# **ETH** zürich

### Weather and Climate Risks

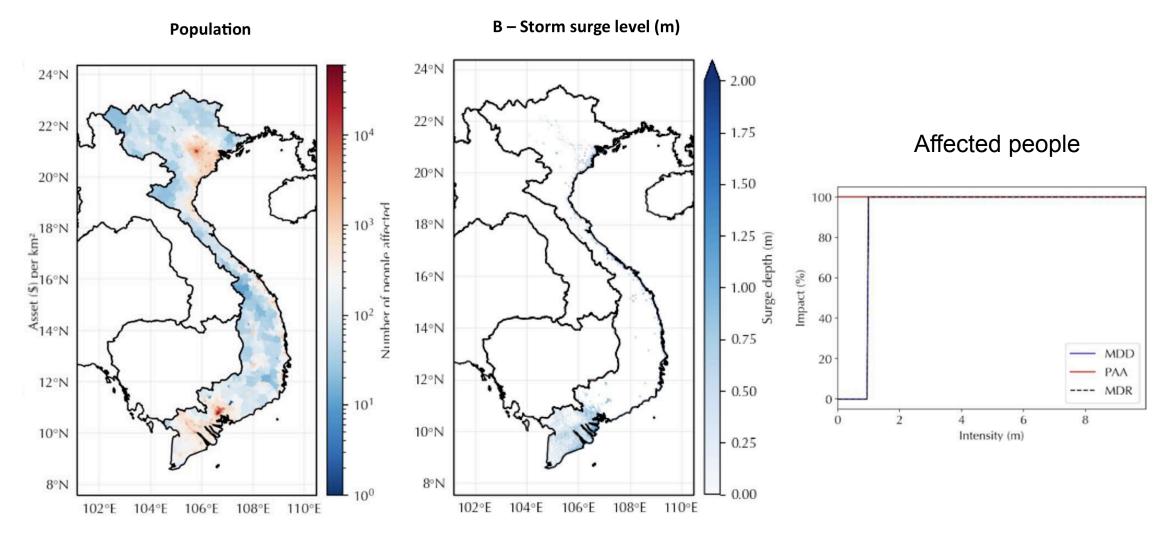


Uncertainty and sensitivity analysis for climate impact assessment and adaptation options appraisal modeling with CLIMADA

> Cross-sectoral ISIMIP and PROCLIAS Workshop 16-19 May 2022 Chahan Kropf



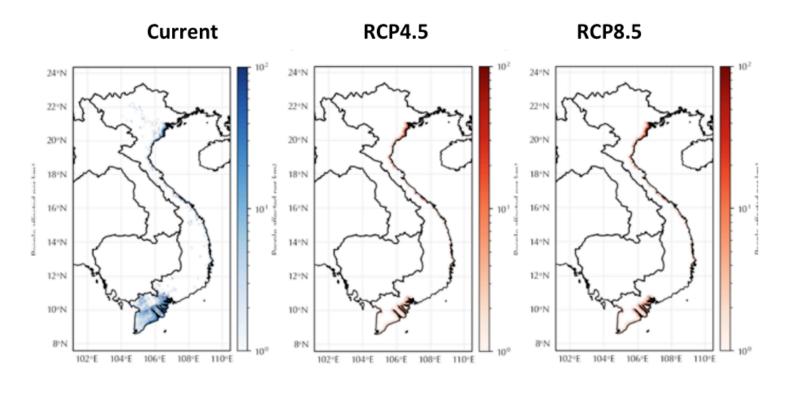
### **Example - storm surge in Vietnam : Input variables**



>10000 events

### **Example - storm surge in Vietnam : Impact**

- Tropical cyclone induced surge
- Current 2020 scenario and additional damage in the future 2050 scenarios under RCP4.5 and RCP8.5.
- Population growth, climate change, sea level rise



• Average annual impact:

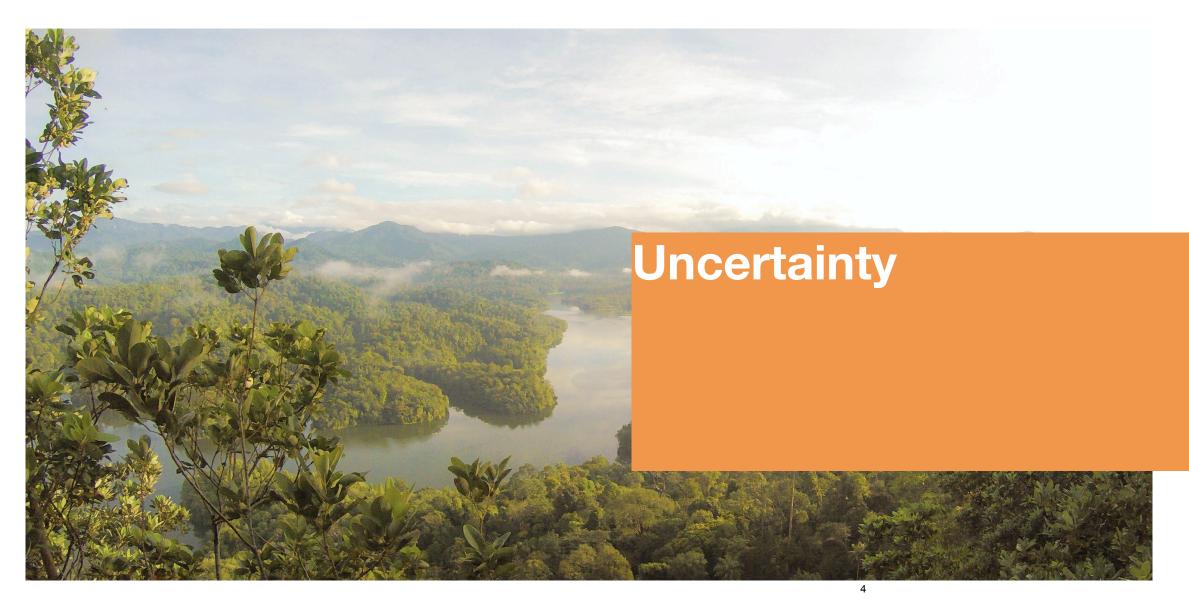
1.94Mio.

+1.08Mio.

+2.16Mio.







### "Uncertainties" in risk models



• Epistemic: "imperfect understanding of the system"<sup>1</sup>



• Aleatoric: "uncertain due to system properties"<sup>1</sup>



• **Predictive:** "different model specifications might seem equally plausible and it is unclear how to best represent the target system for a specific purpose."<sup>1</sup>



• **Normative:** "uncertainty about value itself"<sup>2</sup> and "uncertainties about how to decide and how to act"<sup>3</sup>

1- DOI:10.1016/j.envsoft.2020.104754 3- DOI:10.1093/acprof:oso/9780199964482.003.0003 <u>3. https://doi.org/10.1017/S0266267121000201</u>

