

1 **Modeling runoff and erosion risk in a small steep cultivated watershed using different data**
2 **sources: from on-site measurements to farmers' perceptions**

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26
27 **Abstract**

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29 This paper presents an approach to model runoff and erosion risk in a context of data scarcity,
30 whereas the majority of available models require large quantities of physical data that are
31 frequently not accessible. To overcome this problem, our approach uses different sources of data,
32 particularly on agricultural practices (tillage and land cover) and farmers' perceptions of runoff
33 and erosion. The model was developed on a small (5 ha) cultivated watershed characterized by

34 extreme conditions (slopes of up to 55%, extreme rainfall events) on the Merapi volcano in
35 Indonesia.

36 Runoff was modelled using two versions of STREAM. First, a lumped version was used to
37 determine the global parameters of the watershed. Second, a distributed version used three
38 parameters for the production of runoff (slope, land cover and roughness), a precise DEM, and
39 the position of waterways for runoff distribution. This information was derived from field
40 observations and interviews with farmers. Both surface runoff models accurately reproduced
41 runoff at the outlet. However, the distributed model (Nash-Sutcliffe=0.94) was more accurate
42 than the adjusted lumped model (N-S=0.85), especially for the smallest and biggest runoff
43 events, and produced accurate spatial distribution of runoff production and concentration.

44 Different types of erosion processes (landslides, linear inter-ridge erosion, linear erosion in main
45 waterways) were modelled as a combination of a hazard map (the spatial distribution of
46 runoff/infiltration volume provided by the distributed model), and a susceptibility map
47 combining slope, land cover and tillage, derived from in situ observations and interviews with
48 farmers. Each erosion risk map gives a spatial representation of the different erosion processes
49 including risk intensities and frequencies that were validated by the farmers and by in situ
50 observations. Maps of erosion risk confirmed the impact of the concentration of runoff, the high
51 susceptibility of long steep slopes, and revealed the critical role of tillage direction.

52 Calibrating and validating models using in situ measurements, observations and farmers'
53 perceptions made it possible to represent runoff and erosion risk despite the initial scarcity of
54 hydrological data. Even if the models mainly provided orders of magnitude and qualitative
55 information, they significantly improved our understanding of the watershed dynamics. In
56 addition, the information produced by such models is easy for farmers to use to manage runoff
57 and erosion by using appropriate agricultural practices.

59 1. Introduction

60 Soil erosion and surface runoff are frequent phenomena but their form, intensity, and effects on
61 agricultural land in tropical regions vary considerably (Randria-narijaona, 1983; Roose and
62 Ndayizigiye, 1997; Vezina et al., 2006). Modeling is one way to better understand these
63 processes. Runoff and erosion at the watershed scale can be modelled using different approaches.
64 Non-distributed models estimate both runoff and sediment yield but only at the outlet, while
65 distributed models represent their spatial distribution and account for watershed heterogeneity
66 (Dlamini et al., 2010) especially heterogeneity due to agricultural practices. Distributed models
67 require additional distributed data for their calibration, but simulated sediment yield is

68 nevertheless subject to significant error (López-Vicente et al., 2013). For model extension, the
69 size of the watershed is a key factor for agriculture practices (Valentin et al., 2008). As far as
70 temporal dynamics are concerned, event models of erosion can cope with the brief intense
71 production of runoff.

72 The impact of agricultural practices on erosion has mostly been studied at the plot scale and
73 mainly concerned sheet or inter-ridge erosion (DeLaune and Sij, 2012). At the scale of small
74 watersheds, runoff interconnects the different plots and its spatial distribution has major effects
75 on other forms of erosion (linear erosion, landslides). In small watersheds, distributed models are
76 consequently required and data collection also concerns farmers' practices (Barnaud et al.,
77 2005), which are difficult to measure quantitatively or to extrapolate.

78 The collection of data on topography, rainfall, soil properties and land cover, soil water content,
79 runoff flow, sediment yield, etc. is a major concern for model calibration and validation. In the
80 frequent situations when on-site measurements are lacking, one solution consists in using data or
81 empirical laws from similar situations (Evrart et al., 2009), which however, raises transposition
82 issues. Another solution consists in diversifying the sources of on-site data through quantitative
83 measurements, qualitative observations, and interviews with local people (Etienne, 2011). For
84 instance, farmers can provide useful information about a study site, their plots, and their farming
85 practices that can be compared with on-site observations and satellite images. In addition,
86 building models intended to be useful to stakeholders requires their involvement in the modeling
87 process (Furlan et al., 2012). The stakeholders should already be involved in identifying the
88 issues and in selecting the output form of the model, as well as in collecting data for model
89 calibration and validation. When this approach is used, the models will be more easily
90 appropriated by stakeholders and used as support for discussion and negotiation as appropriate
91 for the Panta Rhei decade « focus on hydrological systems as a changing interface between
92 environment and society » (Montanari et al., 2013),

93 The aim of the present study was to model runoff and erosion risks in a small steep cultivated
94 watershed located on the slope of the Merapi volcano (Java), where available input data is very
95 scarce. In these extreme topographic conditions, farmers perceive runoff and erosion via their
96 impact on agriculture, and try to deal with them using agricultural practices based on their own
97 experience and on traditional knowledge. The aim of our model was therefore to help the local
98 farmers improve management of runoff and erosion in the watershed.

100 **2. Material and method**

101

2.1 Study site

Java is located in a tropical area with high precipitation, and in a subduction area where volcanic reliefs are dominant. Most cultivable land is already cropped and extreme agriculture has taken over the steep slopes of volcanoes. Frequent intense rainfall events cause serious runoff, and erosion is thus a major concern for extreme agriculture (Turkelboom et al., 2008).

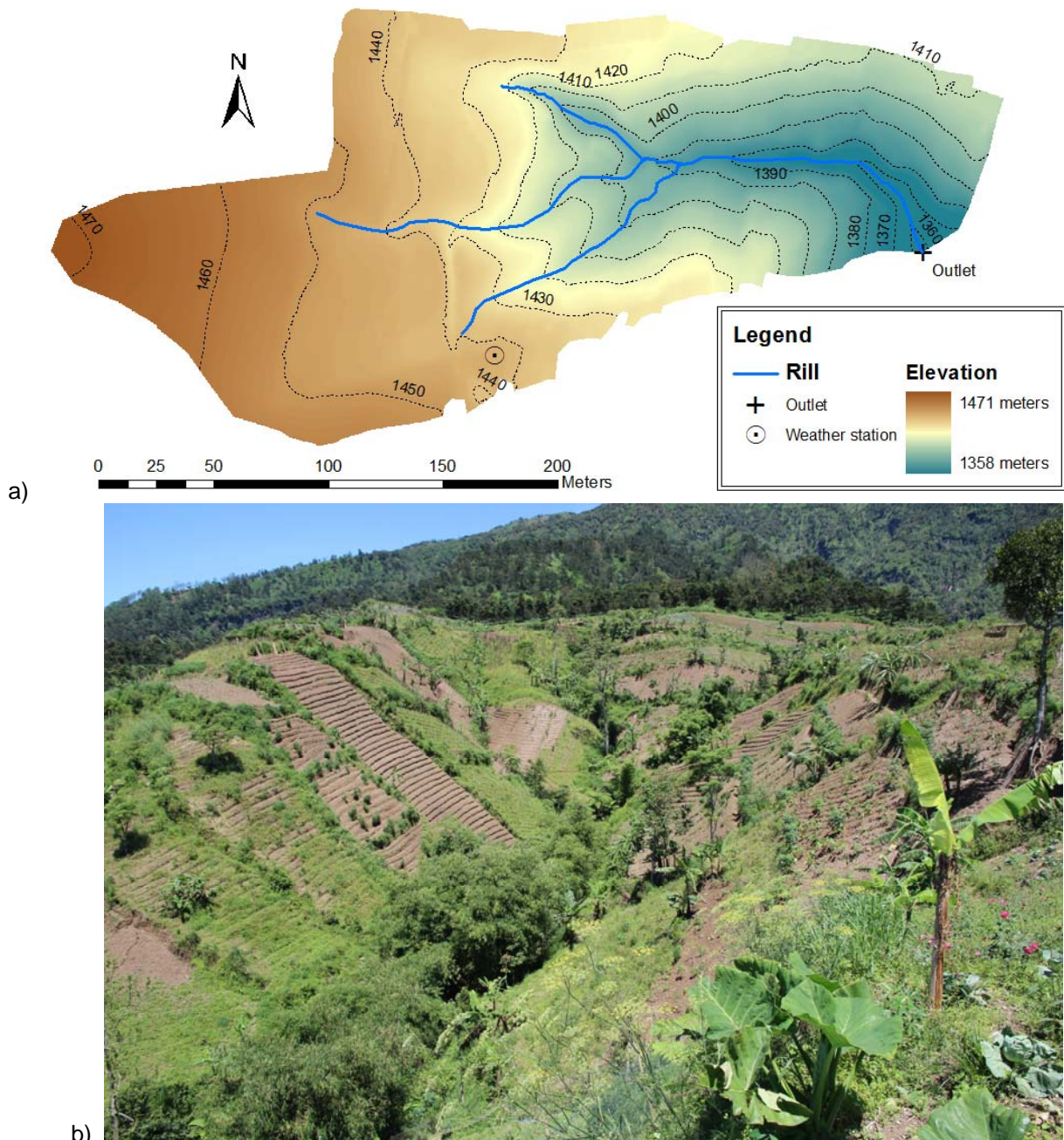


Figure 1: (a) Topographic map of the Gumuk basin and (b) photo taken from the south-east corner looking west.

