>
<td>0.33</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>$CR$</td>
<td>Ratio of the number of children below 18 years to household size</td>
<td>0.22</td>
<td>0.22</td>
<td>0</td>
<td>0.75</td>
</tr>
<tr>
<td>$LA$</td>
<td>Household has liquid assets (D)</td>
<td>0.69</td>
<td>0.46</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>$HR$</td>
<td>Household received house relief (D)</td>
<td>0.35</td>
<td>0.48</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>$SR$</td>
<td>Ratio of the length of smoking consumption period of household head to her/his age</td>
<td>0.21</td>
<td>0.26</td>
<td>0</td>
<td>0.88</td>
</tr>
<tr>
<td>$RO$</td>
<td>Household has relatives who are village officials</td>
<td>0.11</td>
<td>0.31</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>$BA$</td>
<td>Member of breeder associations (D)</td>
<td>0.36</td>
<td>0.48</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>$logPYM$</td>
<td>log of household monthly income/person (Rp)</td>
<td>5.40</td>
<td>0.60</td>
<td>3</td>
<td>6.77</td>
</tr>
</tbody>
</table>

\textit{Notes:} (D) indicates a binary variable (1=yes; 0=no)
Table 2 shows that 28 per cent of the respondents stated that they were interested to participate in the hypothetical disaster microinsurance. Since an actual disaster microinsurance has not developed for the Merapi eruption risk, this figure must be interpreted with care. It addresses a first hurdle that people pass before they actually take the decision to take up this type of insurance scheme. The final decision will depend on obvious factors such as the insurance premium, the amount of money paid by the insurance company in case of a disaster, the details of the insurance contracts, and so forth. Thus, it is plausible that the actual response to an insurance scheme like discussed here will be smaller than the initial response we observe in the present study, or less than 28 per cent.

V. Results and Discussion

Baseline results

To be useful, the output of nonlinear models must be converted into marginal effects so that we can discuss the impact of independent variables on the dependent variable (see, Angrist and Pischke, 2008, p. 76). We present average marginal effects (AME, marginal effect at the mean of each independent variable) of the baseline estimation in Table 3 (column baseline).

The respondent’s perception of natural disaster risks shows statistically significant effects on the probability of a respondent to be interested in participating in disaster microinsurance. Its marginal effect suggests that people’s perception of volcanic risk has a positive influence on the probability to participate in a disaster microinsurance scheme. The positive influence of risk perception also indicates that respondents are aware of important strategies of which disaster microinsurance is a promising option.

The result of the baseline model also shows that insurance literacy has a large influence on the probability of a respondent to be interested in participating in disaster microinsurance. As we can see, the marginal effect of the insurance literacy variable is 0.30 and it is the largest marginal effect among the independent variables in the baseline model.
This marginal effect indicates that a one unit increase in the insurance literacy index increases the probability of a respondent to participate in disaster microinsurance with 30 per cent. This result confirms the important role of insurance literacy in affecting the probability of households to be interested in participating in a disaster microinsurance (see, for instance, Wang et al., 2012). The large effect of insurance literacy also implies that it is important to develop insurance literacy in order to increase the rate of participation in disaster microinsurance.

Table 3. Estimation results (logit model)

<table>
<thead>
<tr>
<th>Variable</th>
<th>Baseline</th>
<th>IV-A</th>
<th>IV-B</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>AME</td>
<td>SE</td>
<td>P&gt;z</td>
</tr>
<tr>
<td>Eruption risk variables:</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Respondent's risk perception-1</td>
<td>0.10</td>
<td>0.06</td>
<td>0.09</td>
</tr>
<tr>
<td>Respondent's risk perception-2</td>
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Notes: SE is Robust Standard Error; AME is Average Marginal Effect (evaluated at the mean of each independent variable).

As expected, access to house relief negatively influences respondent’s interest to participate in disaster microinsurance. Its marginal effect suggests that the probability of
respondents who have access to house relief is 12 per cent lower than for the respondents who do not have access to such relief. This result is not surprising since housing damage has been identified by respondents as their main concern related to the impact of the Merapi eruption in 2010 (see, Figure 4). This result is also in line with other studies (Akter et al., 2008; Wang et al., 2012; Longwell, 2013).