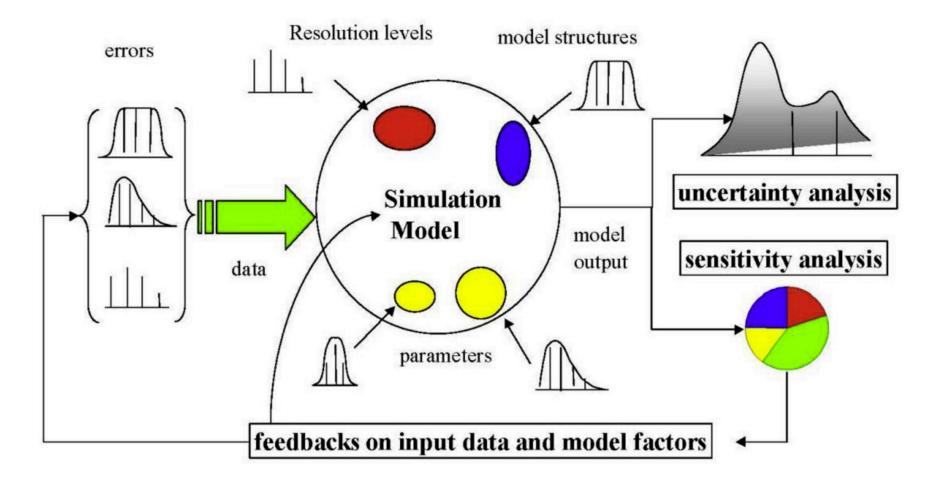


### **Quantitative? Not all**

- Depends on
  - case study
  - ► purpose
- Subset of uncertainties
- General treatment with argument analysis framework<sup>1</sup>

1- In prepatation: Analyzing Uncertainties in Climate Risk Assessment and Adaptation Options Appraisal with a Four-Phase Analytical Framework

