

Added value of citizen science to record extensive disasters

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EXTENSIVE DISASTERS ?

low-severity, high-frequency
hazardous events

highly **localized** hazards

typically **flash floods**
and **landslides** in
mountainous terrains

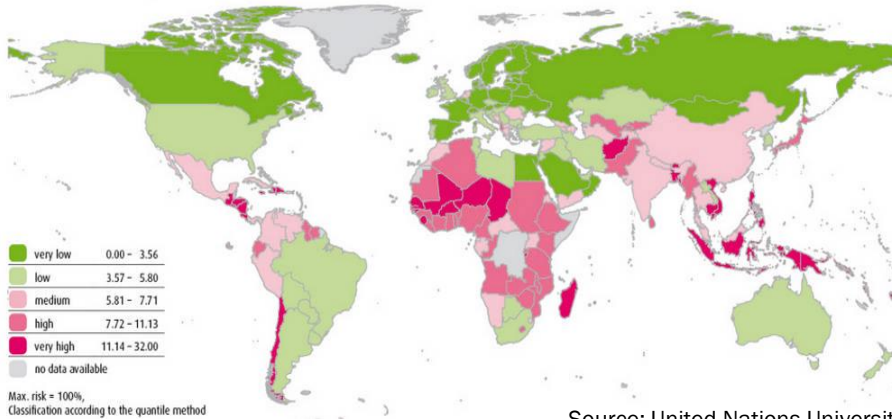
absent of databases with
high entry criteria; but
significant cumulated
impacts



NATURAL HAZARDS/RISK IN THE GLOBAL SOUTH

WorldRiskIndex

WorldRiskIndex as the result of exposure and vulnerability



Source: United Nations University

5x

more **fatalities** in low income countries

25x

larger **impact** of weather-related disasters in low-income countries

40%

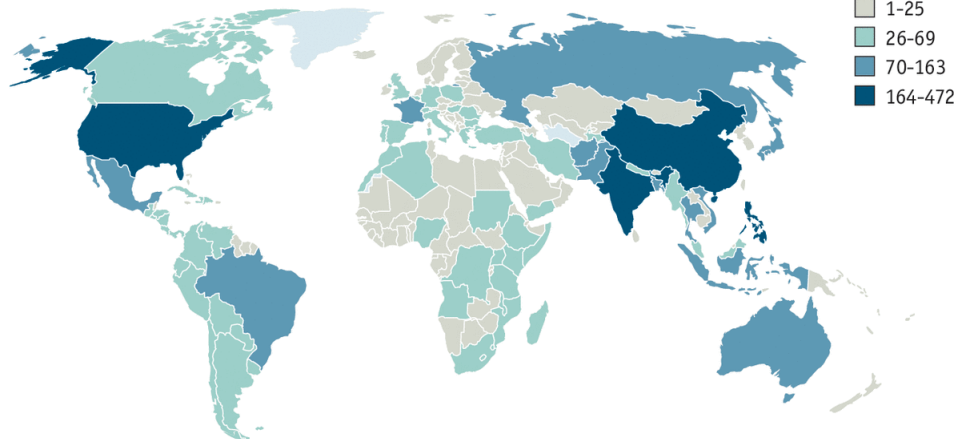
economic loss is due to **extensive risk**

?

underreporting of disasters, blind-spot in disaster research

Disaster zones

Total number of natural disasters* reported per country, 1995-2015



Source: UNISDR

*Hydrological, climatological and meteorological

EXTENSIVE DISASTERS in CENTRAL AFRICA

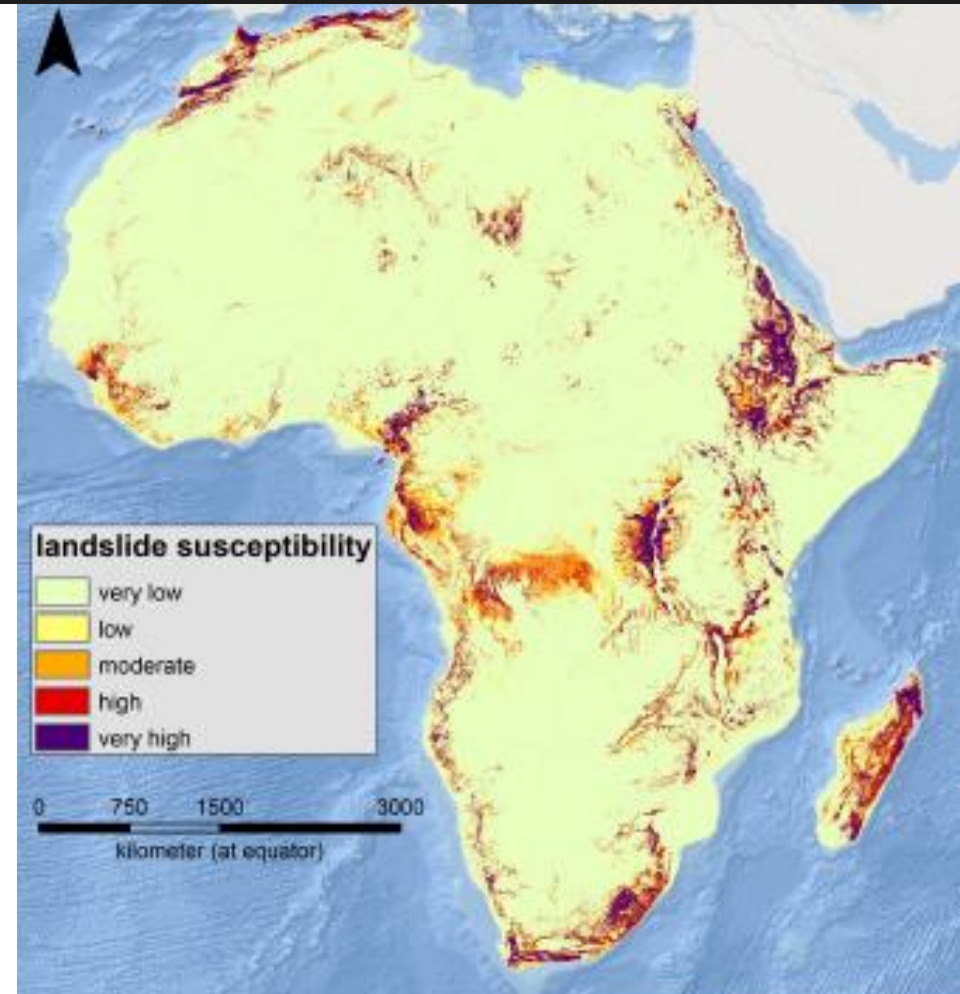
Rapidly increasing & vulnerable populations

High susceptibility to landslides along East African Rift

High mortality per disaster in low income countries

Global blindspot for landslide hazard and risk studies

What hazards? Where? When?
What impacts? Why?



(Broeckx et al. 2018)

DISASTER DATA SOURCES for LANDSLIDES

- EM-DAT
 - DFLD – Durham Fatal Landslide Database
(Froude & Petley 2018)
 - Global Landslide Catalogue (NASA)
- ⇒ Cooperative Open Online Landslide Repository
(Juang et al. 2019)
- DesInventar - National databases
 - Regional to sub-national crowd-sourcing databases
 - Social media

fatalities x15-20
events x3-4

Data sharing/mining
Spatial aggregation
Global scale
Consistent
High entry criteria

Data Mining
Spatial reporting biases
Low entry criteria

Voluntary contributions
Localized events
Multiple potential biases
Limited to no validation
Low entry criteria

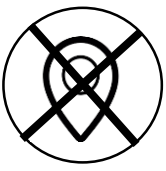
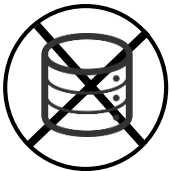
**Very prone to
(multi)- hazards**



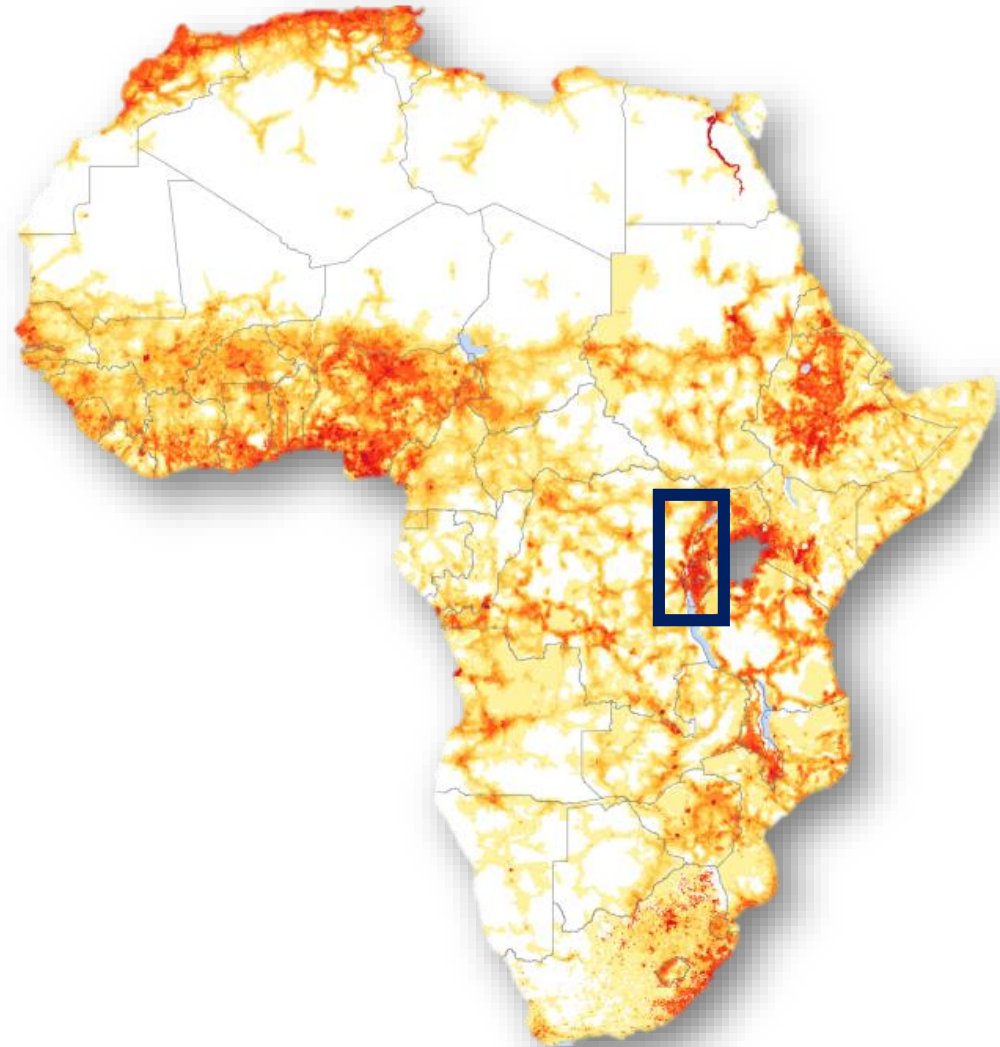
**high population
density and
vulnerability**



