## A systematic assessment of uncertainties in large scale soil loss estimation from different representations of USLE input factors

A case study for Kenya und Uganda



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### Kenia and Uganda - an overview





#### a) Erosion Risk from Topography

- Very gentle inclinations (< 3°) first domain of sheet erosion Moderate to steep slopes (3° to 20°) - domains of active gully
  - erosion & growth Very steep slopes (> 20°) - prone to mass movement, severe

rain splash and sheet erosion

#### b) Mean Annual NDVI (2001-2018) [-]



#### c) Mean Annual Rainfall 1970-2000 [mm]





Data sources; SRTM 90m, MODIS MOD13Q1, MODIS Water Mask, WordClim Version 2, Natural Earth dataset

#### Study goals



- Quantification of uncertainties in soil loss estimation that result from the implementation of different realizations for USLE model inputs.
- Identification of the USLE model inputs that contribute the most to the uncertainties in the soil loss estimates.
- Comparability of soil loss estimates to in-field soil loss data.

### Workflow I – Development of USLE input realizations





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C<sub>MODIS LC</sub>, Monfreda

Panagos et al. (2015)

#### Workflow II – Soil loss estimation and analyses



- 1 Schematic illustration of the USLE input factor realizations. Each dot represents one realization. The color groups them based on the used methods.
- 2 All input realizations are combined to USLE models and the soil loss is estimated for Uganda and Kenya spatially distributed on a 90m grid.
- 3 All 756 USLE combinations are analyzed in each grid cell and aggregated on administrative level.



### Statistical analysis of spatially distributed soil losses





- > Model mean "plausible"
- Ranges in soil loss (uncertainty) exceed the model mean by up to one order of magnitude.

### Soil loss classification: class frequencies in model ensemble





- Almost the entire model ensemble predicts a tolerable soil loss for a large part of study area.
- Steep and complex topographies show strong disagreement between the USLE model realizations.
- Very locally larger numbers of model realizations predict severe soil loss.

### Spatial analysis of dominant soil loss classes





> Patterns of dominant soil loss classes follow the patterns of topography and vegetation.

### Spatial analysis of the most influential USLE inputs





> Vegetation in humid and vegetated regions, and soils in dry regions show large patterns of greater importance.

#### Soil loss on the administrative level



**يa**) Uganda 22 Kyoga 11-16 Kenya 23 Mt. Elgon Nile 17 (4321 m uwenzor Mt. Kenya \*Kampala 109 m(5199 m) 25 26 18 24 20 Nairobi 27 N Kilimanjaro 5895 m)

#### a) Erosion Risk from Topography

- Very gentle inclinations (< 3°) first domain of sheet erosion</li>
  Moderate to steep slopes (3° to 20°) domains of active gully erosion & growth
- Very steep slopes (> 20°) prone to mass movement, severe rain splash and sheet erosion



#### Comparison to in-field data – Farm compounds

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Small scale but long-term soil losses estimated by De Meyer et al. (2011) are substantially larger than model ensemble predictions.

### Comparison to in-field data – Sediment yields from catchments





- Small scale but long-term soil losses estimated by De Meyer et al. (2011) are substantially larger than model ensemble predictions.
- Short-term sediment yield records are lower or in a comparable range to the model ensemble.

#### Summary and conclusions



- The estimation of soil loss with the USLE involves large uncertainties that result from the selected methods to calculate the USLE inputs.
- Steeper and more complex topographies, sparsely vegetated areas show an increased soil loss, but also substantially larger uncertainties.
- > Uncertainties in the C and K factors are relevant on larger scales. The uncertainties introduced by the LS factor shows very small scale patterns.
- > A comparison to single model analyses illustrates the relevance of a comprehensive uncertainty analysis with a USLE ensemble.
- > A comparison of soil loss estimates to in-field data is limited, due to temporal and spatial constraints, but also due to differences in the measured entities.

# **Further Questions?**

This presentation summarizes results that are presented in the manuscript:

Schürz, C., B. Mehdi, J. Kiesel, K. Schulz, and M. Herrnegger (in review, 2019) A systematic assessment of uncertainties in large scale soil loss estimation from different representations of USLE input factors - A case study for Kenya and Uganda, In: Hydrol. Earth Syst. Sci. Discuss., doi: 10.5194/hess-2019-602



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