## Quantitative: uncertainty and sensitivity analysis with quasi-Monte-Carlo sampling





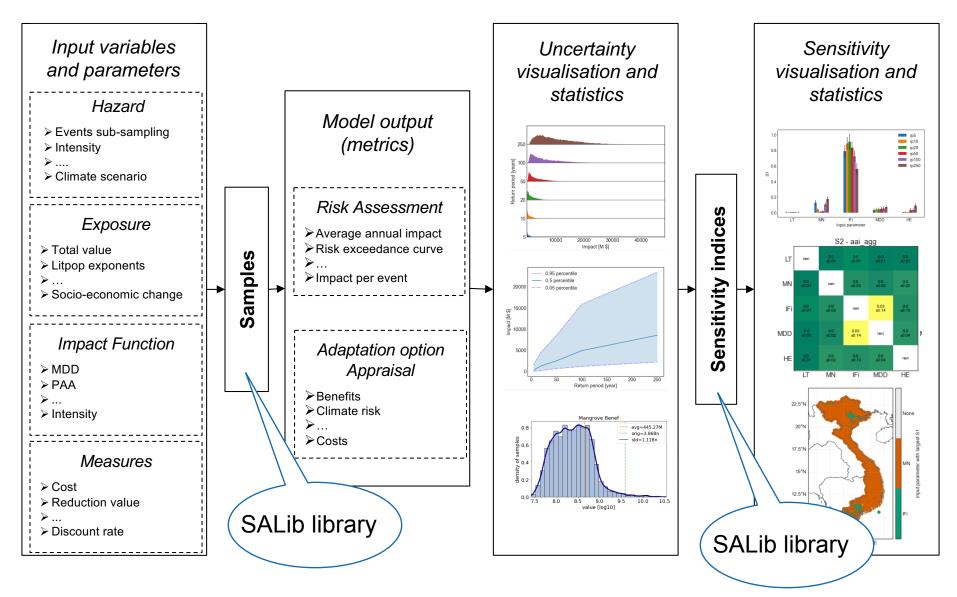


# unsequa module

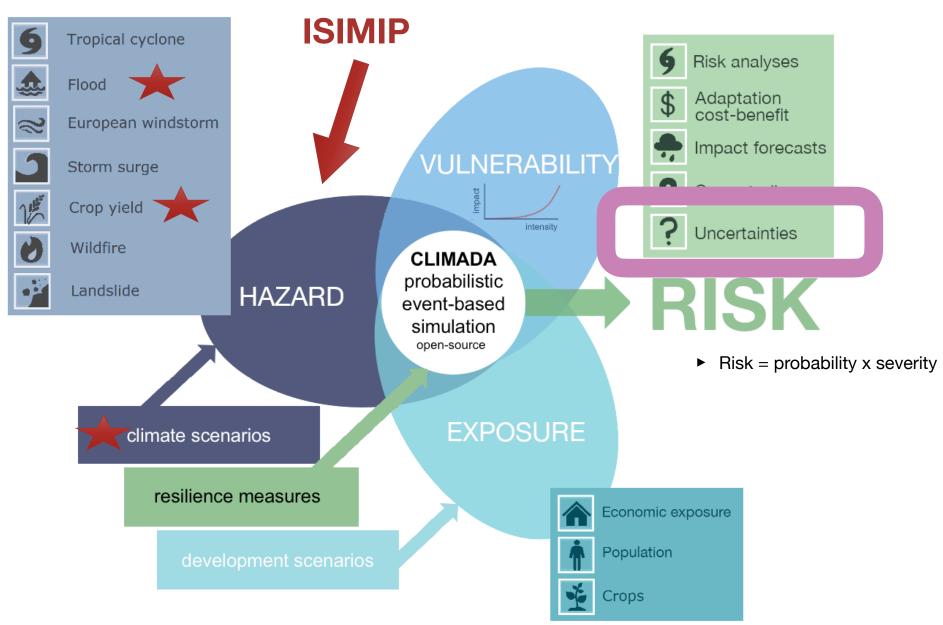
Module for uncertainty and sensitivity quantification in CLIMADA

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## unsequa (uncertainty sensitivity quantification) module

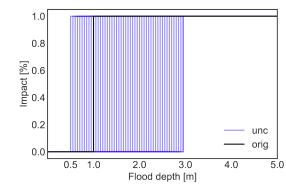


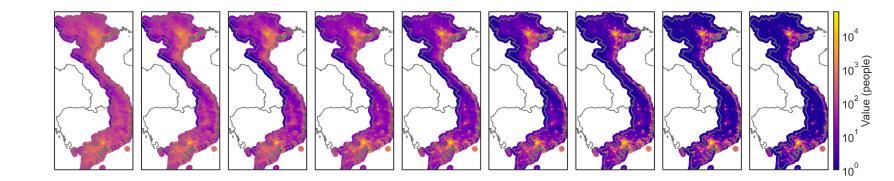