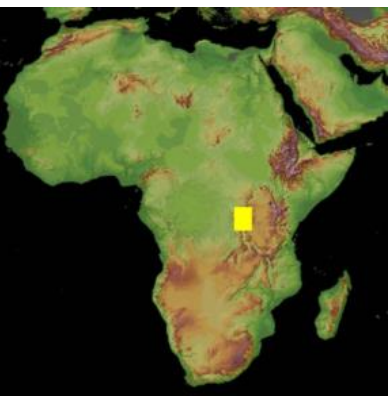
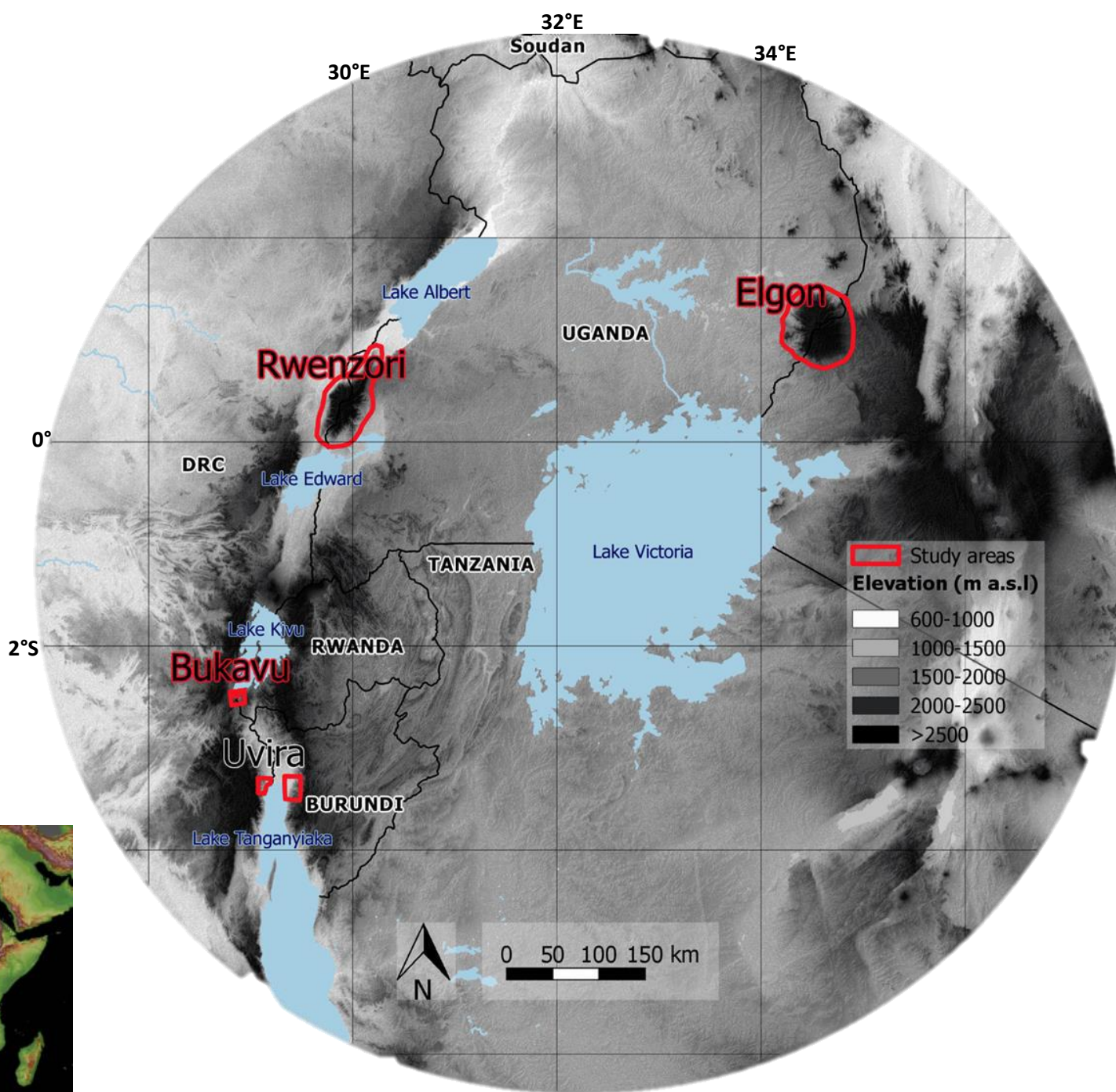
**No centralized
database/information**

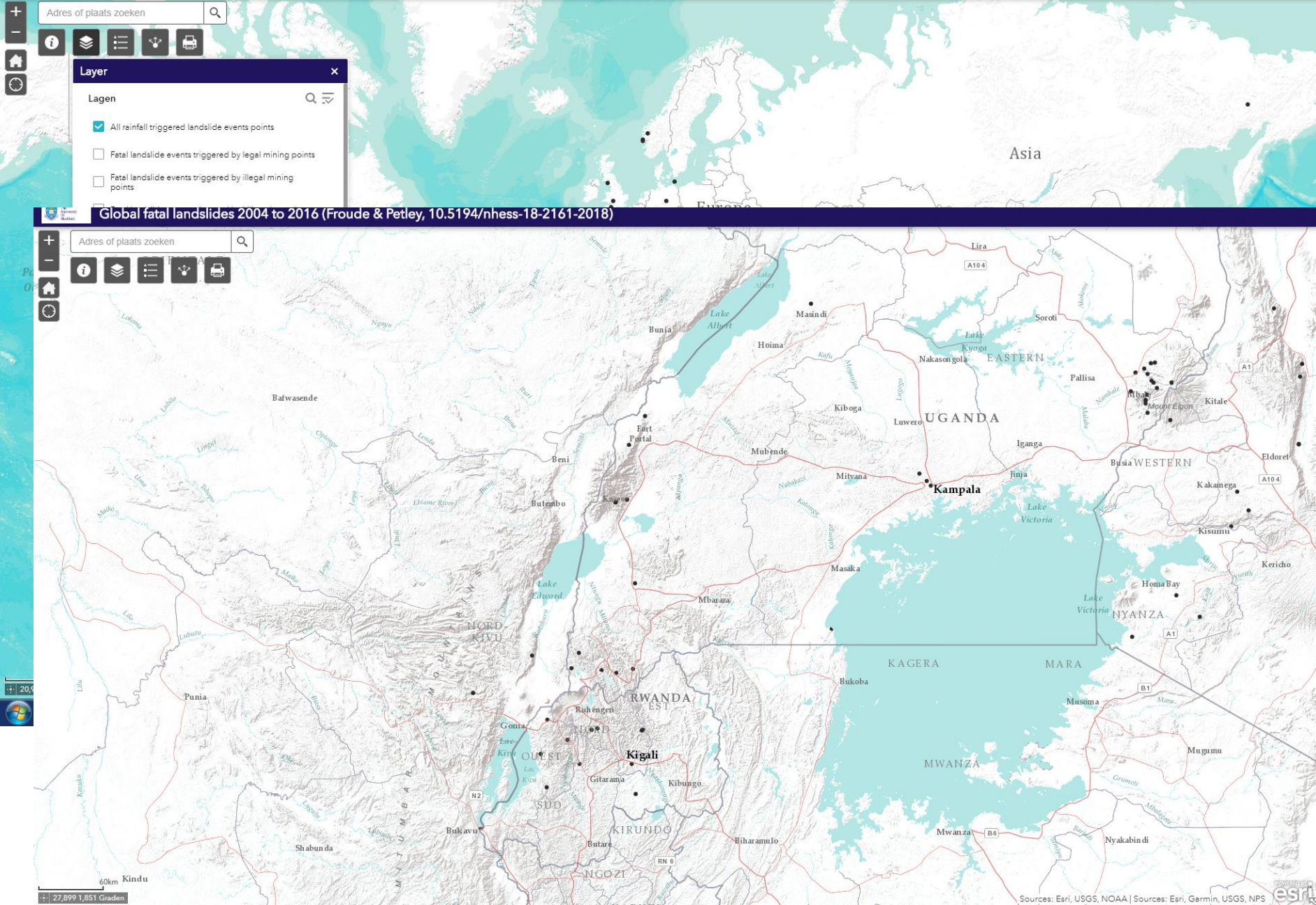


**Our study area:
Kivu & Rwenzori Mountains:**



STUDY AREA

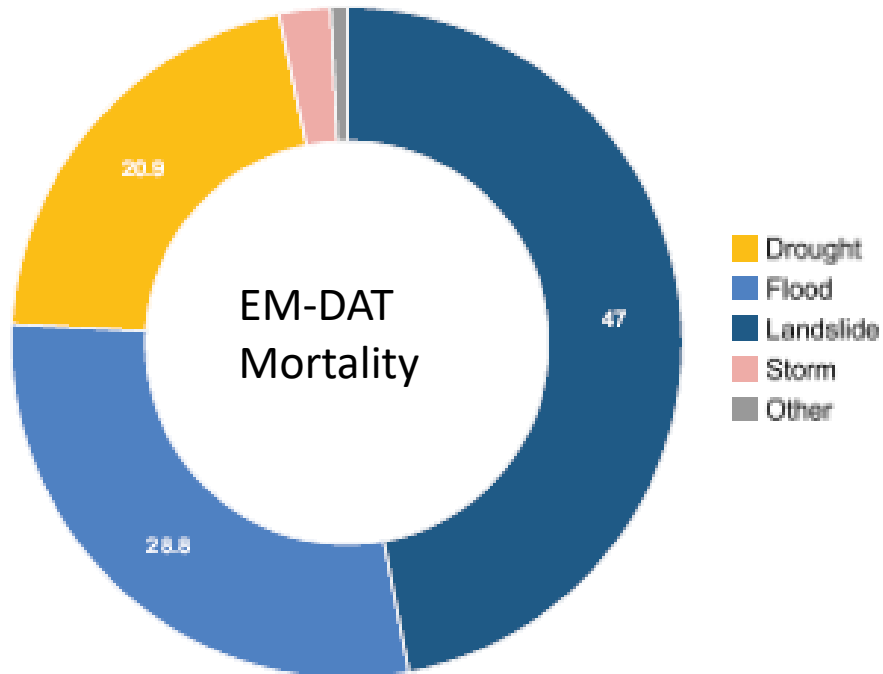
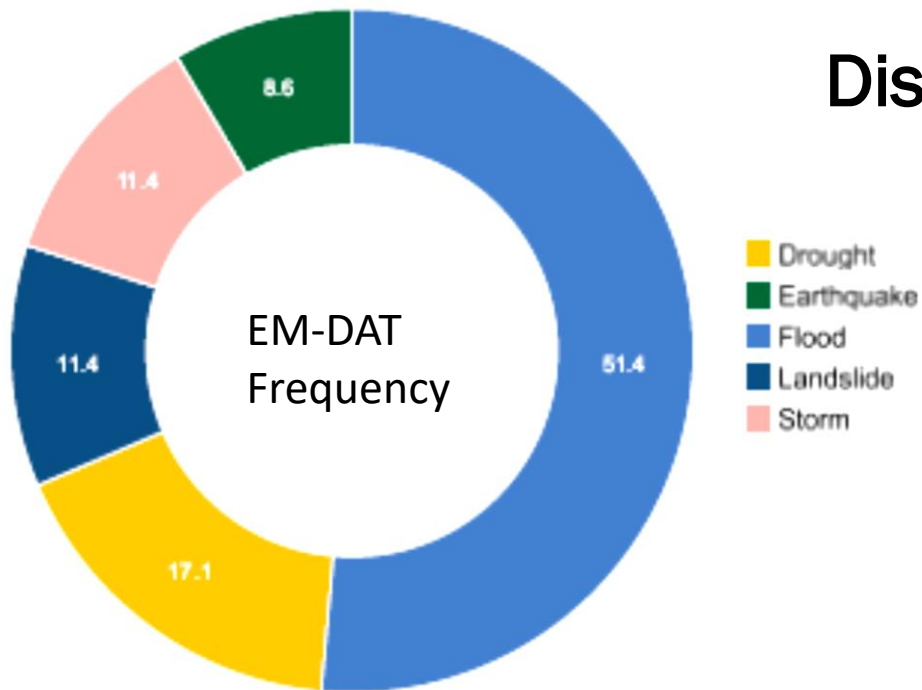