The Gumuk watershed is an example of extreme agriculture. It is located on the east-south-eastern slope of the Merapi volcano. The coordinates of the outfall are 7° 32' 33.21203" S and 110° 29' 2.0486" E and an elevation of 1,471 m asl. The watershed (Fig. 1) is approximately 400 metres long and 150 metres wide and covers 4.5 hectares. The watershed is very steep: the average slope is 23° and 20% of the area has a slope steeper than 40°. The steepest slopes are concentrated in the centre of the watershed (Fig. 1a).

The watershed substratum is an andesitic lava flow covered by ash and pyroclastic deposits. Deposition, driven by the volcano activity, and erosion, driven by intense rainfall events, has shaped the geological structure of the watershed. Soils are andosols mainly composed of deposits with very low organic matter content. They are very rich in crystalline materials due to their young age.

Agriculture uses most of the watershed for cultivation on terraces whose steep slopes can reach 50° (Fig. 1b). The watershed comprises more than 80 plots ranging from 50 m² to 700 m² in size (Fig. 2). Most of the plots are ridged to give a preferential direction to the flow, and are drained by ditches. The land cover varies over the course of the year depending on the cropping system. Typically, maize is cultivated at a low planting density at the beginning of the rainy season (October to January) followed by market gardening of different vegetables in the same field (December to March). Tobacco is then cultivated from February until the end of the rainy season (mid-June) in almost all the plots and harvested during the first month of the dry season. No new crop is planted during the dry season, which lasts from June to October.

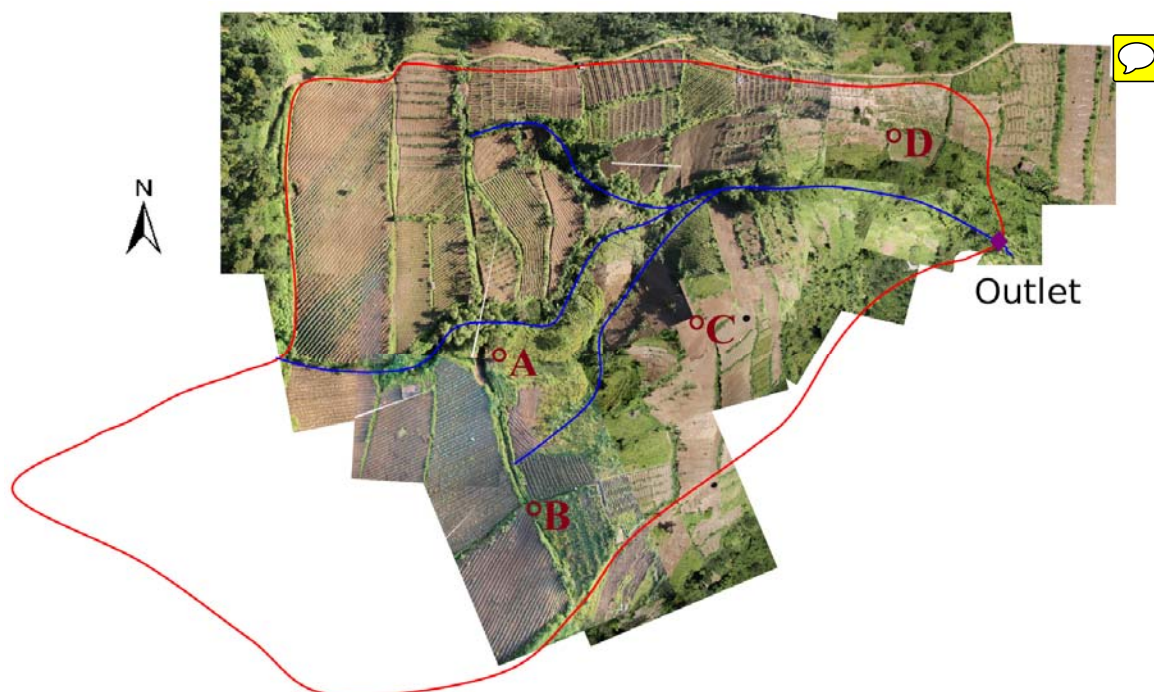


Figure 2: Assembly of aerial photographs. The basin boundaries are in red and the main hydrography in blue. Runoff was observed on the A, B, C and D locations.

In 2012, 1,700 mm of rain was measured at the local weather station. Only ten days of heavy rainfall accounted for 663 mm (38% of total rainfall) and only 12 mm of rain fell during the dry season. The data collection campaign ended in July 2011. The weather in 2012 and 2013 was compared to long term data from neighbouring weather stations. Weather stations at Boyolali (20 km from Gumuk, alt. 400 m), Yogyakarta (32 km from Gumuk, alt. 100 m) and Surakarta (40 km from Gumuk, alt. 100 m) are located at lower altitudes in the same valley. Due to the impact of relief on the weather, they are not representative of the study site. Farmers were questioned instead. They qualified the weather in 2012, especially the rain (total amount or extreme events), as a “typical”. Farmers cited the length of the rainy season, i.e., the exact starting and ending dates, as the main inter-annual weather variability. Indeed, the length of the rainy season determines the agricultural calendar and the water resource. Farmers described the 2012 rainy season as starting “on time”, and the 2013 rainy season as being one month longer, as it ended at the beginning of July whereas it usually ends at the beginning of June. Extreme rainfall events in 2012 were described as “typical” in terms of number and intensity. The rainfall events in 2012 produced runoff that triggered erosion, which, in turn, had a major impact on agricultural activities. Runoff and erosion were therefore also considered to be “typical” by the farmers.

2.2 Analysis of rainfall-runoff data

Rainfall was measured in a receptacle with 0.2 mm capacity at one-minute intervals by a pluviometer that was an integral part of an automatic weather station (Vantage Pro2 - Davis). The weather station was located on the top of the watershed (see fig.1a). Runoff was measured at the outlet by measuring the level of water at a gauging station that had been converted to measure discharge using rating tables calibrated in the laboratory.

The aim of data analysis was to identify the rainfall events that produced runoff. Monitoring began in September 2011 and ended in March 2013. Fifty-six days with runoff were counted at the outlet, but rainfall and runoff data were only both correct on 29 days. The start and end of a rainfall event were defined as a period of three minutes with rainfall intensity equal to or greater than 0.1 mm.h^{-1} (5 minutes moving average). Data analysis was limited to runoff produced by a single rainfall event. For these events, the antecedent precipitation index (API: Descroix et al., 2002) was calculated at a one-minute time step for a period of 24 h. The API was defined as

166 follows (Eq. 1) where t is the time in minutes before the beginning of the rainfall event and p_t is
 167 precipitation during this minute:

$$168 \quad API = \sum_{t=1}^{1440} p_t a^t \quad (\text{Eq. 1})$$

169 The recession factor (a) of API was determined as 0.9984023. It corresponds to 90% infiltration
 170 of rain 24 hours before the main rainfall event, due to the high hydraulic conductivity of soils.

171 The total amount of rain and the duration of the rainfall events were also calculated. The total
 172 volume of runoff associated with a given rainfall event corresponded to the output flow between
 173 the beginning and end of runoff.

174 175 **2.3 – Runoff modeling based on STREAM model**

176 The STREAM model (Souchère et al., 1998; Cerdan et al., 2002; Souchère et al., 2003) uses a
 177 raster approach to calculate the spatial distribution of runoff volume at the time scale of a rainfall
 178 event based on the hydrological process and expert rules. STREAM architecture is separated into
 179 two components: the production of runoff and the transport of the runoff water.

180 **Lumped version**

181 The lumped version (Fig. 3) considers the watershed to be homogeneous. We only focused on
 182 total runoff volume at the outlet using the characteristics of the rainfall event. This version
 183 therefore only used production components. The STREAM production component is based on
 184 physical processes instead of multivariate adaptive regression splines (Sharda et al., 2006). Rain
 185 fills the imbibition tank until it overflows; runoff corresponds to this overflow minus the
 186 infiltration volume, which is derived from the soil infiltration capacity and the duration of the
 187 rainfall event. The initial filling of the imbibition tank depends on the API. Four levels of
 188 imbibition capacity were distinguished corresponding to four categories of API.

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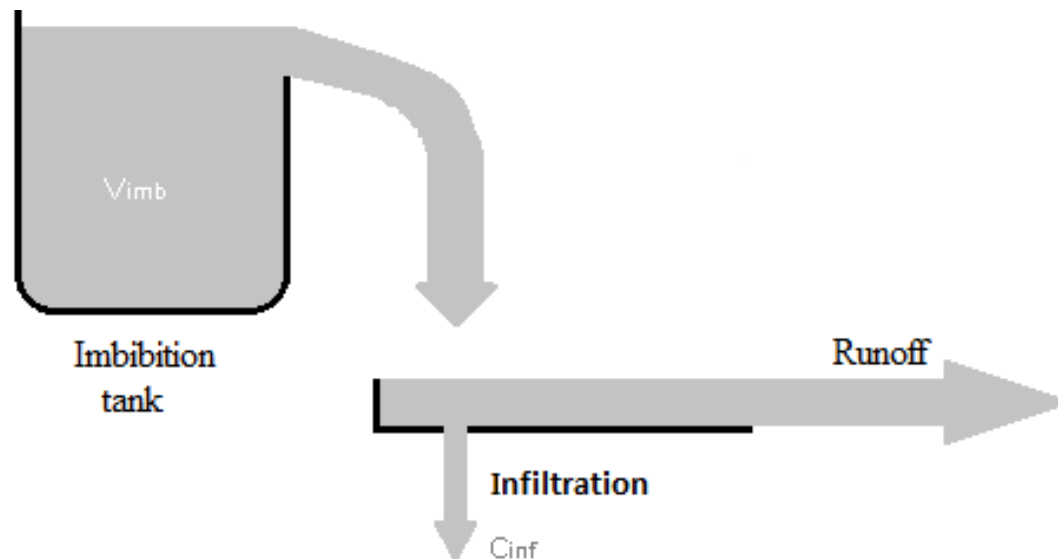





Figure 3: Runoff production in the non-distributed model 


Equation 2 represents the runoff volume production V (in cubic meters) per unit of surface area (S in thousands of square meters) as a result of the rain (R in mm) minus the imbibition volume (V_{imb} in mm) per unit surface area and the infiltration capacity (C_{inf} in $\text{mm}\cdot\text{h}^{-1}$) times the duration (D in hours) of the rainfall event.

$$\frac{V}{S} = R - V_{imb} - C_{inf} \times D \quad (\text{Eq. 2})$$

Development of a spatially distributed runoff model

~~Modelling runoff in Gumuk watershed aimed to spatially distribute the lumped version.~~ The raster grid is based on a digital elevation model of the watershed. In each cell, runoff production is determined with the same architecture as in the lumped model (see [fig. 3](#)). The runoff volume produced in each pixel is then transported to one of the eight neighbors following the steepest slope. The natural flow direction  can be modified by humans (through soil tillage and water channels) leading to preferential flow directions. In the absence of soil tillage, the flow direction  follows the natural slope. The map of flow accumulation enables evaluation of the runoff volume transiting each pixel.