### CLIMADA v3.1.2



### **Case study: Vietnam storm surge current risk**

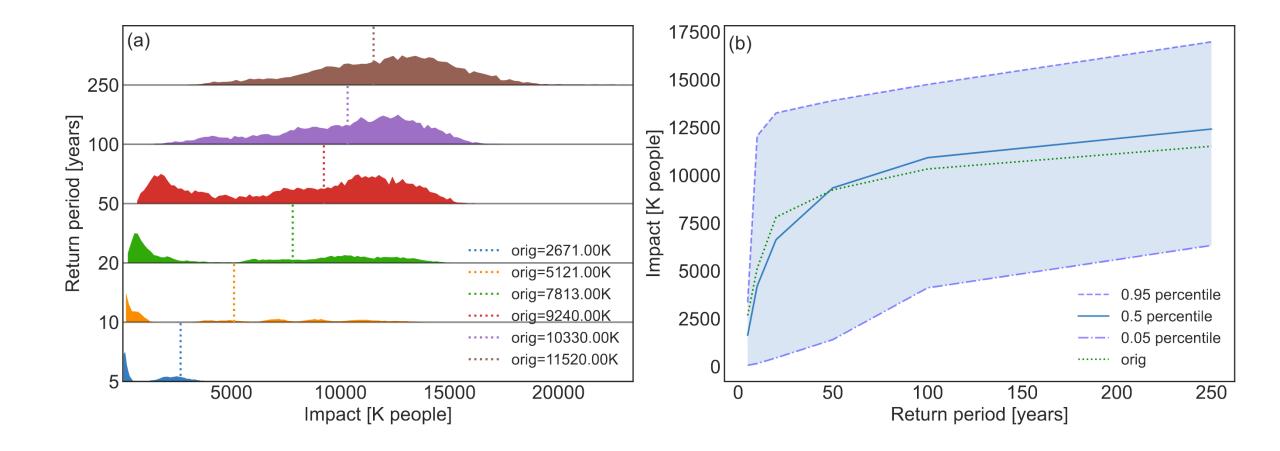
Risk assessment				
Exposures	total value	Т	truncated Gaussian multiplicative	clip:[0.9, 1.1] ; $\mu$ : 1, $\sigma$ : 0.05
	spatial distribution	L	LitPop layers exponents	$m \in (0, 0.5, 1); n \in (0.75, 1, 1.25)$
Hazard	event set bootstrapping	Η	re-sampling the event set with replacement	
Impact function	threshold shift	S	uniform range	[0.5m, 3.0m]