DFLD – Durham Fatal Landslide Database

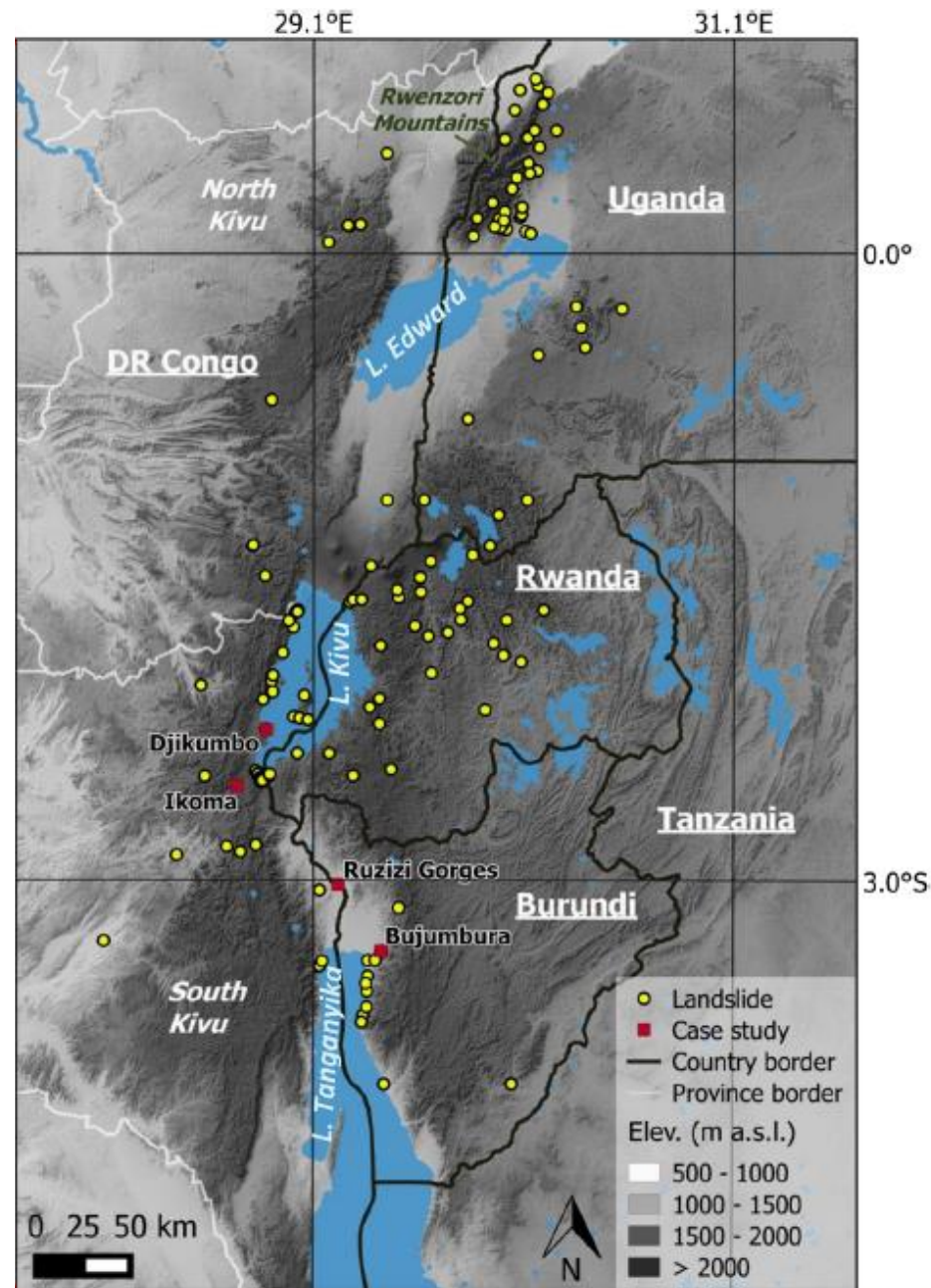
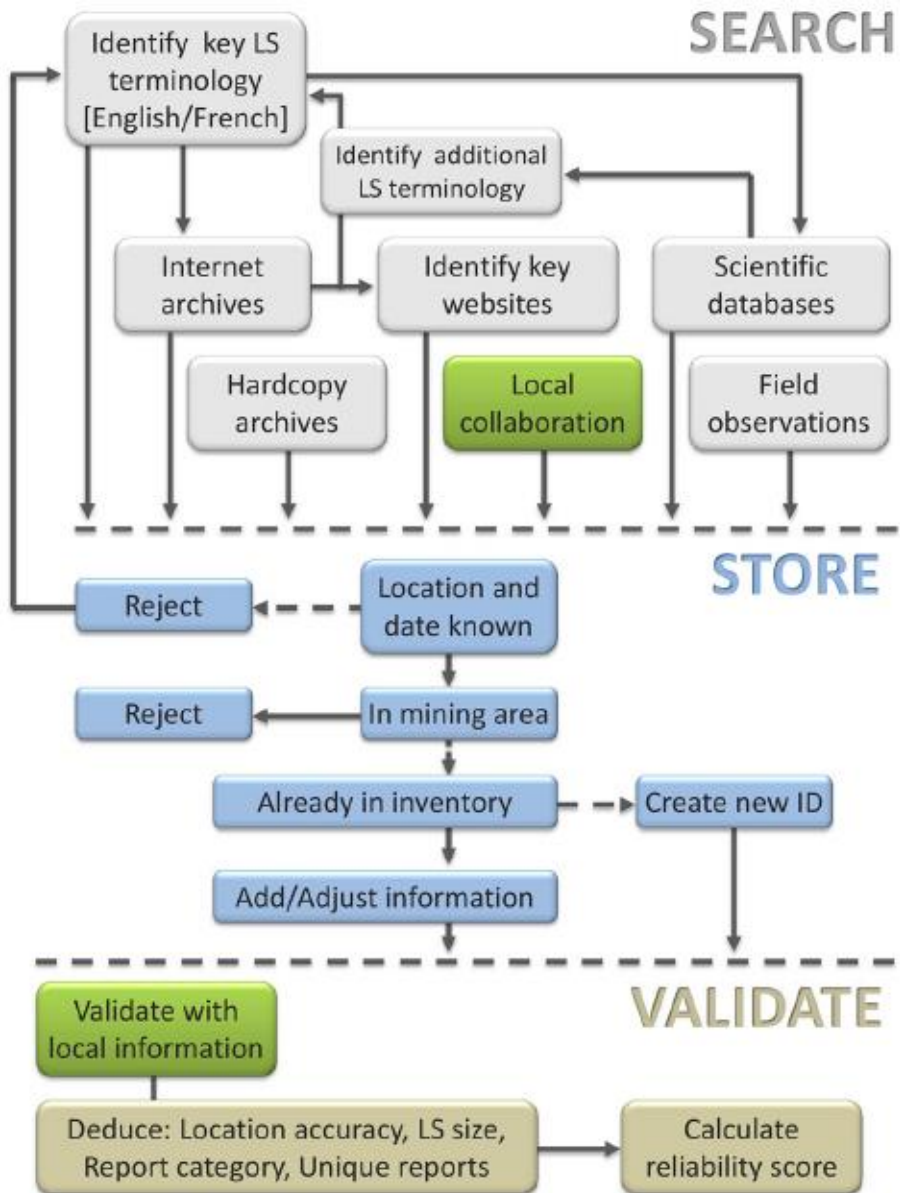
(Froude & Petley 2018)

Disaster Statistics in Uganda:



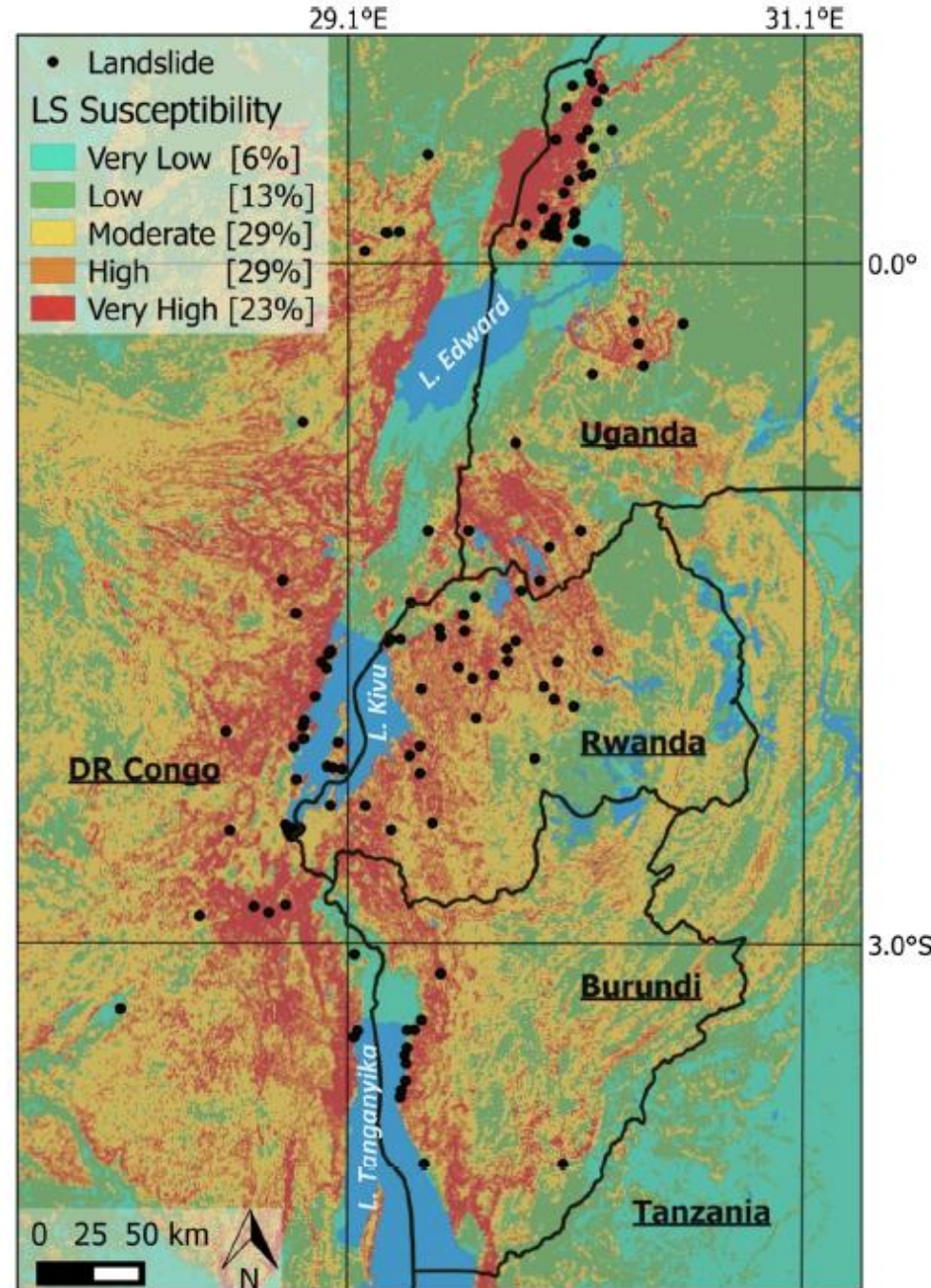
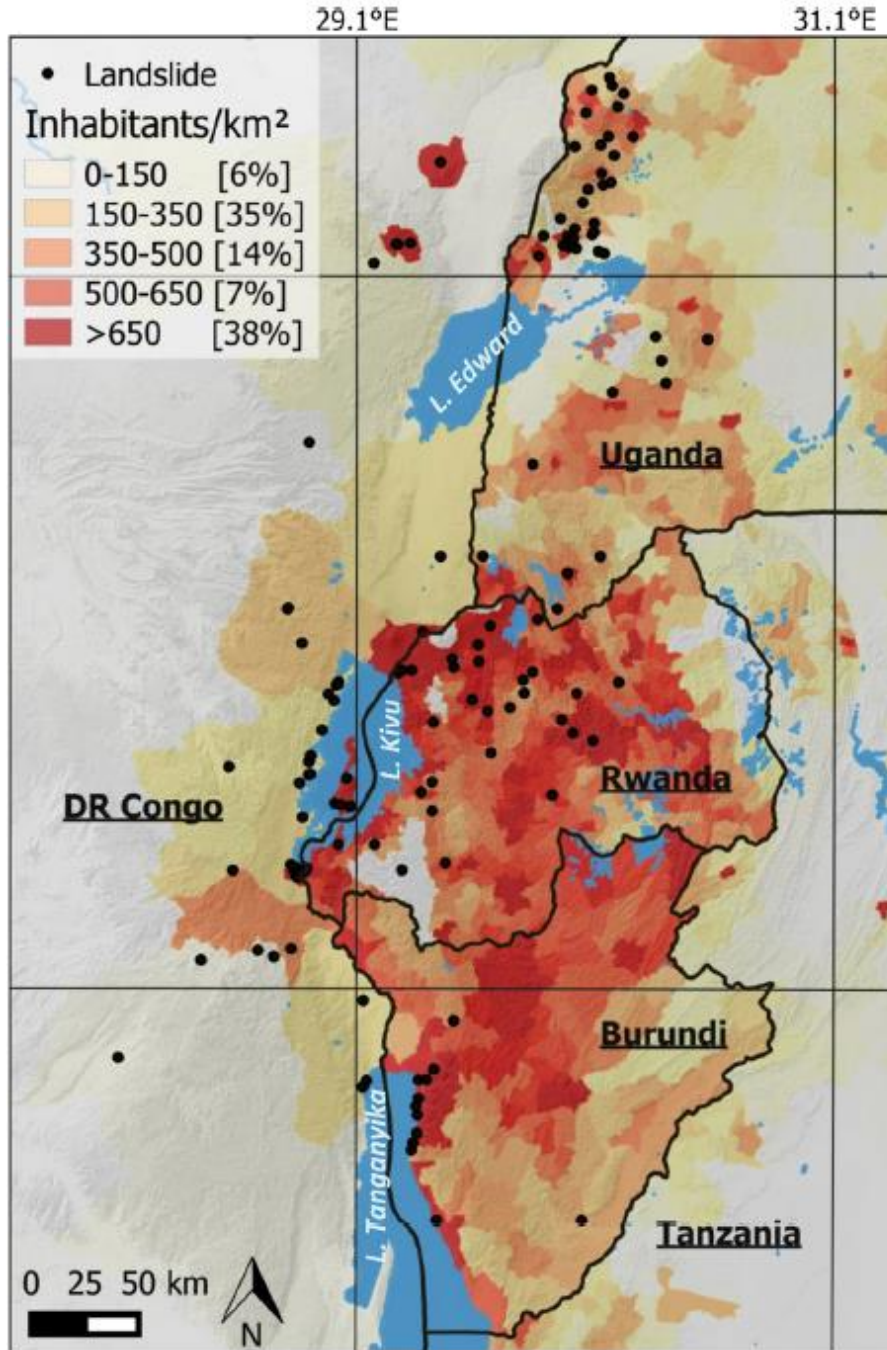
Bududa District, Eastern Uganda, 2012
Daily monitor