Based on the characteristics of the Gumuk watershed (slope variability, size of plots, and water channels), a precise digital elevation model was built at a resolution of 25 centimetres using more than 2,000 elevation measurements with D-GPS and a Leica total station. Main waterways, water channels, terraces and border of plots were also mapped.

The runoff production component of STREAM was adapted to the specificity  of the study site. Based on field observations and interviews with the farmers, the slope, vegetal cover, soil

roughness and tillage direction were identified as main factors influencing runoff production and flow direction. These four characteristics were then used to build a decision table to determine the imbibition volume and infiltration capacity at plot scale: the values of these two parameters were estimated on the basis of the average imbibition volume and infiltration capacity determined by the lumped model. Data at plot scale (soil properties, slope, vegetal cover, soil roughness, tillage direction, waterways) resulting from observations or collected in interviews with farmers were used to build decision tables to determine the imbibition volume and infiltration capacity of each plot according to its characteristics. Imbibition volumes without antecedent rainfall were determined from the combination of the infiltration capacity and the antecedent rainfall. Assuming that the antecedent rainfall filled the imbibition tank to capacity and that filling was homogeneous across the watershed, the imbibitions volume for each pixel was determined using the same framework as for infiltration capacity.

2.4 Calibration and validation of the runoff models

The STREAM versions (lumped and distributed) of runoff were both calibrated with 13 rainfall-runoff events and validated with nine other events. The quality of the simulated runoff volumes was measured using the Nash-Sutcliffe coefficient (Nash and Sutcliffe, 1970) and the average quadratic error.

The map of runoff accumulation simulated with the distributed model version for each rainfall-runoff event was compared with field observations and information gathered in interviews with farmers about the location of runoff flow in each plot.

2.5 – Modeling erosion risks

Erosion is a major concern in agriculture in steep tropical landscapes [Vietnam (Vezina et al., 2006); Madagascar (Randria-narijaona, 1983); North Cameroon (Abbot et al., 2001); Rwanda (Roose and Ndayizigiye, 1997); Ethiopia (Nyssen et al., 2000); Thailand (Forsyth, 1994); Malaysia (Midmore et al., 1996)] and its spatial distribution is also a major issue. Erosion events, particularly those caused by storms in tropical areas, cause major land degradation [Turkelboom et al., (2008)]. As our watershed only produced runoff during extreme rainfall events, we focused on the subsequent erosion events.

Distributed numerical erosion models require multiple calibration and validation data and have difficulty representing the different erosion processes (hysteresis issues) (Giménez et al., 2012). There may also be significant errors in the location of erosion and in sediment yield simulated by distributed numerical erosion models (Jetten et al., 2003; López-Vicente et al., 2013). On the

other hand, the farmers' representation of erosion is more concerned with agricultural problems (sediment losses, destruction of the crop) at plot scale than at watershed scale. For these reasons, we decided to focus on the farmers' representations of erosion and on the location of erosion patterns within the watershed. Rather than computing sediment losses, we decided to model the different erosion risks and to build accurate maps of these risks in the watershed.

~~Modeling erosion risk is based on hazard vulnerability analysis (Turner et al., 2003; Prasannakumar et al., 2011).~~ Hazard and vulnerability are specific to each type of erosion the farmers described as being one of the major ones during the survey, namely: linear erosion in plots and waterways, and landslides. For both types of erosion risk, the vulnerability was constrained by the erosion susceptibility. Peak discharge determines linear erosion (Souchère et al., 2003), but the STREAM model only calculates runoff accumulation as the total volume of the event. For a small watershed (less than a few dozen hectares) with short rainfall-runoff events (a few hours or less), we assumed the runoff volume is a good indicator of the intensity of runoff flow.

For linear erosion (ephemeral rill intra-field and permanent gully extra-field), the hazard was evaluated as the runoff volume simulated by the STREAM model for a typical intense rainfall-runoff event. Simulated runoff volumes were classified in 3 or 5 (regular) classes to build a hazard map. Linear erosion susceptibility was derived from vegetal cover and slope (Souchère et al., 2003). The combination of these two characteristics at pixel scale was classified in four categories and resulted in a susceptibility map. The map of the risk of linear erosion was built using raster calculations (multiplication and categorization) combining the hazard and susceptibility maps.

For landslide, the hazard was evaluated as the combination of soil saturation (infiltrated volume) and overflow (runoff volume) (at plot scale). Landslide vulnerability related to vegetal cover, slope, and the difference in angle between the direction of tillage and of the slope. Landslide risks within the plot and at its border (where runoff has a major impact) were distinguished. The map of landside risk was built in the same way as linear erosion risk, by crossing the hazard and susceptibility maps.

2.6 – Soil properties, vegetal cover and tillage

The soil composition was analysed by the regional agricultural service BPTP, in Yogyakarta. Thirteen soil samples were taken in different cultivated plots located in the watershed. The organic carbon of each sample was measured by spectrometry, while other analyses focused on hydrological soil properties (primary porosity, effective porosity, ineffective porosity and

281 permeability). The national hydrological service IAHHI conducted a field experiment to
282 determine soil hydraulic conductivity using a permeameter disc. Measurements were taken in
283 seven plots, some cultivated, some ~~not~~ in March 2013.

284 The vegetal cover, soil tillage (type and azimuth direction) and roughness recorded during the
285 last two cropping seasons (mid-April 2013 to end-June 2013) were **precisely mapped** the
286 watershed. A total of 90 homogeneous plots were defined and these parameters were recorded in
287 a GIS. Vegetal cover was classified in four categories in accordance with agricultural criteria: (i)
288 “bare soil” with less than 20% of the surface covered by vegetation, corresponding to no crop or
289 a weak crop cover with no weed infestation; (ii) “medium vegetal cover” with 20% to 70% of the
290 ground covered by vegetation, corresponding to a weak crop with medium weed infestation or a
291 dense crop with no weeds; (iii) “high vegetal cover” with more than 70% of the ground covered
292 by vegetation, corresponding to a crop with high weed infestation; (iv) “very high vegetal cover”
293 where surface was completely covered by vegetation, corresponding to forest or uncultivated or
294 abandoned plots. Roughness was classified in five categories in accordance with agricultural
295 land preparation and height of ridges: (i) no roughness; (ii) natural small roughness (less than 5
296 cm); (iii) small roughness resulting from agricultural operations (5 to 10 cm); (iv) medium
297 roughness resulting from agricultural operations (more than 10 cm); (v) high roughness resulting
298 from agricultural operations (more than 10 cm and the presence of puddles after rainfall). Photos
299 were taken of the vegetal cover and of the roughness categories to discuss the soil surface and
300 vegetal cover in the plots with the farmers. The properties from 2012 to April 2013 were
301 assessed in interviews with the farmers.

303 **2.7 – Field observations on runoff and erosion**

304 Field surveys were conducted during rainfall events in May and June 2013 to assess the
305 contribution of slope, tillage direction, roughness and vegetal cover to runoff and erosion in the
306 plots. A sample of 18 plots with different slopes, tillage direction, roughness and vegetal cover
307 were chosen. Observations were made during rainfall events to compare runoff and erosion in
308 the field to ensure the different factors involved in runoff production (occurrence and intensity)
309 roughly corresponded and to identify the impact of each on erosion (change in the soil structure).
310 Observations were also made at four junctions (A, B, C, and D) in the upstream waterways (see
311 Fig. 2) in order to roughly estimate the runoff flow and the contribution of each upstream branch
312 during intense rainfall events. The survey was not possible downstream in the main waterway
313 because of the high runoff flow.

2.8 –Interviews with Farmers

The aim of the interviews with farmers was to gather two types of information. The first type concerned the farmer's perception of runoff and erosion and what kind of model outputs farmers would consider useful. The second type concerned their agricultural practices and land management. Individual semi-open interviews were carried out with a sample 19 farmers who cultivated 87% of the total cropped area. Each farmer was interviewed in his/her field.