We also find that the smoking risk variable positively affects respondent’s probability to be interested in participating in disaster microinsurance. It should also be noted that health problems are the second important impact of the Merapi eruption (see, Figure 4). A higher value of the smoking indicator indicates a higher health risk and volcanic eruptions increase this health risk. The marginal effect of this variable suggests that a one unit increase of smoking risk of the head of household increases the probability of a respondent to participate in disaster microinsurance with 17 per cent.

![Figure 4. Main concerns related to impacts the eruption](image)

With regard to the membership in local associations, the result of the baseline model shows that for respondents who are member of farmer associations, the probability to be interested in participating in disaster microinsurance decreases by 11 per cent. The negative effect of
membership in farmer associations supports the argument that this local association plays an alternative role for villagers to cope with the impact of volcanic eruption. This causes disaster microinsurance to be less important. The farmer association may also play a role as a channel of delivery of disaster relief specifically to recover farming activities in this area. But a different role is found for membership in breeder associations. Being a member of a breeder association increases the probability to be interested in participating in disaster microinsurance by 11 per cent. One possible reason for this difference is that breeding activities—including the rearing of dairy cows—have relatively important role compared to traditional farming for villagers in the Merapi area. It is also possible that breeder associations are better organised than the farmer associations.

Taking all together, the findings suggest that the most important variables that affect respondents’ interest to participate in disaster microinsurance after evaluating their financial situation and their experiences with the eruption, are their perception of natural disaster risks, insurance literacy, house relief, smoking risk, and local social networks (memberships in farmer and breeder associations). Since the financial situation has been considered when respondents answered the question whether they are interested in participating in disaster microinsurance, it is not surprising that both income and liquid assets variables have no statistically significant coefficients. The negative sign of the coefficients indicates that respondents with a better financial position are less inclined to participate in disaster microinsurance.

Robustness checks

We have presented the results of our baseline model. Two issues remain which are important to be investigated to check whether the baseline result is robust. The first issue is related to variable of perception of risks, and the second one is the issue of insurance literacy. Both issues concern the measurement error of these two important independent
variables that can affect the estimates of the baseline model. Although the field survey has been carefully conducted, one may suspect that there is still a problem of measurement errors. We use instrumental equations in investigating these two issues on the robustness of our baseline specification. The results are also presented in Table 3.

As already discussed in Section 3, respondents’ perception of natural disaster risks refers to their former place of residence. However, since the survey was conducted after the eruption in 2010, it is possible that the responses have been influenced by respondents’ experiences with past natural shocks. Some studies also argue that past shock may affect the perception of risk (see, Brody et al., 2004; Brody et al., 2008; Brilly and Polic, 2005; Siegrist and Gutscher, 2006). In the present study, the past natural shock is the Merapi eruption in 2010. We instrument the respondent’s perception of risk with a dummy variable that represents whether respondent’s place of residence was affected by the eruption. This dummy variable is based on the households’ response to this question: "Was your place of residence affected by the eruption in 2010?". The first-stage estimation result for respondents’ perception of natural disaster risk is provided in Appendix A1 and the results of an instrumental variables equation is presented in column IV-A in Table 3. This specification (IV-A) shows that marginal effect of respondent’s perception of disaster risks is similar to the effect as found in the baseline model. We use a Hausman endogeneity test to compare the estimates of respondents’ perception of risk since these estimates are our parameter of interest (see, Wooldridge, 2002, p. 120). The Hausman $t$ statistic is calculated by using this formula: $(\beta_{1,\text{baseline}} - \beta_{1,\text{IV}}) / \sqrt{\text{se}(\beta_{1,\text{IV}})^2 - \text{se}(\beta_{1,\text{baseline}})^2}$\textsuperscript{1/2}. The 5 per cent critical value is 3.84 [chi-squared value($d$=1)]. The Hausman $t$ statistic is $-0.24$ indicating that the impact of respondent’s perception of risks is not statistically different from the impact in the baseline model. This test result supports the robustness of the baseline estimation.