DATA MINING

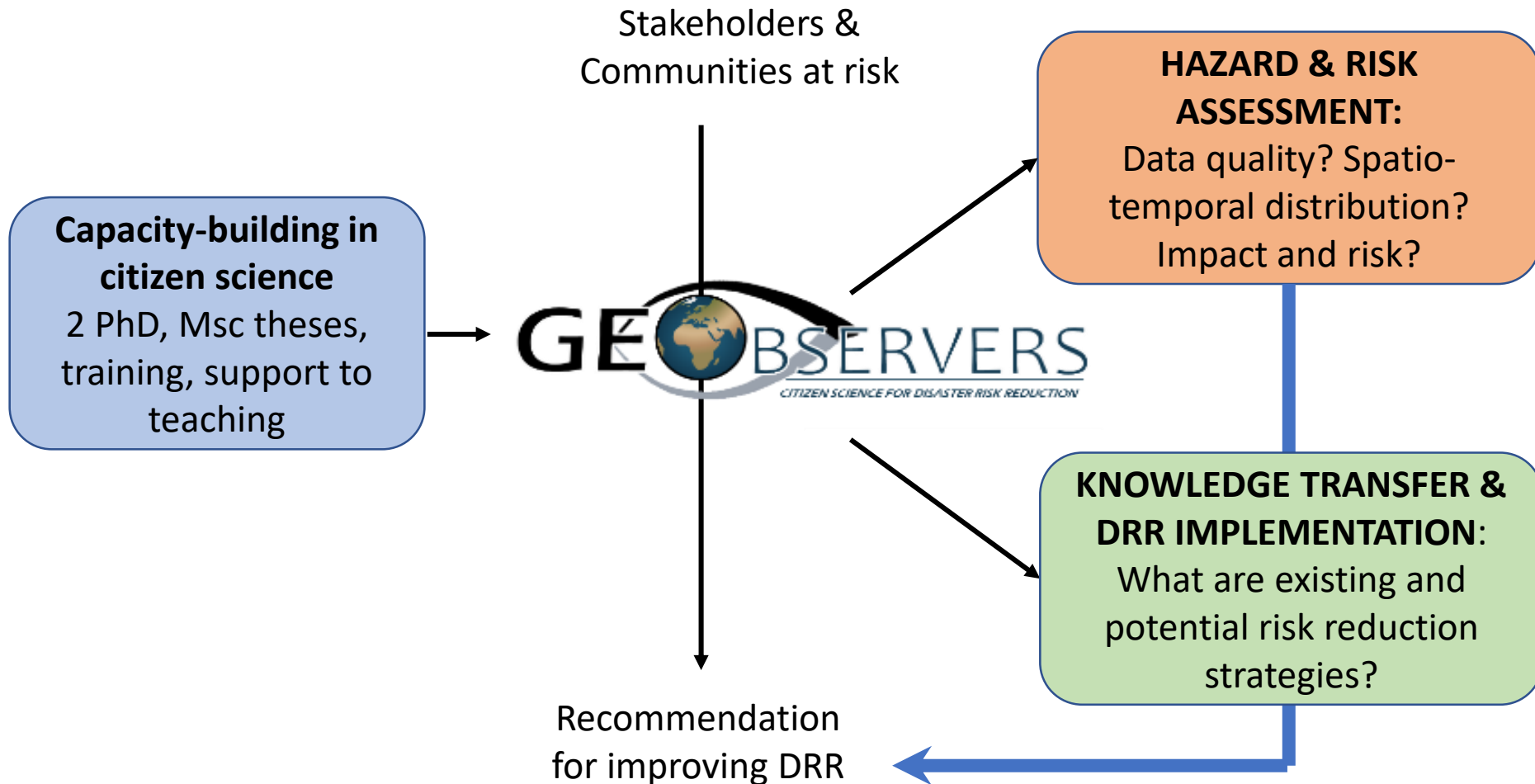


156 identified LS events in the study area (13 in EM-DAT)

Monsieurs et al. 2018



Can **Participatory Sensing** by Citizen Scientists alleviate the **data bottleneck** in evidence-based policy development?



Geo observer network

Knowledge on the **timing and location** of hazards and the **impact** they cause is the **first step** towards disaster **risk reduction**

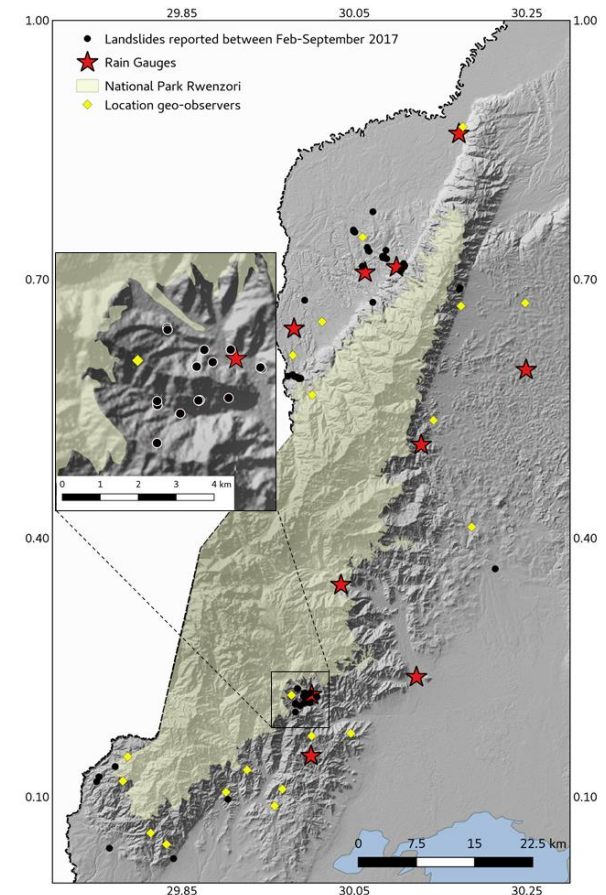
Local stakeholders are often **consulted** but rarely **involved** during the data collection phase

Local people are a valuable asset in collecting data on their environment:

- thorough **knowledge** on their environment
- in **direct contact** with the inhabitants of the region
- direct **access** to the field



THE SET UP OF A GEO-OBSERVER NETWORK

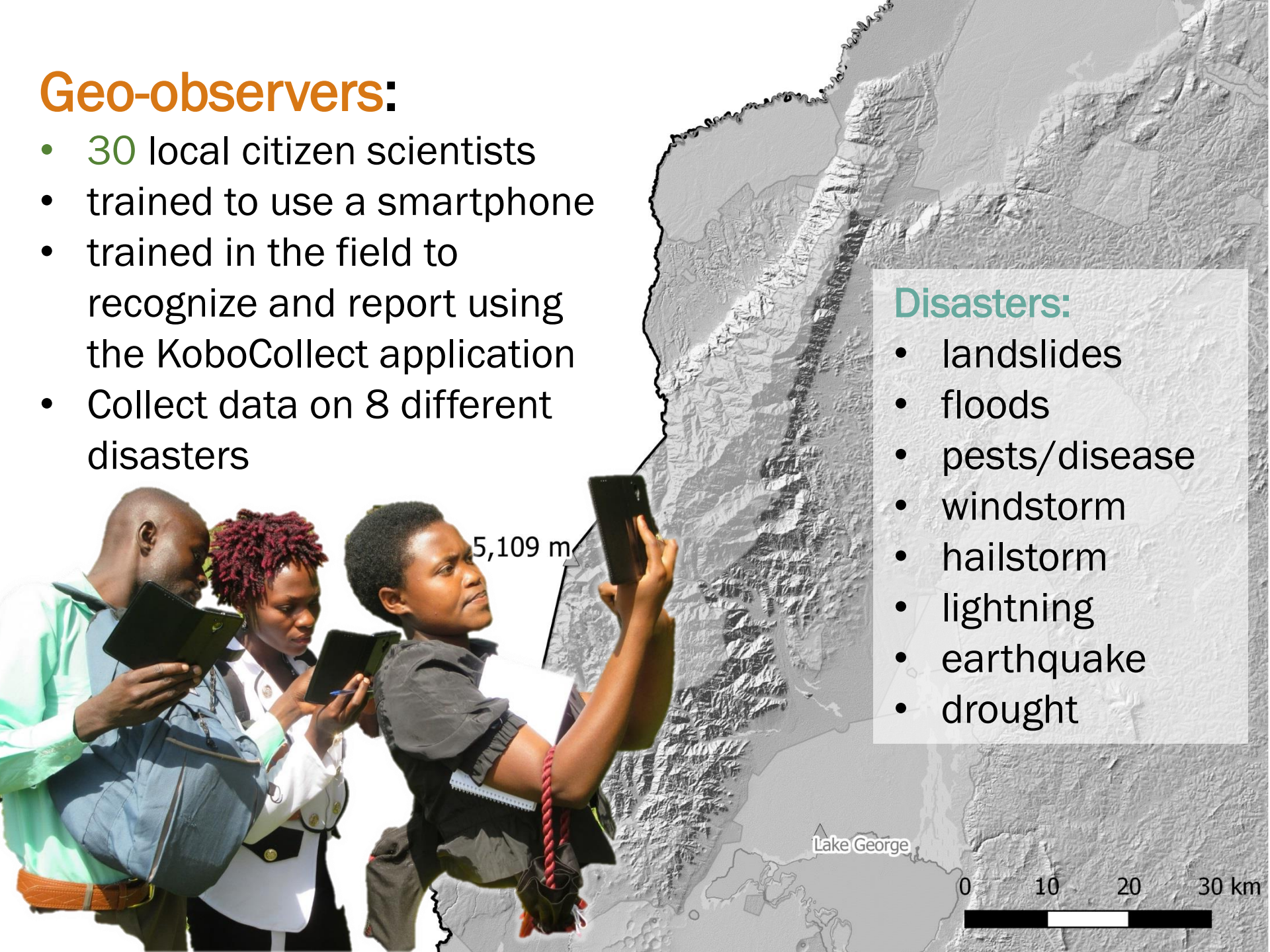


Geo-observers:

- 30 local citizen scientists
- trained to use a smartphone
- trained in the field to recognize and report using the KoboCollect application
- Collect data on 8 different disasters

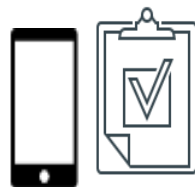
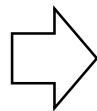
Disasters:

- landslides
- floods
- pests/disease
- windstorm
- hailstorm
- lightning
- earthquake
- drought

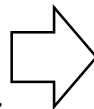




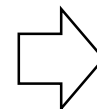
1. Geo-observer receives information



2. Geo-observer reports using the app



3. Geo-observer sends report via mobile network



4. Quality check by Mountains of the Moon University



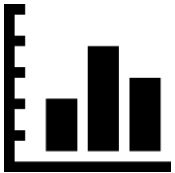
METHODOLOGY: the network

DATA ANALYSIS: analyzing incoming reports



Quality control on reports:

- accurate GPS coordinates
- clear pictures
- realistic report content



Analysis of validated reports:

- which areas?
- when reported?
- which events?



Damage assessment

- which events cause which damage
- added value compared to national database?

META ANALYSIS: analyzing geo-observer's activities



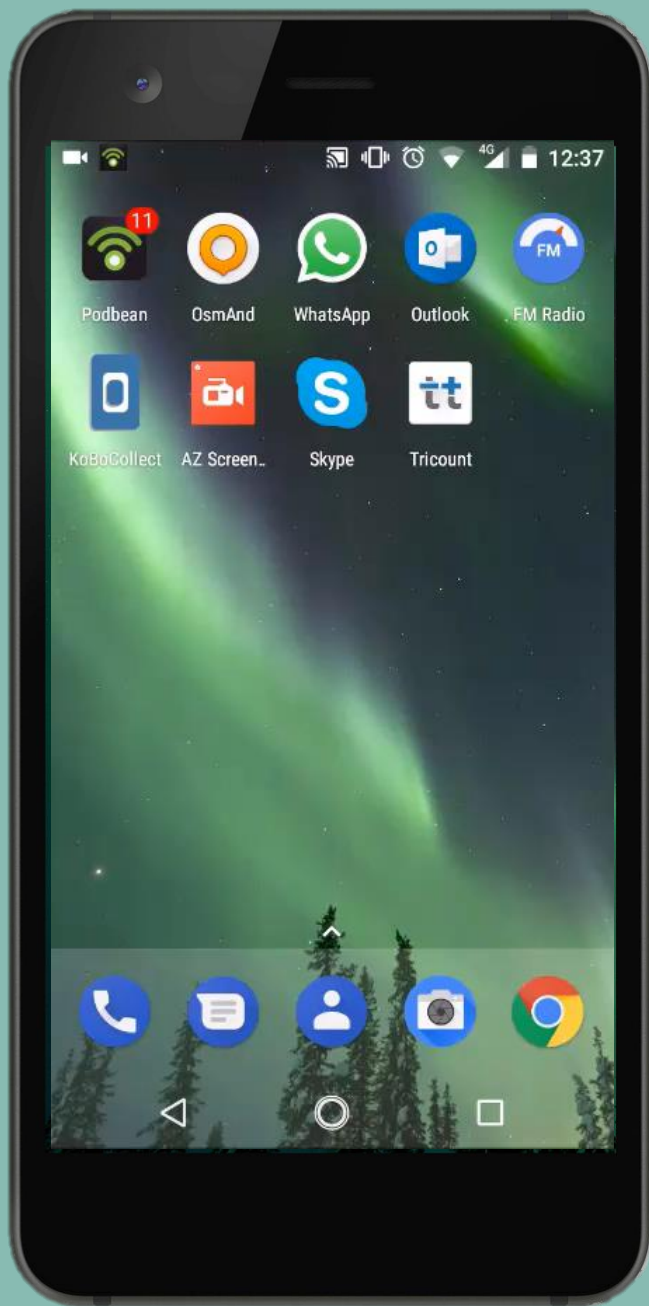
Check of bias:

- contribution bias
- spatial bias in reporting
- age bias

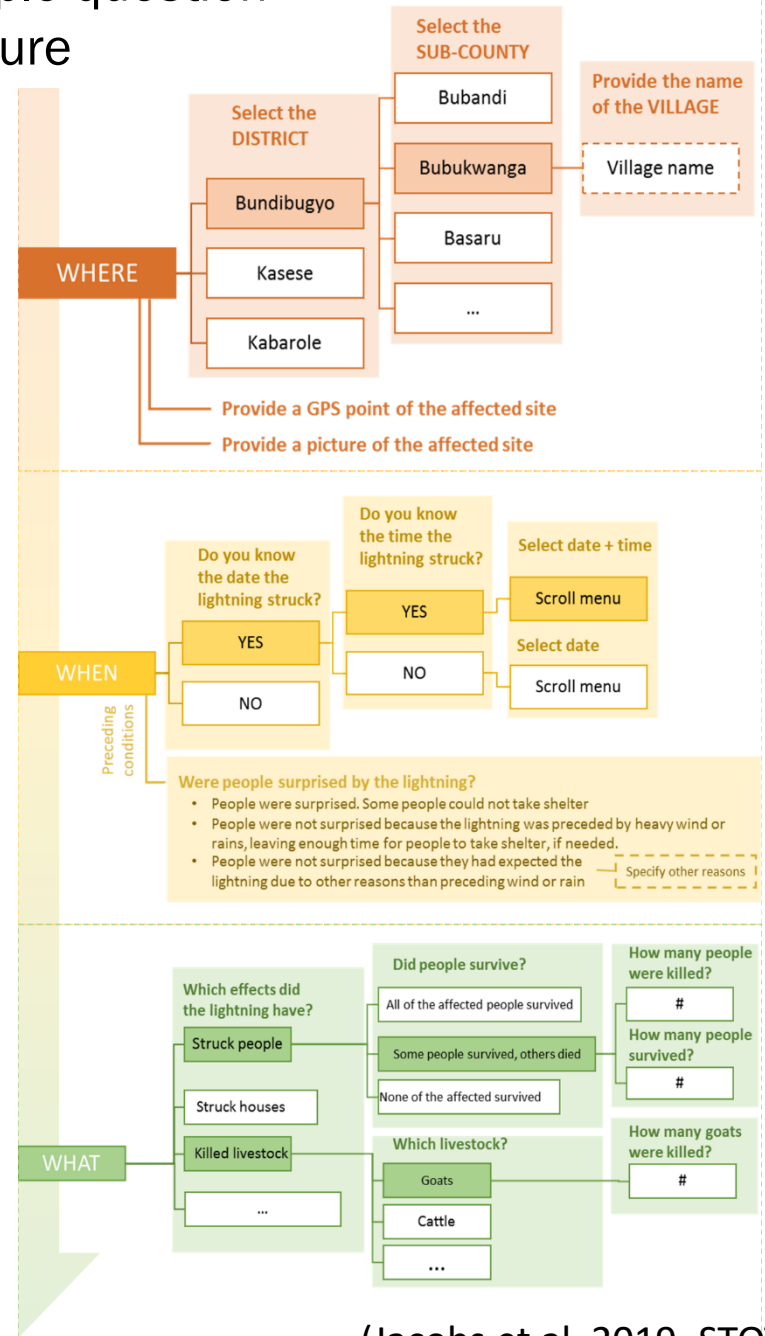


Survey among geo-observers

- major bottlenecks
- attitudes towards project
- motivations



Example question structure

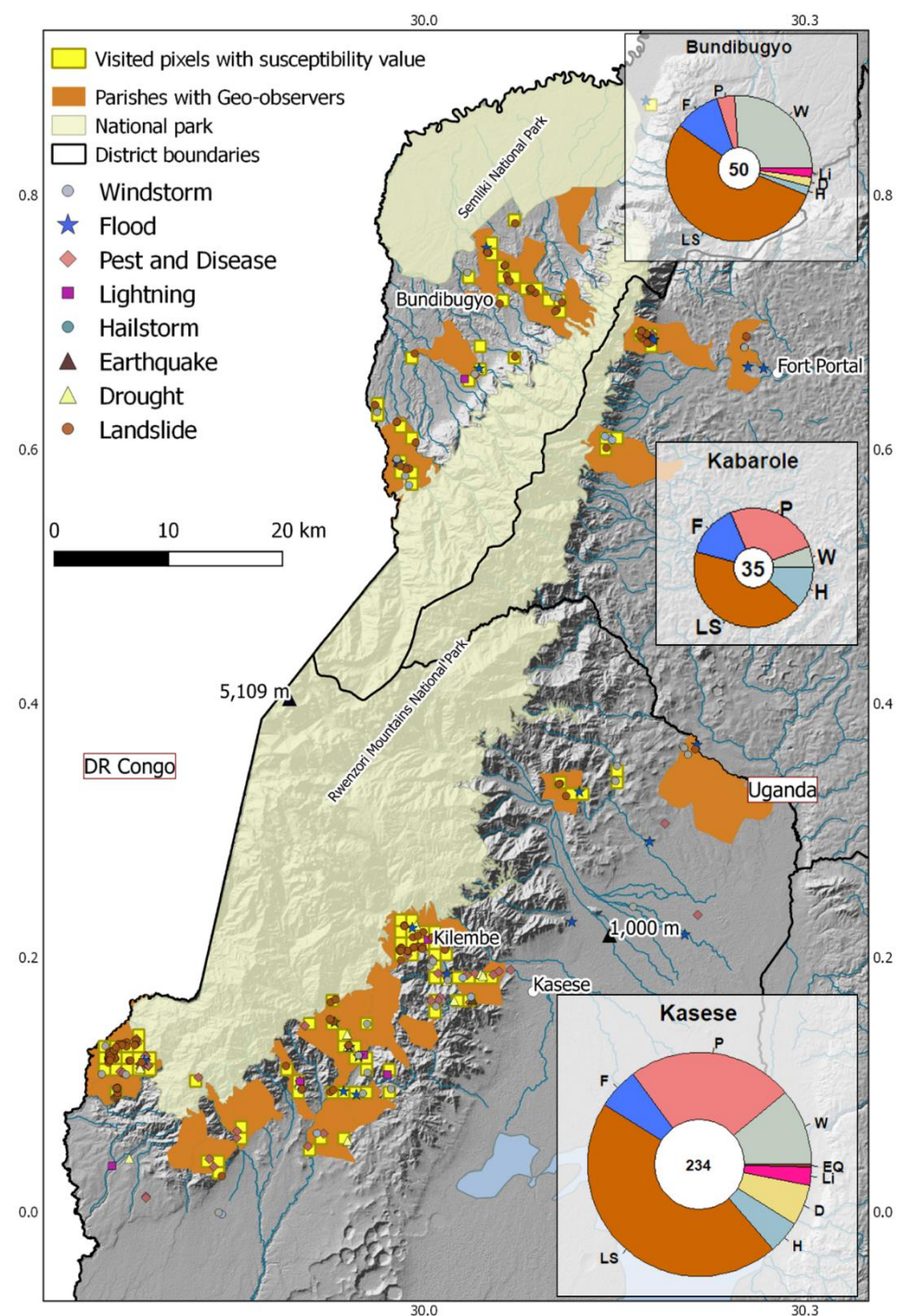


DATA ANALYSIS

FEB 2017-APRIL 2018

- 319 validated events
- predominantly landslides, pests-and diseases and floods
- Interactive content:

HERE or, here:

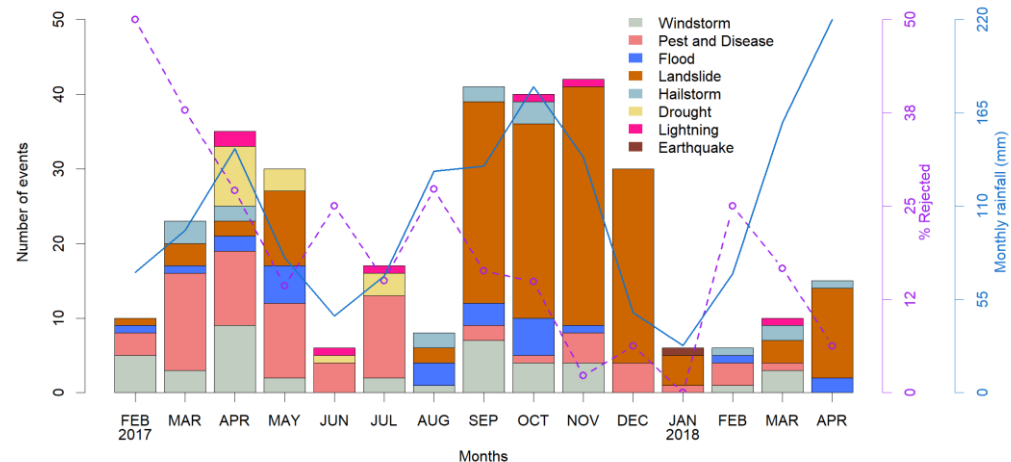


DATA ANALYSIS

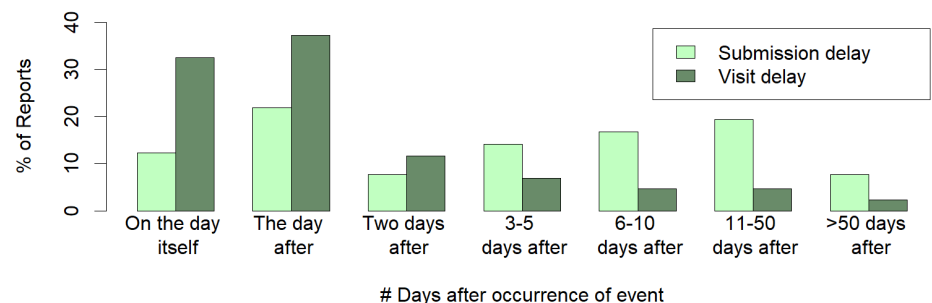
FEB 2017-APRIL 2018

- 319 validated events:
 - over time, quality increased
 - predominantly in rainy seasons
 - reporters reach the site 1-2 days after the event
 - reports send reports in days-weeks after event

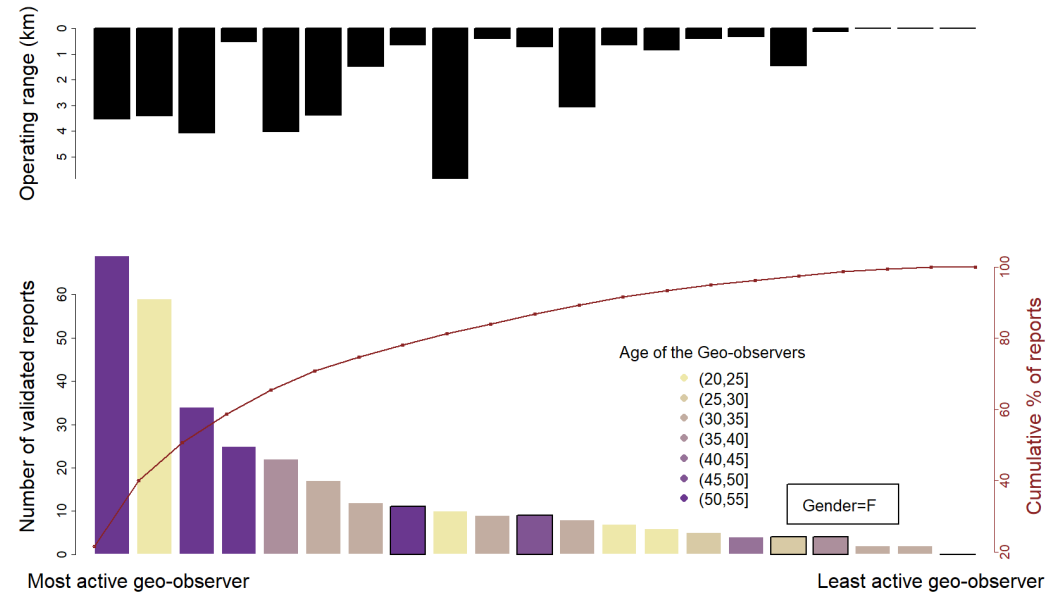
Validated reports over time:



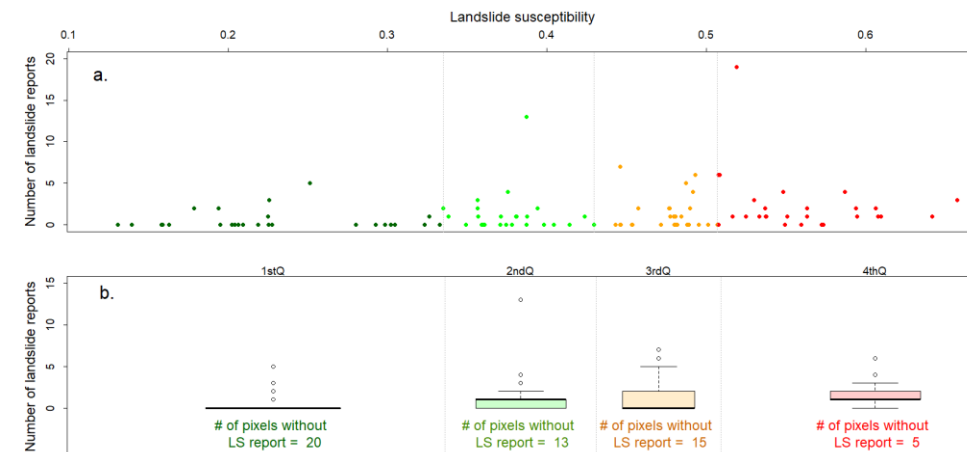
Visitation time and reporting delays:



Reporting contributions:



Spatial bias?



META-ANALYSIS REPORTER ACTIVITIES

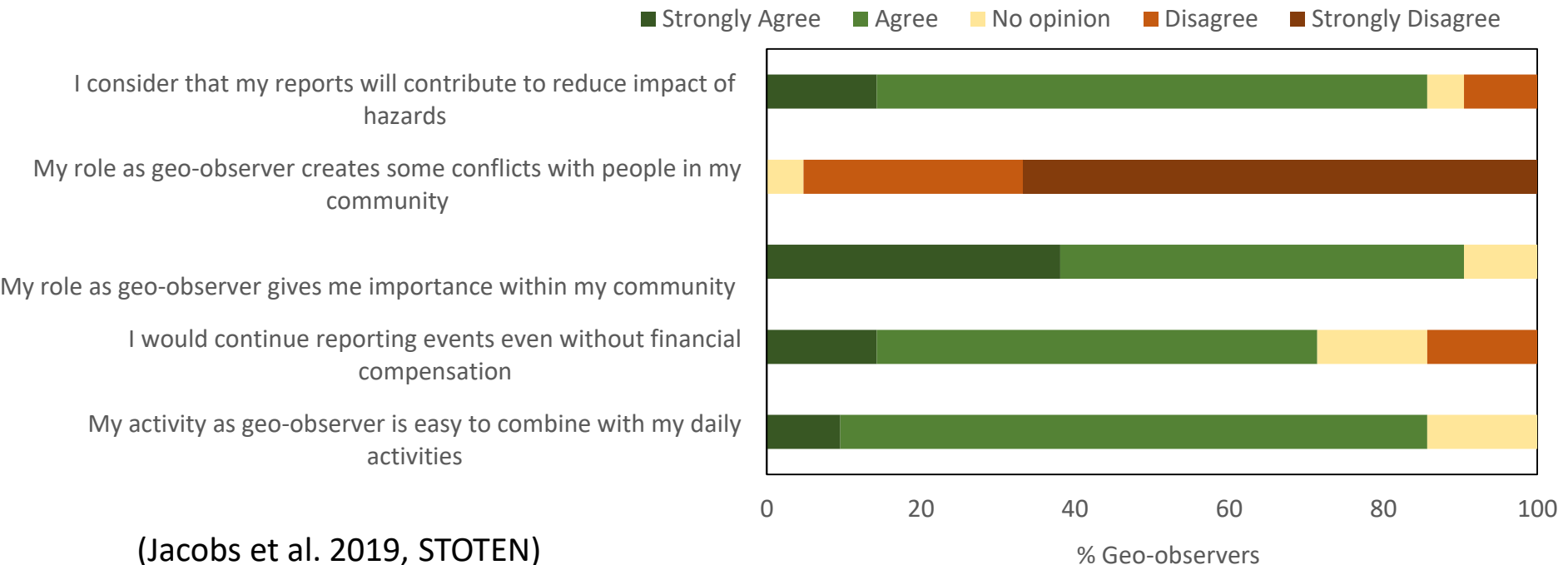
- one third of reporters contribute 75% of data
- older reporters report more
- more active reporters also travel further to report
- Higher landslide susceptibility = more landslide reports

(Jacobs et al. 2019, STOTEN)

META-ANALYSIS

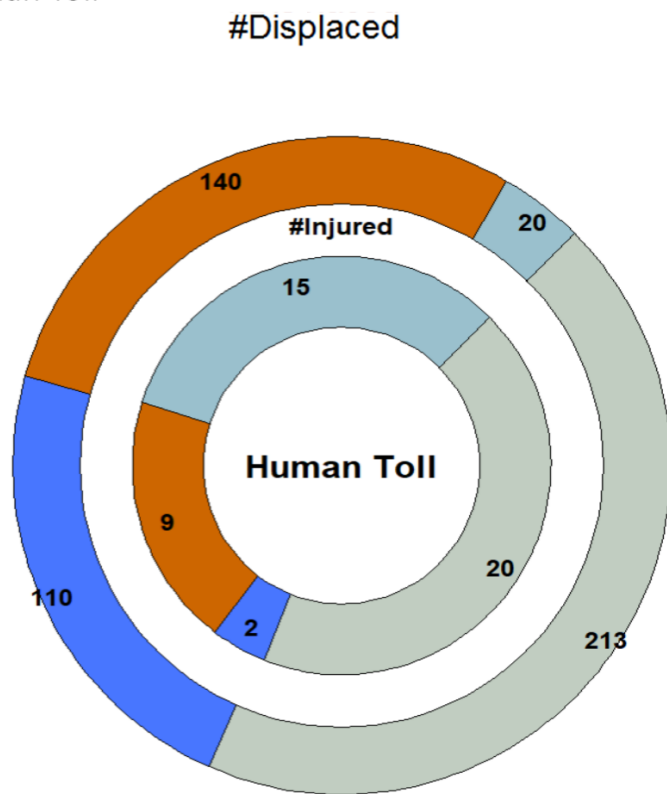
Survey of geo-observers

- geo-observers mostly driven by non-pecuniary incentives:
 - ➔ role in community, learning from reporting, contributing to research
- lack of events is major reason for non-reporting