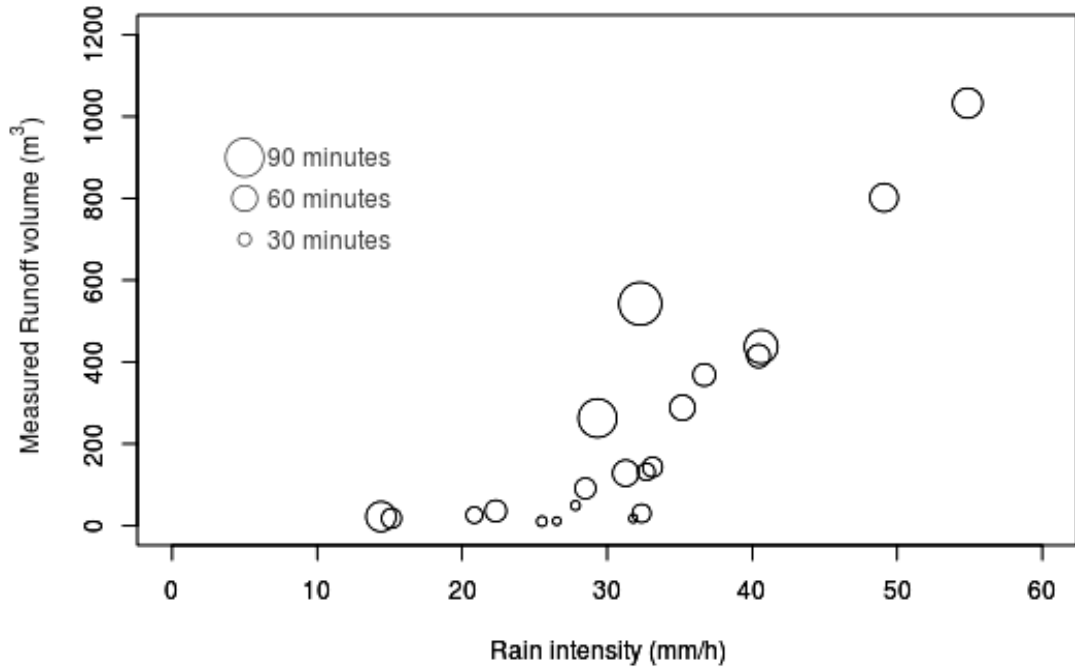
The first part of the interview concerned the farmer's perception of runoff and erosion, the different types of erosion, their impacts on farming activities, and their main triggers. For each type of erosion, the farmer showed us the exact place where erosion occurred in his/her field, its frequency and its intensity. The farmer also showed us the traces left by of erosion in the plot and qualified its intensity. In addition, the farmer located and qualitatively evaluated the concentration of runoff in the field during the 2012 and 2013 rainy seasons. The farmer's answers were compared with field observations. This information was mapped and used to validate the runoff and erosion risk models.

The second part of the interview was ~~directive and~~ concerned the farmer's crop and runoff management practices used in his/her plot in the 2012 and 2013 cropping seasons. The farmer described the tillage, weeding, agricultural calendar, weather and the differences between the current and the past cropping seasons. This information was compared with in situ observations and was used as model inputs.

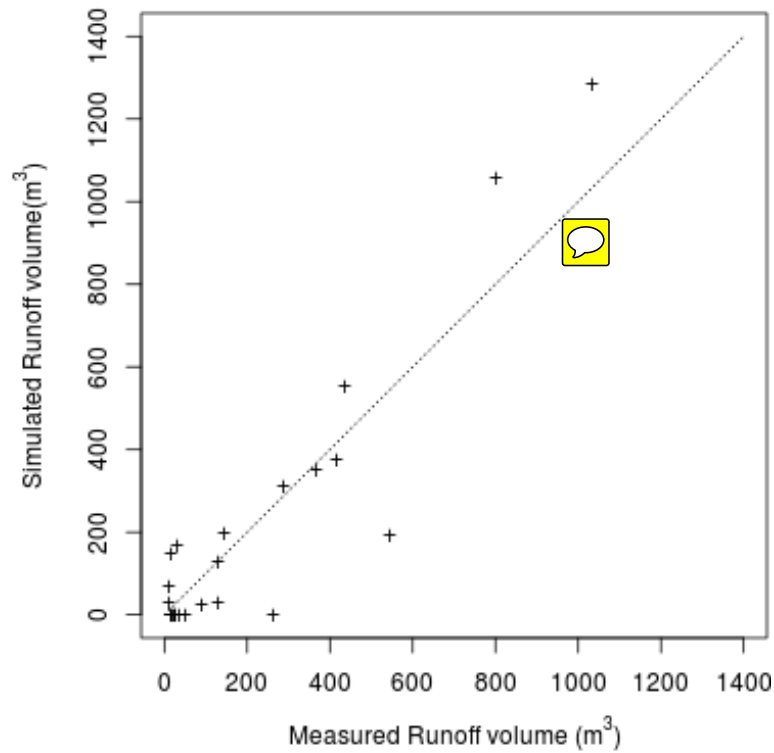
3 – Results

3.1 – General watershed dynamics and lumped version

The measured runoff volume produced by the Gumuk watershed ranged from 10 to 1,000 m³ depending on the intensity (rainfall amount and duration) of the rainfall event (Fig. 4-a). Rainfall events with an intensity of less than 28 mm.h⁻¹ produced no runoff. The volume of runoff produced by rainfall events whose intensity ranged from 28 to 43 mm.h⁻¹ depended on the duration: only rainfall events that lasted more than 40 minutes produced runoff, and major runoff volumes resulted from intense rainfall events that lasted for more than an hour.



a)



b)

Figure 4: (a) Measured runoff volume versus rainfall intensity and duration for the identified rainfall events, and (b) simulated versus measured runoff volume for identified events using the lumped version.

The lumped version of runoff correctly represented the general dynamics of the Gumuk watershed, especially for medium runoff volumes (between 100 and 500 m^3). Average

352 infiltration capacity and average imbibition volume without antecedent rainfall were determined
353 to be 30 mm.h⁻¹ and 10 mm, respectively. But the model clearly over-estimated high runoff
354 volumes and was highly inaccurate for small volumes (Fig. 4-b): simulated runoff volumes were
355 either under- or over-estimated. The average error was 122 m³ and the Nash-Sutcliffe coefficient
356 was 0.85.

358 **3.2 Decision tables in the distributed version and land cover scenarios**

359 The field survey showed that the natural slope combined with the slope in the direction of tillage,
360 and the vegetal cover were the main determining factors of runoff production. Indeed, soil
361 analyses revealed that the soil in the fields was homogeneous, i.e., andosol mainly composed of
362 ash deposits with little organic matter and no clay, characterized by high infiltration capacity and
363 20% effective porosity. However, at the bottom of the valley in the main waterway, the lava
364 substratum was visible at the surface and created in an impermeable a-strip of soil about 1 meter
365 in width and 100 meters in length. Five categories of average natural slope and slope in the
366 tillage direction (less than 10°, 10° to 15°, 15° to 20°, 20° to 25°, and more than 25°) were used
367 as a slope index. The vegetal cover was classified in four categories and surface roughness (that
368 included the height of ridges) in five categories.

369 The decision table used these three categorized factors to determine the infiltration capacity in
370 each plot. Values were distributed around the average infiltration capacity of the watershed (30
371 mm.h⁻¹) based on field observations and expert information (Table 1). Indeed, an increase in
372 slope or a decrease in roughness or in the vegetal cover logically increased runoff, thereby
373 reducing infiltration capacity.

Table 1: Decision table for the infiltration capacity (mm.h-1) according to the slope index, soil surface roughness, and plant cover.

Roughness	Plant cover	Slope index in degrees				
		0 - 10	10 - 15	15 - 20	20 - 25	>25
zero	Sparse crop with no weeds	35	25	15	5	5
	Crop with some weeds	35	30	20	10	5
	Crop with weed or grassland	40	35	25	15	10
	Highly vegetated	40	35	30	25	15
low (natural)	Sparse crop with no weeds	40	30	20	10	5
	Crop with some weeds	40	35	25	15	10
	Crop with weeds or grassland	40	35	30	20	15
	Highly vegetated	45	40	35	25	20
medium (small rill)	Few crop with no weeds	40	35	25	15	5
	Crop with some weeds	40	35	30	20	10
	Crop with weeds or grassland	45	40	35	25	20
	Highly vegetated	45	40	35	30	25
strong (dug rill)	Sparse crop with no weeds	40	35	25	15	10
	Crop with some weeds	40	35	30	20	10
	Crop with weeds or grassland	45	40	35	30	25
	Highly vegetated	50	45	40	35	30
strong and irregular	Sparse crop with no weeds	40	35	30	20	15
	Crop with some weeds	40	35	30	25	20
	Crop with weeds or grassland	45	45	40	35	30
	Highly vegetated	50	45	40	40	35

Imbibition volumes derived from infiltration capacities (Table 2). A plot with a high infiltration capacity also had a high imbibition volume, which decreased with antecedent rainfall. However the variability of imbibition volume was small (3 mm).

Table 2: Decision table for determining the imbibition volume (mm) according to categories of infiltration capacity and API.

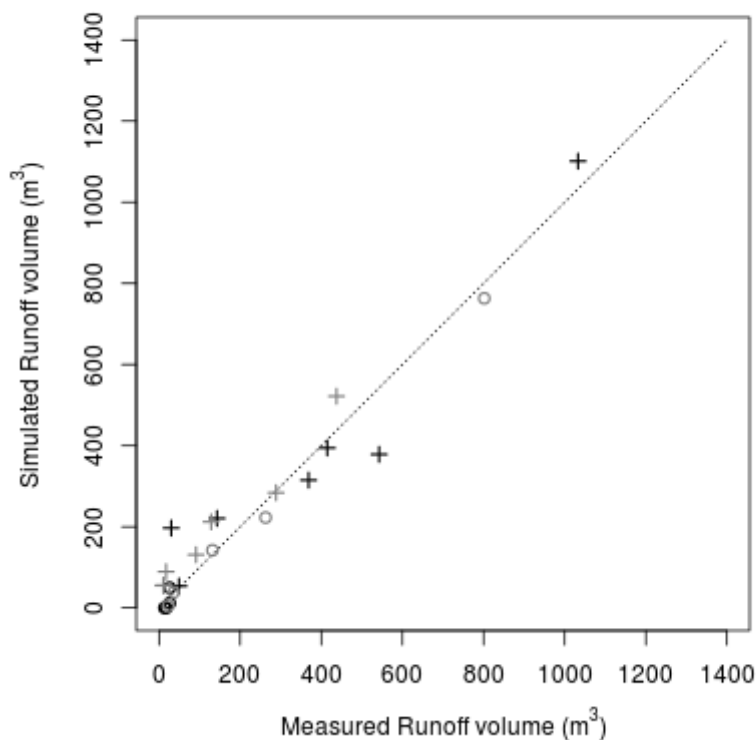
Infiltration capacity (mm.h ⁻¹)	API (mm)			
	0 - 2	2 - 6	6 - 9	> 9
0	0	0	0	0
5	7	1	0	0
10	8	2	0	0
15	8	2	0	0
20	9	3	0	0
25	9	3	0	0
30	10	4	0	0
35	11	5	1	0
40	11	5	1	0
45	12	6	2	0
50	13	7	3	1

During calibration of the STREAM model (distributed version), only a few modifications in the decision tables were required to obtain a satisfactory solution for the simulated runoff at the outlet. The main modification was to reduce the variability of the imbibitions volume from 5 mm to 3 mm.