With regard to insurance literacy, we have three potential variables to be used as instrumental variable for the insurance literacy index. The first is a dummy variable
indicating if the household uses and is familiar with mobile phone and/or internet. The reason for using this variable is that it may indicate the ability of a respondent to search or to find information about insurance. As indicated by McCormack et al. (2002), knowledge about some basic elements of insurance can be increased if there is a flow of information on insurance and the policy holders or potential policy holders are also aware of this information. The other candidates are two variables reflecting the mental health status of respondents. The first is “self concentration” and the second one is “self confidence”. Both are in ordinal values from 1 (the lowest) to 4 (the highest). The idea for using these variables is that if the respondents stated that they were able to focus on anything and they had a good self confidence during the last week before the survey was conducted, we expect that these respondents could minimise errors in selecting the best answer to a question about a general definition of insurance.

The results for the first stage estimation are presented in Appendix A2. We find that dummy variable of telecommunication consistently shows a statistically significant impact on insurance literacy. We then use this telecommunication variable to instrument the insurance literacy index and the result for the IV model is presented in column IV-B in Table 3. As we can see, the marginal effect of the insurance literacy index is almost four times as large as the marginal effect in the baseline estimation. This supports the role of telecommunication facilities on insurance literacy. However, the Hausman $t$ statistic ($-2.55$) is smaller than the 5 per cent critical value (3.84) suggesting that the impact of insurance literacy in the IV-B is not statistically different from the impact in the baseline model. In other words, the test again confirms the robustness of the baseline estimation.

**A simulation exercise**

The findings show that perception of natural disaster risk as our variable of interest influences participation in disaster microinsurance. Besides, insurance literacy also
affects respondents’ decision to participate in this microinsurance. Since insurance literacy has a large impact, we use this variable to simulate the probability of respondents to be interested in participating in a disaster microinsurance. Such a simulation is useful to identify policy implications with regard to the opportunity to establish this type of insurance for people living in disaster-prone areas.

Figure 5 provides the observed probability and three simulated probabilities. The observed probability is 28 per cent and is based on the survey result. We use the baseline result of the logit model (the fitted model) to calculate simulated probabilities. In the simulation 1, all independent variables used are at their median values. Evaluating Table 3 we find that the standard of deviation of some independent variables is larger than their mean. In the case where data are skewed, the use of the median value is better than the mean. Using the median value of insurance literacy the simulated probability is 38 per cent (Simulation-1). We then calculate the simulated probabilities for the extreme values of

![Figure 5](image-url)
insurance literacy (0 and 1). As can be seen in Figure 5, the simulated probability for these two situations are 21 per cent (Simulation-2) and 59 per cent (Simulation-3).

The difference between the simulated probability of Simulation-2 and of Simulation-3 is large (38%). Therefore, our simulations confirm that insurance literacy plays an important role in determining the rate of participation in disaster microinsurance. The policy implication of this finding is that insurance literacy of potential participants should also be developed along with designing and implementing a relevant microinsurance for people living in a specific disaster-prone area like the Merapi volcano.

VI. Conclusion

The main objective of this paper is to investigate the influence of perception of natural disaster risks on the probability of a respondent to be interested in participating in a hypothetical disaster microinsurance. We use household data to assess a specific disaster risk, notably the risk of an eruption of the Mount Merapi in Java.

On the basis of a logit model we find that respondent’s perception of natural disaster risks positively influences the respondents interest in participating in disaster microinsurance. This result also suggests that households living in disaster prone areas are aware of the risks of their location. This awareness then increases the probability of a respondent to be interested in participating in disaster microinsurance as a tool to minimise the impact of natural disasters. It implies that local people’s perception of natural disaster risks should be taken into account in designing and proposing a specific insurance for people living in vulnerable areas to natural disasters.

We also find that insurance literacy has a strong positive relationship with the respondent’s interest to participate in disaster microinsurance. The large effect of insurance literacy implies that it is important to develop insurance literacy to increase the rate of participation in disaster microinsurance. Whether respondents showed an interest to
participate in disaster microinsurance is also determined by their access to house relief, membership in local associations and health risk that is related to smoking behaviour. We conclude that these variables should also be considered in studying and implementing microinsurance scheme as a formal coping strategy for people living in disaster-prone areas.