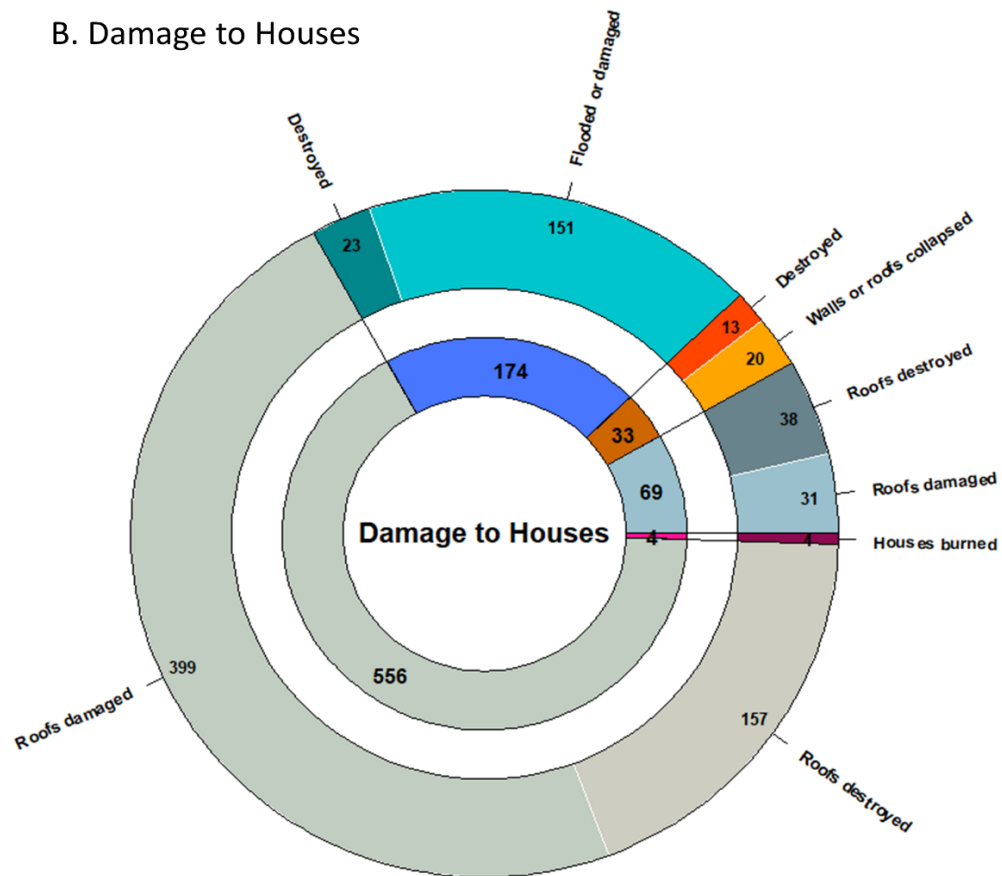


DAMAGE ASSESSMENT

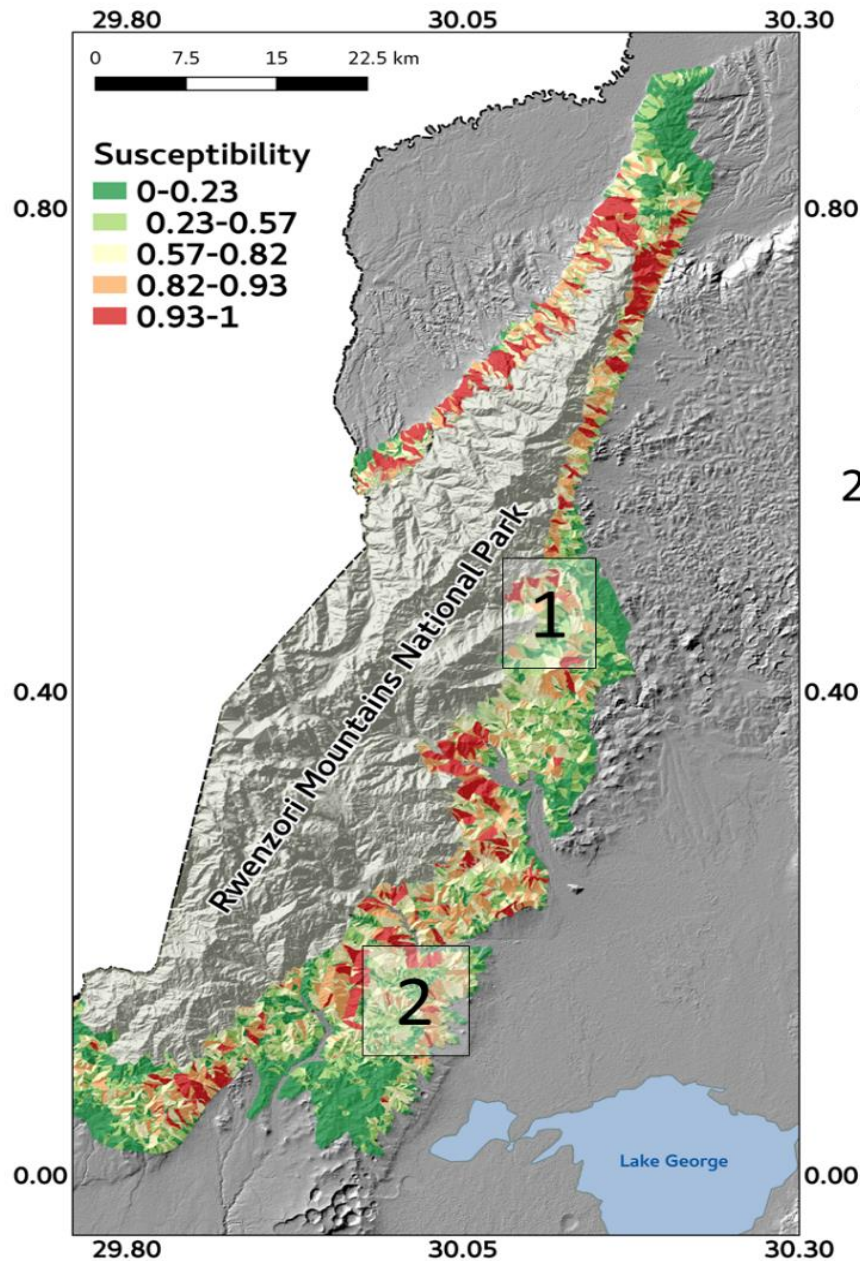
A. Human Toll



B. Damage to Houses

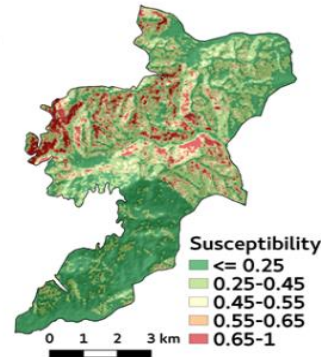


Landslide Susceptibility Modelling

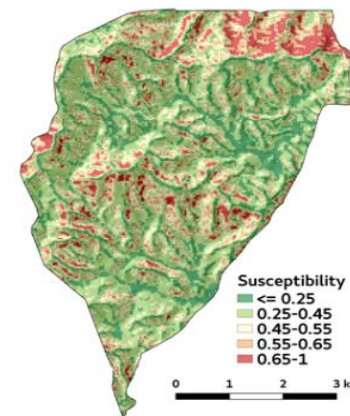


Pixel-based:

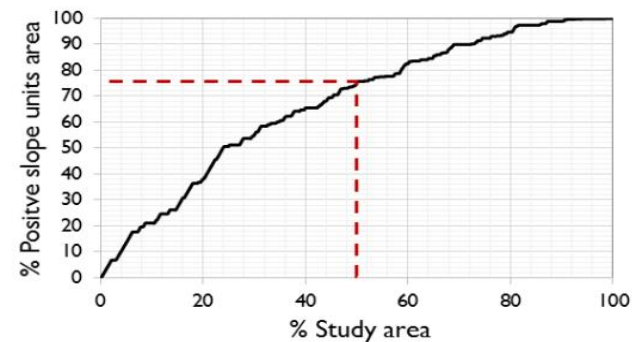
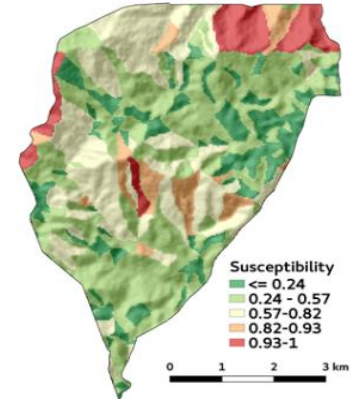
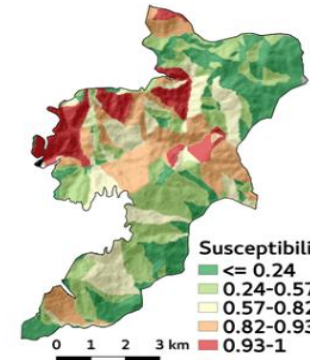
1