Interviews with the farmers and plot surveys revealed no significant differences in the location of each plot, nor in surface roughness and soil tillage direction, but did reveal changes in the vegetal cover due to changes in crop or weed infestation during the rainy season. In the first part of the rainy season (from October to January) agriculture was extensive, with sparse crops and high weed density, resulting in high vegetal cover in the cultivated plots. In the second part of the rainy season (from February to mid-June) farmers plowed their plots in preparation for cultivating vegetables and tobacco, and controlled weeds, resulting in low vegetal cover or bare soil in the cultivated plots. Two vegetal cover scenarios were therefore used: one using a vegetal cover (the one observed in the second part of the rainy season) that remained the same throughout the rainy season, and one that adapted the vegetal cover to the date of the rainfall-runoff event.

3.3 – Runoff volume simulated at the outlet with the distributed model version and non-linear response of the basin

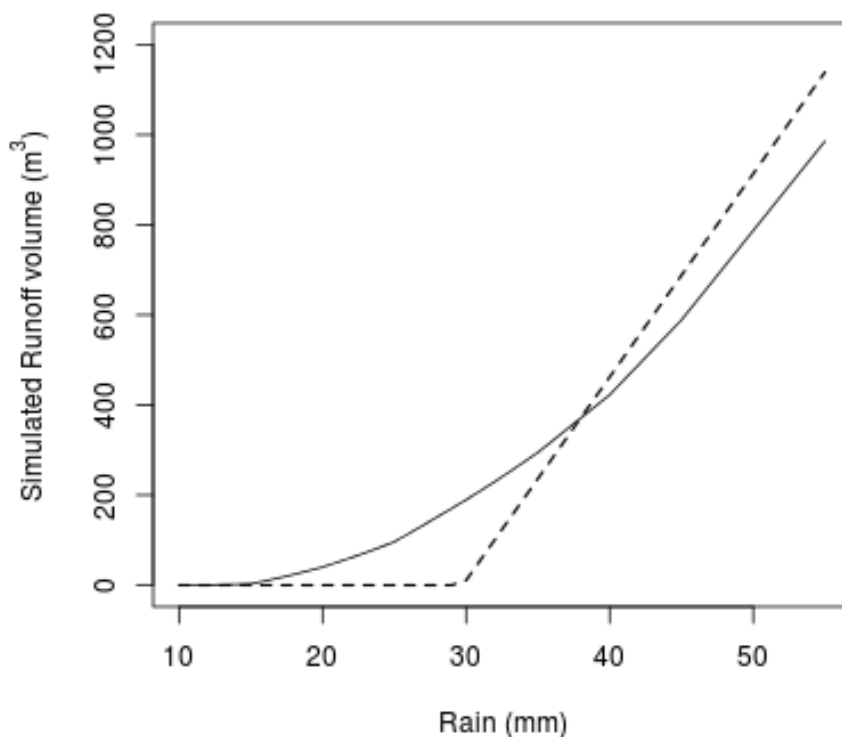
423 The runoff volume simulated with STREAM (distributed version) took into account the
 424 distribution of soil hydrologic properties and the vegetal cover (Fig. 5). Estimations computed
 425 with this distributed hydrological model were more accurate than with the lumped model. Using
 426 the same vegetal cover for all rainfall-runoff events, the average error was 75 m^3 and the Nash-
 427 Sutcliffe coefficient was 0.91. Taking changes in the vegetal cover into account slightly
 428 decreased the simulated runoff volume (less than 70 m^3 , which was within the initial average error
 429 error). However, this increased the model accuracy, as the average error decreased to 62 m^3 with
 430 a Nash-Sutcliffe coefficient of 0.94.
 431



432
 433 Figure 5: Simulated versus measured runoff volume for identified events using STREAM model
 434 with different land cover according to the season (grey-first half of the rainy season, black-
 435 second half of the raining season).
 436

437 Both runoff model versions were used to simulate one-hour rainfall events of different intensity
 438 with a full imbibition tank (Fig. 6). The distributed version of STREAM showed a non-linear
 439 response of the basin corresponding to the spatial distribution of the infiltration capacity,
 440 especially due to the spatial distribution of slope. Low intensity rainfall events produced runoff

441 only in the steepest plots, which were mostly located at the center of the watershed, where
 442 infiltration capacity was low. Increasing rainfall intensity produced runoff in plots with a less
 443 steep slope (that had a higher infiltration capacity) and hence increased runoff in the steepest
 444 plots. Under very intense rainfall (more than $45 \text{ mm}\cdot\text{h}^{-1}$) all the plots, including the flat plots
 445 located at the watershed border, were saturated and produced runoff: any millimeter per hour of
 446 rain above this threshold immediately produced runoff. This differential production of runoff
 447 was validated by our field observations during rainfall events. The steepest plots located in the
 448 center of watershed produced runoff even during short low intensity rainfall events, whereas the
 449 flat fields located at the watershed border only produced runoff during long or high intensity
 450 rainfall events. The distributed version satisfactorily simulated this contrasted effect, which was
 451 not accounted for by the lumped version.

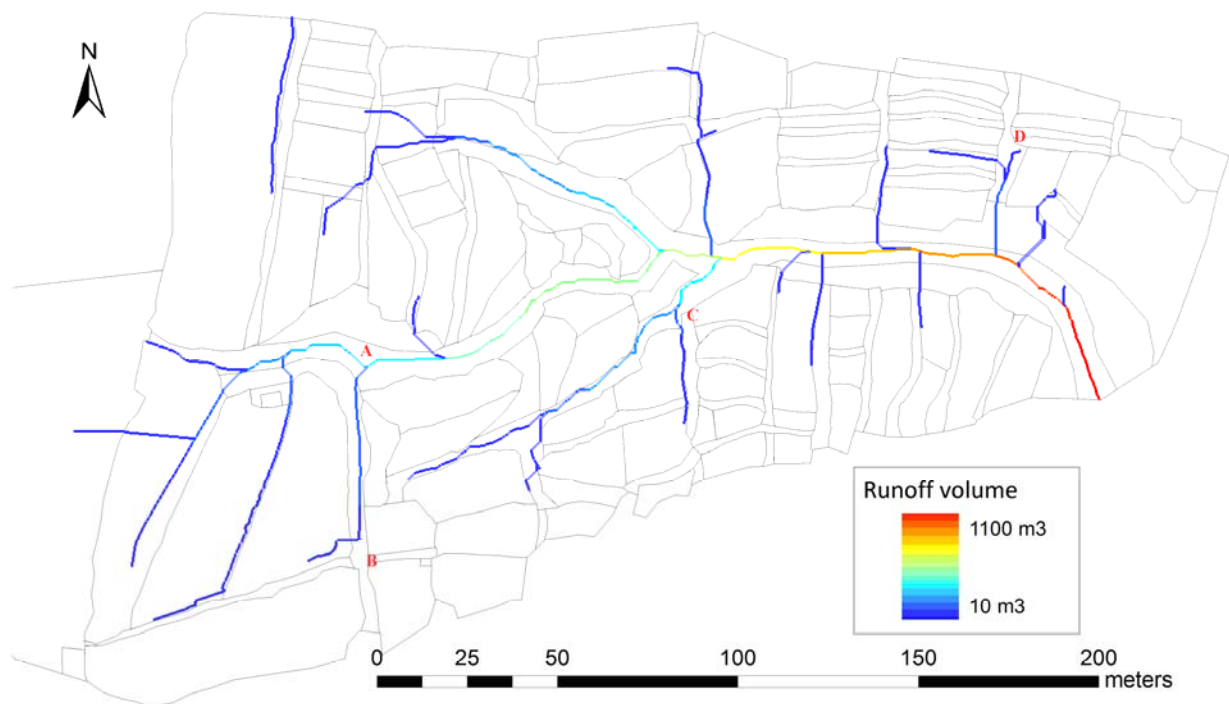


453 Figure 6: Simulated runoff volume for a virtual one-hour rainfall event with different rainfall
 454 intensities with a full imbibition tank (high API) with the non-distributed model (dotted line) and
 455 with the STREAM model (black line)

457 3.4 – Spatial distribution of runoff

458

459 The distributed version simulated runoff production at the pixel scale and consequently
 460 reproduced the spatial variability of runoff production at the basin scale (Fig. 7). The production
 461 of runoff was higher in the middle of the basin where the steepest plots are located. Agricultural
 462 practices impacted runoff through both vegetal cover and the tillage direction. However,
 463 variation in vegetal cover linked to the presence of a crop and weeds had less impact on runoff
 464 production than slope. Tillage direction influenced flow direction and hence the concentration of
 465 runoff in the furrows or channels at the borders of plots. Farmers' interviews and field
 466 observations confirmed model predictions of runoff production.
 467



468
 469 Figure 7: Spatial distribution of runoff volume over 10 m³ simulated by the STREAM model for
 470 the strongest rainfall event (64 millimeters in 70 minutes).
 471

472 The hydrographic network of the basin was highly developed and concentrated: simulated results
 473 showed that three main branches (west of the watershed) contributed to the majority of runoff
 474 (700 m³ corresponding to 60% of the total runoff volume), and after their confluence, seven
 475 waterways accounted for 30% of the total runoff volume (350 m³). Simulated runoff volumes in
 476 upstream waterway channels at point A, B, C and D were compared to observations made during
 477 intense rainfall events. For instance, runoff volumes at junction C simulated with a very intense
 478 rainfall event (50 mm.h⁻¹ during 56 minutes) were 122 m³ coming from the west and 33 m³
 479 coming from south (see Fig. 7). Our observation at junction C during similar rainfall events were

480 in accordance with the simulated result: it showed that about three-quarters of runoff came from
481 the west and one quarter from the south.