Finally, this study uses household level data to investigate the impact of a specific natural disaster, viz. a volcanic eruption of Mount Merapi. It should be noted that a volcanic eruption is only one of the natural disasters in Indonesia that can negatively affect the lives of local people. Furthermore, we argue that natural disaster risks and their influence on participation in disaster microinsurance are closely related to the specific location.

Notes
1. A microinsurance against floods was developed by Wahana Tata (an insurance company) and this product was established in partnership with GTZ (the German development agency) and Munich Re (a reinsurer company). This product is designed as livelihood coverage against floods. The price of this product (card) is IDR 50,000 and the card holder will receive IDR 250,000 when flooding in Jakarta reaches a certain predefined level (above 950 centimetres at the Manggarai Water Gate in Jakarta). Only about 500 cards were sold and this weak demand was one of the reasons to end this pilot program in 2010 (see, World Bank 2011: 27).
2. We use data at the household level provided by a joint team of Gadjah Mada University of Indonesia and the Indonesia Project of ANU Canberra who conducted a survey in 2012 to investigate the impact of the eruption in 2010 on the welfare of people living around the volcano, especially the people who were affected by the eruption. Hence, the disaster microinsurance section is not its main aim. The survey did not offer insights from a specific operational disaster microinsurance scheme for people living on the slope of Mount Merapi since such microinsurance scheme has not been developed yet. The lists of households in the Cangkringan and the Prambanan subdistricts based on the 2010 Population Census were provided by the local government offices. The Cangkringan subdistrict is situated closest to the summit of Merapi volcano while the Prambanan subdistrict is quite far from the volcano. Therefore the surveyed households in Prambanan can be partly used as the control group. According to the official data, there are 8,727 households in the Cangkringan subdistrict in 2010 and about 36 per cent of them were affected by the Merapi eruption in 2010. Most affected households in the Sleman district were in the subdistrict of Cangkringan. There are five villages in the Cangkringan subdistrict, but the team focused on three villages: Glagaharjo, Kepuharjo and Umbuharjo. Most inhabitants of these three villages were evacuated during the volcanic crisis in 2010 (see, Figure 2 and 3). A random sampling has been used to create 300 samples (about 3.4% of the sampling frame for the Cangkringan subdistrict); 20 of them are used for testing the questionnaire and the rest are for an interview. In the
Prambanan subdistricts, 80 households are randomly selected from 191 households in the Jamusan subvillage in the Bokoharjo village (see Figure 2). The total number of households that finally could be traced for an interview are 276 households (216 in the Cangkringan and 60 in the Prambanan subdistrict). We excluded five respondents from the district of Cangkringan due to missing information and outlier for our relevant variables resulting in 271 observations for empirical estimations.


4. This map has low resolution. For a high resolution map see http://geospasial.bnpb.go.id/wp-content/uploads/2011/05/2011-04-29_KRB_Area_Terdampak_Langsung_Merapi.pdf

5. Seven respondents reported zero income. Since we use the log-income, the zero income of these seven respondents are imputed with the lowest income in the sample. This treatment allow us to keep these seven observations, instead of excluding them from the sample.

6. Brody et al. (2008) include property damages in estimating risk perception. They find no statistically significant influence. Meanwhile, Wang et al. (2012) argue that experience with significant losses of houses is a dominant subjective factor affecting people’s participation in disaster insurance.

Appendix

A1 First stage result for perception of natural disaster risk

| Variable          | Coef. | SE  | P>|t|
|-------------------|-------|-----|-----|
| House is affected | 2.79  | 0.32| 0.00|
| Constant          | −1.18 | 0.26| 0.00|
| Wald chi squared(1) | 74.40 |     |     |
| Prob > chi squared | 0.00  |     |     |
| Pseudo R-squared  | 0.26  |     |     |
| Log pseudolikelihood | −130 |     |     |

Note: SE is robust Standard Error

A2 First stage result for insurance literacy index

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<tr>
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<td>Root MSE</td>
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Note: SE is robust Standard Error
References


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