2

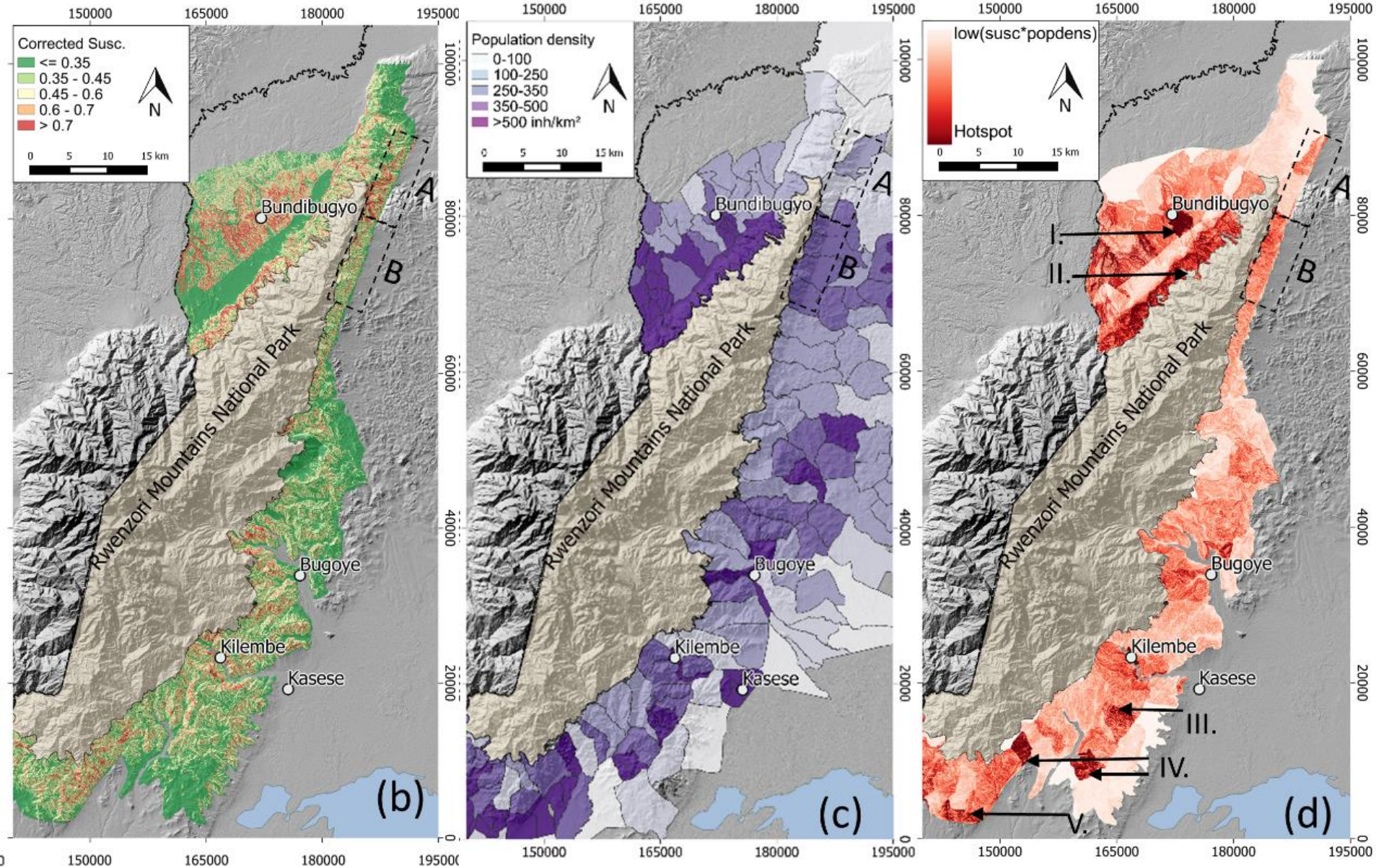


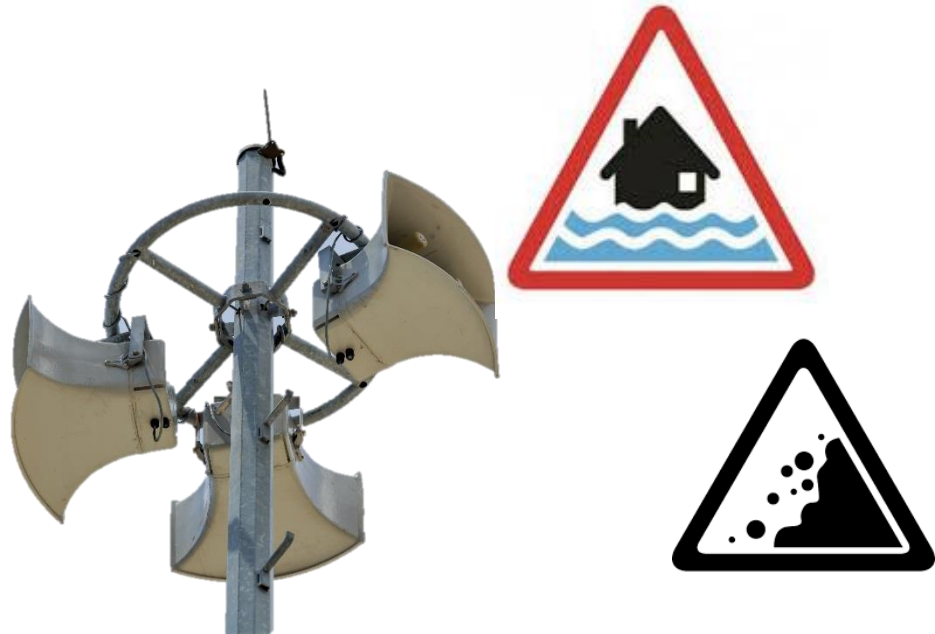
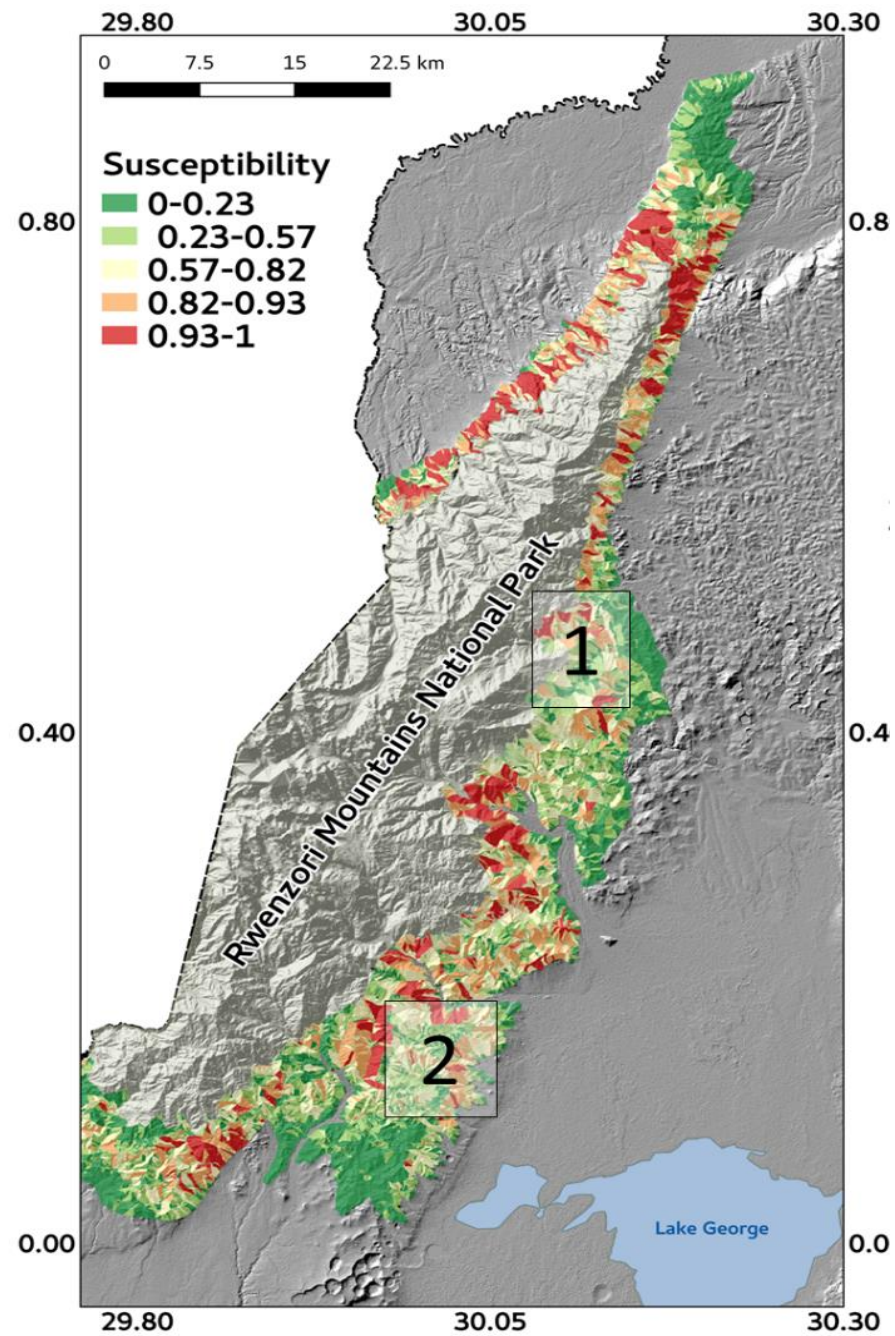
Slope unit-based:



(Jacobs et al. 2018, NHESS)

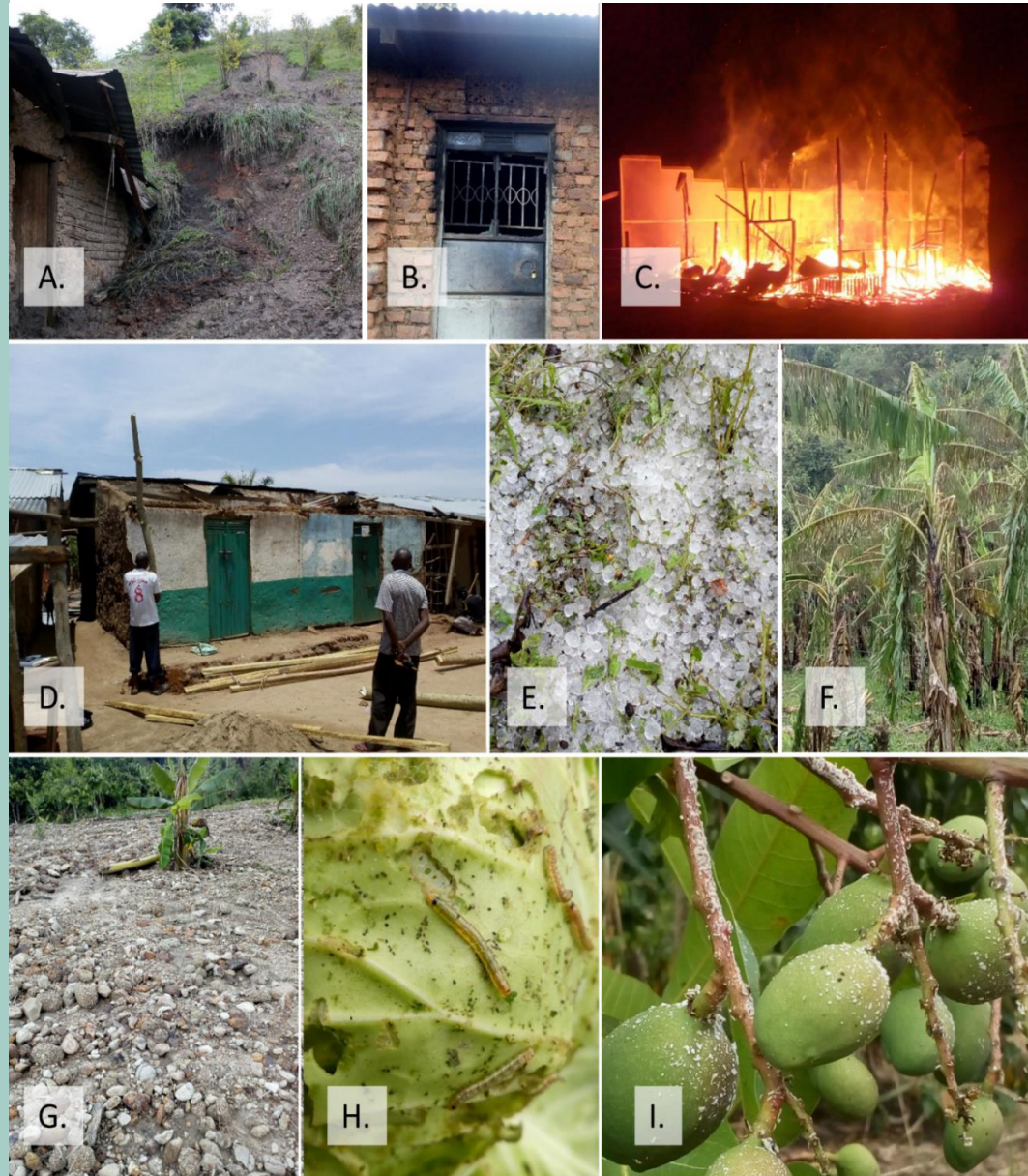
Landslide Exposure and Risk Modelling





CONCLUSIONS

- Citizen science provide on-site information on disasters with short response time
- Intrinsic motivation of geo-observers
- Citizen science as a way to raise DRR awareness
- Differences in motivation and/or mobility can lead to reporting bias
- High data density for a data poor region
- Issues of up-scaling, sustainability and data validation





Thank you

KU LEUVEN

ROYAL MUSEUM
FOR CENTRAL
AFRICA

Africa

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DE BRUXELLES