482 Different simulations identified the main impact of agricultural practices, i.e. tillage determined
483 preferential flow direction and waterway channels drained the plots. These anthropogenic
484 practices were responsible for the high runoff concentration.

486 **3.5 – Farmers’ perception of erosion**

487 The farmers distinguished between three types of erosion processes.

488 (i) Landslides were the most problematic for farmers because they destroyed the crop in their
489 plot, reduced the cultivated area, and required earthworks to repair the damage. Landslides
490 destroyed not only the plot in which it started but also the plot located in the downstream
491 deposition area. In the farmers’ opinions, there are two types of landslide: those caused by a
492 rupture of the plot border and those that start inside the plot.

493 Linear erosion was a less serious problem for the farmers who described two phenomena: (ii)
494 linear erosion of the inter-ridge inside the plots, caused by a rapid concentrated flow that could
495 harm the crop and destabilize the plot, and (iii) linear erosion of the waterways between the
496 plots, which can break through the man-made waterways and then overrun the plots, destroying
497 the crop or causing landslides inside the plots.

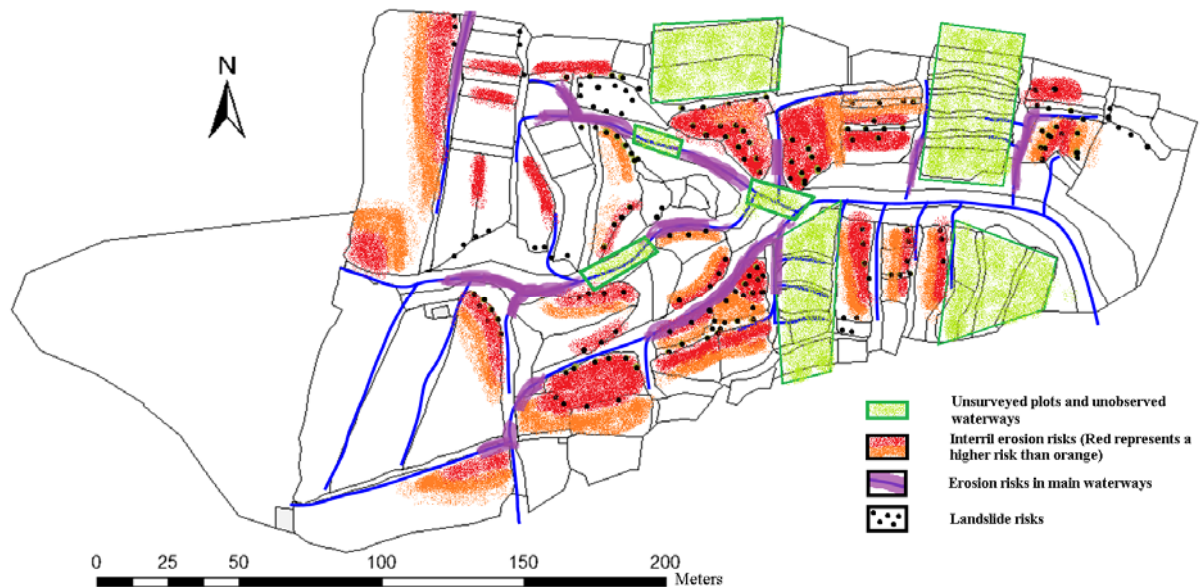
498 Farmers connected the intensity and frequency of erosion events: the location with the most
499 intense erosion was also the place where erosion occurred most frequently. The map in Fig. 8
500 presents the information provided by the farmers and our observations of erosion features in the
501 plots.

503 **3.6 –Mapping erosion risks**

504 Maps of erosion susceptibility and risk were created for the three types of erosion phenomena
505 described by the farmers: landslides, linear erosion within the plots, and linear erosion in the
506 waterways.

507 Concerning the risk of linear erosion in the permanent gully outside the field (Fig. 9)
508 susceptibility was zero in the main waterway because it was located on andesite lava (Fig 9-a).
509 Susceptibility was higher in the other waterways because they were located on andosol, which is
510 mainly composed of ash deposits. In addition, susceptibility to linear erosion was connected with
511 the steepness of the slope of the waterway. Combining the susceptibility map with the hazard
512 map (runoff volume distribution) resulted in a risk map (Fig. 9-b). Erosion risk was high in the
513 three main tributary waterways where runoff volume was high, but it was also very high in very

514 steep secondary waterways where the runoff volume was low. Field observations during intense
 515 rainfall events confirmed the location of erosion risk in the waterways (see Fig. 8). Field
 516 observations also showed that the places with high risk (see Fig. 8) corresponded to a steep
 517 waterway or high runoff flow. In addition, observations at point A during intense rainfall events
 518 showed that erosion led to a digging of the existing waterway channel.
 519
 520



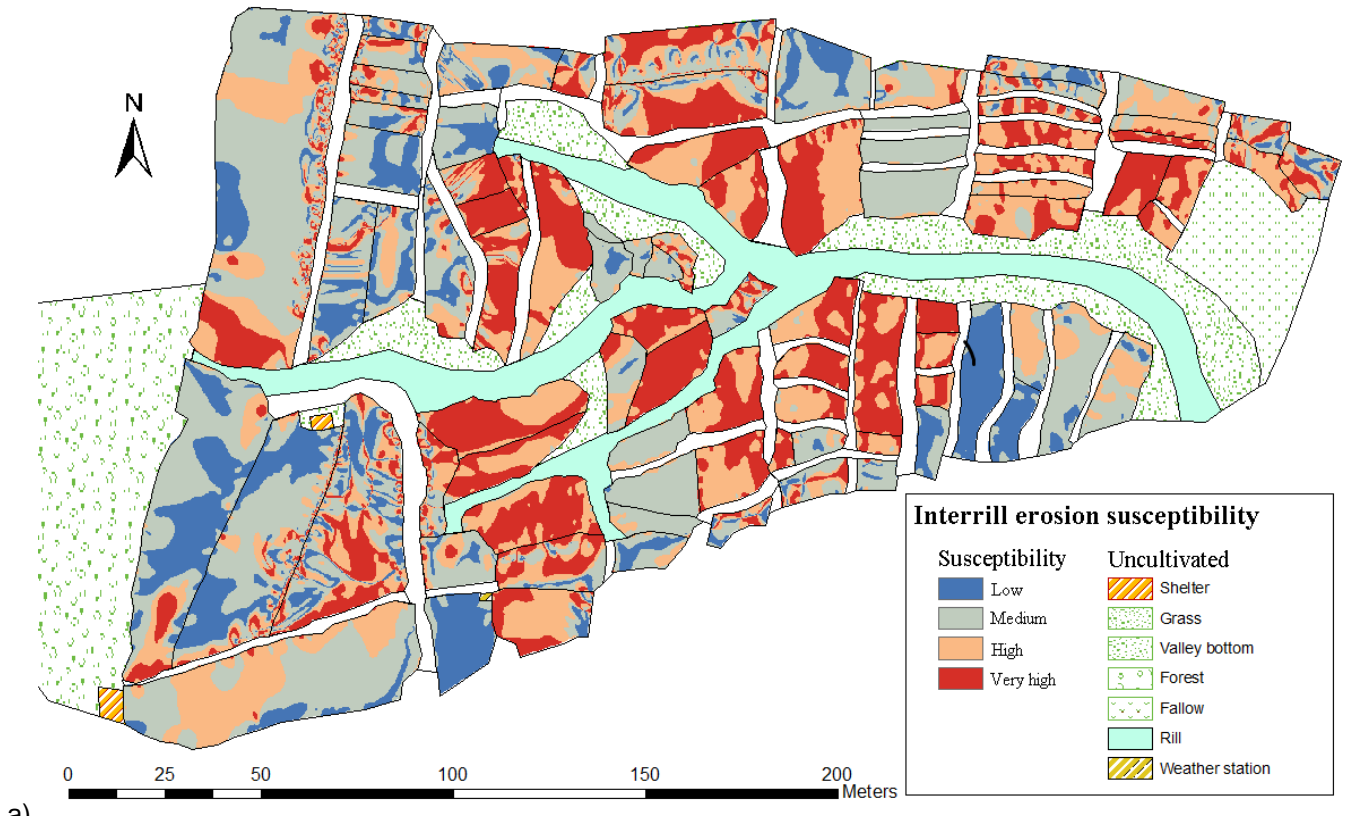
521
 522 Figure 8: Map of erosion risks based on farmers' interviews and field observations *In the figure*
 523 *below, it should be* 'Meters' 'Unsurveyed plots and unobserved waterways' 'Red represents a
 524 higher risk than orange'
 525

526 Concerning linear erosion in the *transient intra-field rill* (Fig. 9) *the* main factor influencing
 527 susceptibility was slope, which can be tempered by tillage across the natural slope and vegetal
 528 cover. Susceptibility was consequently high in steep areas, and these were quite widely
 529 distributed in the basin (Fig. 9-a). The main risk of linear erosion inside the plots was
 530 concentrated in the center of the basin where runoff production was higher (Fig. 9-b). The map
 531 shows increasing risk of erosion caused by the accumulation of runoff downstream (green to
 532 yellow to red). Comparing field observations and information collected in interviews with the
 533 farmers (see Fig. 8) confirmed the location of the risk of linear erosion in the plots. Places with
 534 high, medium and low linear erosion risk described by the farmers (Fig. 8) corresponded to those
 535 predicted by the risk model.
 536
 537

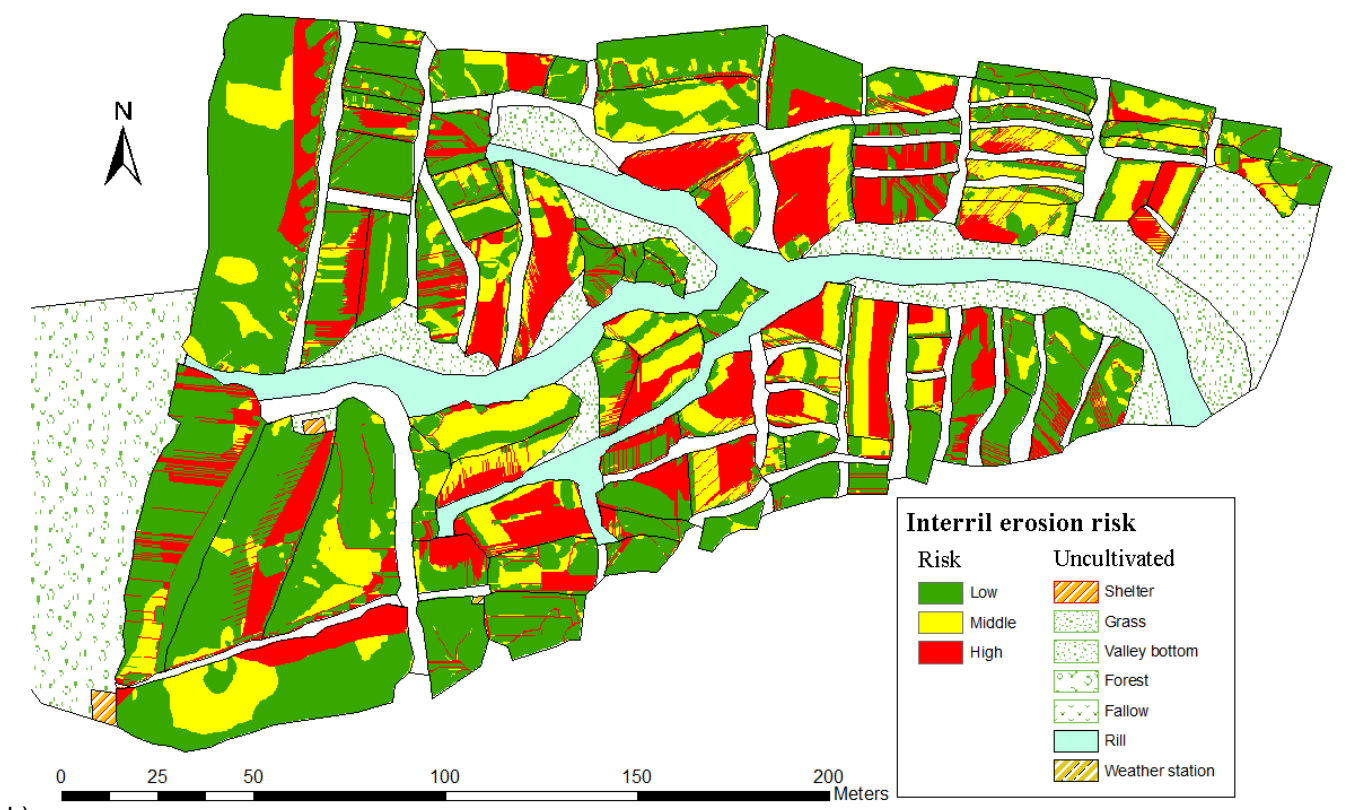


Figure 9: Susceptibility of waterways to linear erosion (a) and risk (b) maps

538
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a)

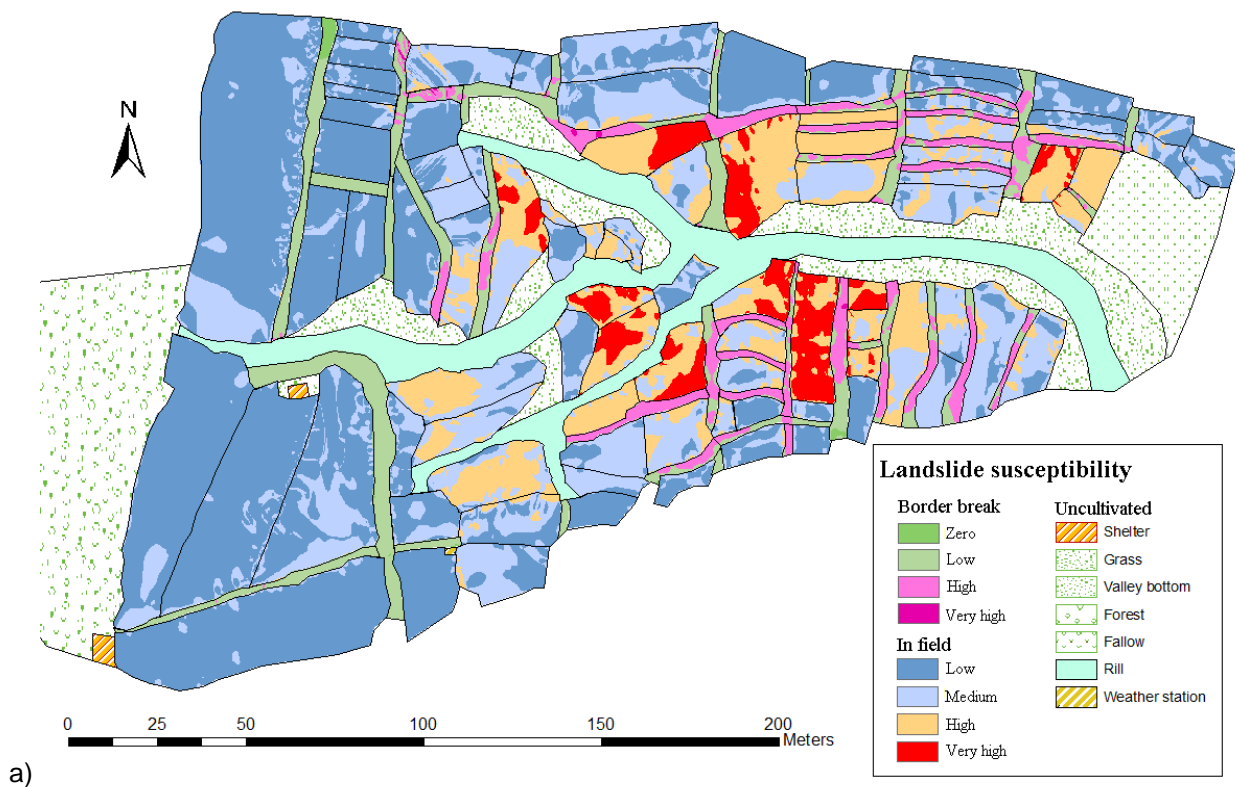


b)

544
545
546

Figure 10: Map of inter-rill susceptibility (a) and risk map (b) *In the key in the top panel it should be 'Uncultivated'*

547 Susceptibility to landslides was concentrated in steep areas, which are mainly located in the
 548 center of the basin (Fig. 10-c). Flat plots located at the edge of the basin were less susceptible.
 549 Plots located on the banks of the main waterways were uniformly highly susceptible in steep
 550 areas but only slightly susceptible elsewhere. Hazard (infiltration and runoff) concentrated the
 551 risk in the center of the basin (Fig. 10-d). Only a few plots were subject to high risk of landslide,
 552 but in these plots the risk was nevertheless severe. All five landslides that occurred during our
 553 field observations happened in this highly risky area. Moreover, except for two plot borders, the
 554 locations of the landslides described by the farmers (Fig. 8) corresponded to highly risky areas
 555 on the simulated risk map.



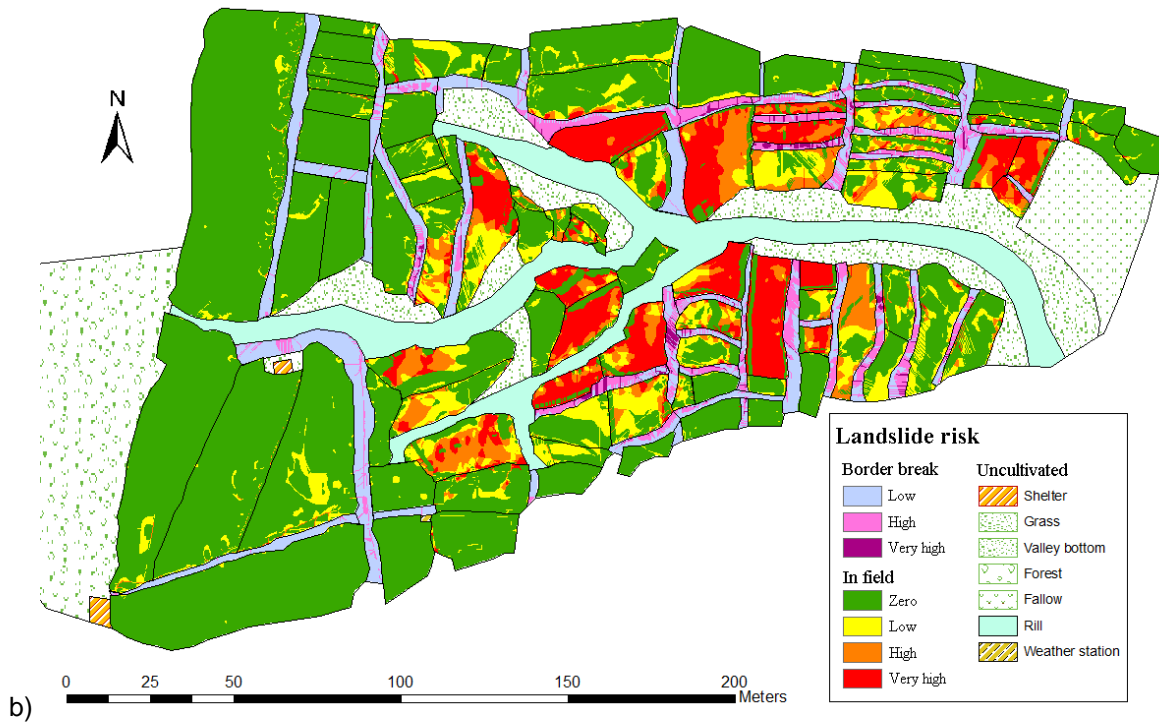


Figure 11: Map of landslide susceptibility (c) and risk map (d)

Erosion maps revealed the major impact of tillage direction on the different erosion risks. The risk of linear erosion inside the plot decreased if the slope in the tillage direction was less than the natural slope. But if it was too flat (perpendicular to the natural slope), the risk of landslide in the plot increased. Tillage direction determined runoff direction and concentration and therefore influenced the hazard maps (runoff and infiltration). It consequently increased the risk of erosion in the plot or in the waterway that received the runoff flow. Moreover, linear erosion in waterways created deep gullies with instable flanks, and, as a consequence, increased the risk of soil losses (which can lead to landslides) at the plot borders.

4. Discussion

4.1 Watershed response

Intense tropical rainfall events produce flash runoff in steep micro-watersheds in Indonesia (Rijsdijk et al. 2007; Van Dijk, 2002). The lumped runoff production model is quite simple, uses few parameters and satisfactorily simulates runoff production at watershed scale. The distributed STREAM model uses qualitative or quantitative variables classified in 4 or 5 categories based on simple field observations and farmers' interviews. Accounting for the spatial distribution of

577 hydrological properties increases the quality of the runoff simulations, as shown by Turkelboom
578 et al. (2008). The expert based distributed model with only a few infiltration capacity classes was
579 consequently more accurate than the adjusted linear lumped model in simulating runoff in this
580 steep cultivated watershed.

581 The STREAM model predicted total event runoff accumulation without taking the temporal
582 dynamics of runoff during the rainfall events into consideration. This can have a major impact on
583 erosion assessment because peak discharge plays a determining role in the evaluation of linear
584 erosion in the waterways. However, runoff models that represent peak discharge require a lot of
585 spatially distributed data, which, in our case, were not available. Moreover, as Gumuk is a very
586 small watershed (4.5 ha), short intense rainfall events (less than 2 hours) produced short runoff
587 events (less than 4 hours), and total runoff volume at outlet was therefore a good indicator of the
588 intensity of runoff flow (an affine regression between total runoff and peak flow on 9 events
589 gave a positive relation with $r^2=0.81$). However modeling longer runoff events in a larger
590 watershed would require taking temporal dynamics into account (Morgan et al., 1998).

591 592 **4.2 Spatial distribution of runoff production and accumulation**

593 Runoff production was modeled using a limited number of input parameters. Based on field
594 observations and laboratory experiments, the soil parameters in the watershed were considered to
595 be homogenous. However, a more detailed analysis could reveal some heterogeneity of the soil
596 hydrodynamic properties, and could also reveal the contribution of soil variability to the
597 variability of runoff production. This soil heterogeneity could be then added in the STREAM
598 model through the properties of plots.

599 Human activities can greatly modify runoff concentration pathways (Souchere et al., 1998;
600 Moussa et al., 2002). In this study site, which is characterized by small size, high slope
601 variability, and the effect of many human actions on the topography (construction of waterways,
602 terraces, and ridges), runoff circulation was not clear. A high resolution DEM was therefore
603 required to simulate the flow directions at plot and waterway scales. A precise description of the
604 topography was possible because the watershed is small. In a larger watershed, remote sensing
605 could be used to produce a DEM, but its precision would not be sufficient to represent the impact
606 of farmers' practices on the distribution of runoff.

607 Precipitation was considered to be homogeneous throughout the watershed, but in fact, the relief
608 and the wind may have a major impact on the distribution of precipitation within the watershed.
609 A better representation of rainfall distribution would require several weather stations (for
610 instance on both banks of the main stream) and a distributed model of precipitation.

4.3 – Validation of distributed runoff model

Measurements of runoff accumulation in plots, and in natural and human-made waterways made it possible to quantitatively validate the runoff model. However, these measurements are delicate and time consuming and their coherence at different scales is not guaranteed (Le Bissonnais et al., 1998). The low accuracy measurements made during the present study could only be used for the purpose of comparison and to give an order of magnitude. The use of farmers' knowledge on runoff concentration and production in their plots was also delicate, mainly because perceptions may vary between farmers and differ from scientific measures and observations. However, interviews conducted with the farmers in their fields enabled us to compare and calibrate farmers' perceptions with our own observations. The validation of runoff estimated by the distributed STREAM model was not based on accurate measures but on the consistency of the simulated results, field observations, and farmers' knowledge. Nevertheless, the distributed runoff model enabled us to understand the runoff distribution within the watershed. It showed how the farmers managed runoff, built ridges and water channels to control the runoff that accumulated in their plots and to redirect the runoff flow away from the downstream plots. Quantitative validation of the distributed runoff model would require measurements of runoff flow inside fields and waterways.

4.4 – Erosion models

Susceptibility to erosion did not account for the impacts of erosion (or resilience to erosion) like vulnerability studies (Leone and Vinet, 2006). Including vulnerability would require assessing the impacts of erosion in the short term (on farm income) and in the long term (loss of arable land). An erosion event could have a cascade of consequences and the different forms of erosion were not independent. For instance, field observations revealed that linear erosion in waterways could cause a landslide, a small landslide could cause a bigger one during the following rainfall event, and an erosion event could modify both the topography and the vegetal cover. Assessing these cumulated risks would require other field surveys and further interviews with farmers. Quantitative validation of the erosion risk models would require measurements of sediment yield and topography in different locations inside the plots and in waterways. One of the main difficulties would be identifying the contribution of the different forms of erosion to sediment yields using isotopic C-137 or granulometry, for instance. It would require long term on-site measurements that are both technically difficult and expensive.

644 Qualitative validation of erosion risk maps was based on our field observations and farmers'
645 interviews. Field observations concerned only the current cropping season, and the farmers
646 provided historical information. The information provided by the farmers about the preceding
647 cropping season had to be checked and harmonized by comparing it to field observations and
648 using photos of traces of erosion. More complex physically-based models would be extremely
649 difficult to calibrate and validate because of the lack of data and due to the specificities of the
650 watershed: marked variability in slope and the impacts of agricultural practices and topography
651 that can differ from year to year. Moreover, the simulated sediment yield would be the product of
652 different forms of erosion.

653

654 **5. Conclusion**

655 The spatially distributed models (runoff and erosion risks) developed in this study used a variety
656 of data sources for their calibration and validation when classical data were missing. The
657 distributed runoff model enabled us to build maps of three major forms of erosion risk. These
658 models mainly provided an order of magnitude or qualitative results. However they enabled a
659 better understanding of phenomena, particularly their distribution. First, models enabled the
660 researcher to locate erosion issues especially those combining forms of erosion that had not been
661 identified during the field observations. The farmers confirmed the importance of those locations
662 for the management of erosion. Second, models gave a distributed and global representation of
663 erosion that differed from the farmers' scale. The aim of the agricultural practices used by the
664 farmers in their fields was managing runoff concentration, diverting flow from the main slope
665 through tillage, and channelling the water in waterways. But the risk of erosion increased with
666 slope and runoff flow was therefore higher in downstream plots. Thus, managing erosion risk
667 calls for coordination at basin or sub-basin scale. The modelling approach we developed was
668 therefore an appropriate way to get round the lack of quantitative data.

669 Maps of erosion risks could be used to draw up plans for coordinated practices and their location.
670 Runoff models could be used to test the effect of these practices on runoff flow, the hazard part
671 of erosion risk. Designing strategies to reduce erosion risks could be done classically by external
672 experts (Turkelboom et al., 2008) or with the participation of farmers (Souchère et al., 2010;
673 Furlan et al., 2012). The second approach would have the advantage of envisaging and
674 discussing solutions that can be implemented by the farmers themselves. In this approach, a map
675 of erosion risk could be a support for reflection by the farmers: the maps we produced are
676 accurate, simple, intuitive, and compatible with the farmers' perceptions.

677

678 6. Acknowledgments

679 This research was funded by the project “*Integrated and Participatory Water Resources*
680 *Management towards effective agricultural system in Klaten Regency*” financed by Danone-
681 Aqua. We would like to thank the Balai Pengkajian and Teknologi Pertanian of Yogyakarta
682 (BPTP DIY) for their support in the field. We are very grateful to Pak Bruto and his wife for
683 their warm welcome and their three month hospitality. We are deeply grateful to Putra Nasution
684 for his help in the field and thank the inhabitants of Gumuk for their kindness and availability.

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