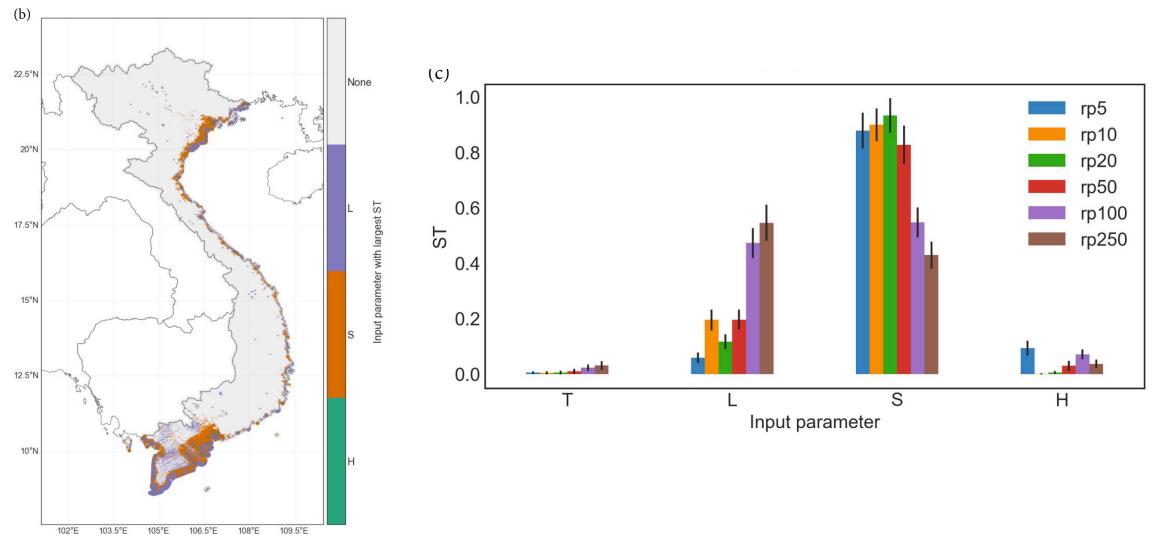




### **Impact uncertainty**

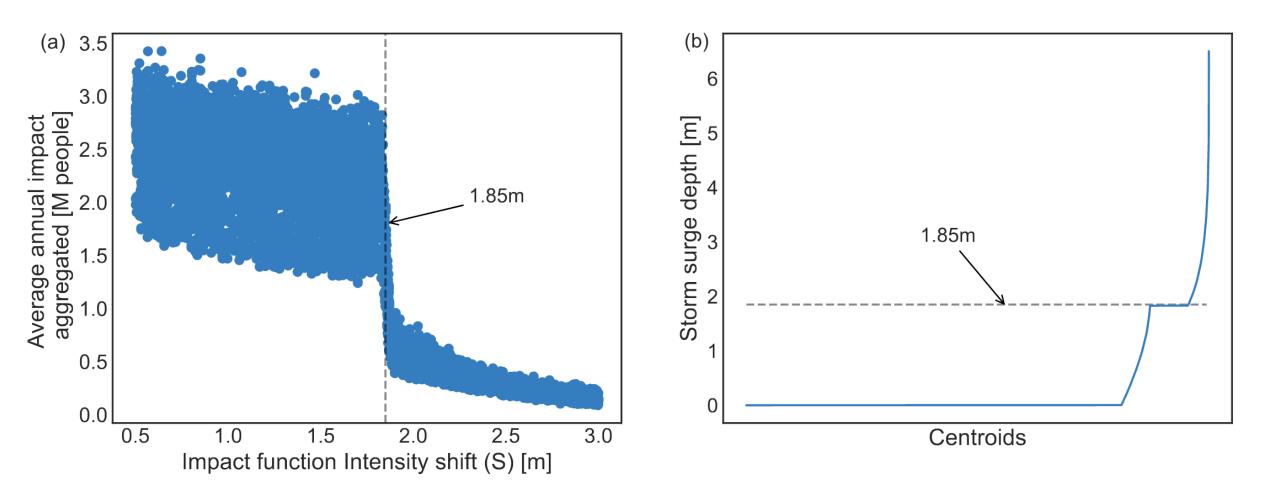


## Sensitivity



Taken from: https://gmd.copernicus.org/preprints/gmd-2021-437/

### **Bi-modal distribution**



Taken from: https://gmd.copernicus.org/preprints/gmd-2021-437/







### **Caution on Uncertainty Quantification**

- Be vary of false certainty from uncertainty!
- You need to have a good idea of the uncertainty of the uncertainty.

- For **ISIMIP**-data, good **uncertaint**y estimates are sometimes **missing**
- How to treat **ensembles** combining different models, different scenarios?
  - "Equi-probably" ensemble of opportunity?
- Uncertainty requires more than hazard model.
- Uncertainty of **modelling choices** are rarely reported

## Thank you!

Kropf, C. M., Ciullo, A., Otth, L., Meiler, S., Rana, A., Schmid, E., McCaughey, J. W., and Bresch, D. N.: Uncertainty and sensitivity analysis for probabilistic weather and climate risk modelling: an implementation in CLIMADA v.3.1.0, 1–32, <u>https://doi.org/10.5194/gmd-2021-437</u>, 2022.







# **Supplementary slides**



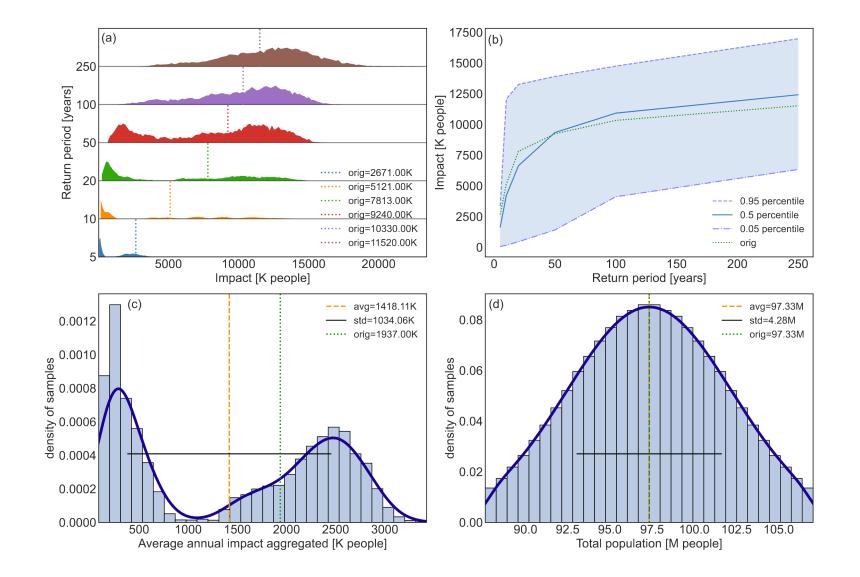
### **Uncertainty vs Sensitivity analysis**

• **Uncertainty**: distribution of output metric due to a distribution of input variables/parameters

• **Sensitivity**: attribution of the output metric variation to the input variable/parameters

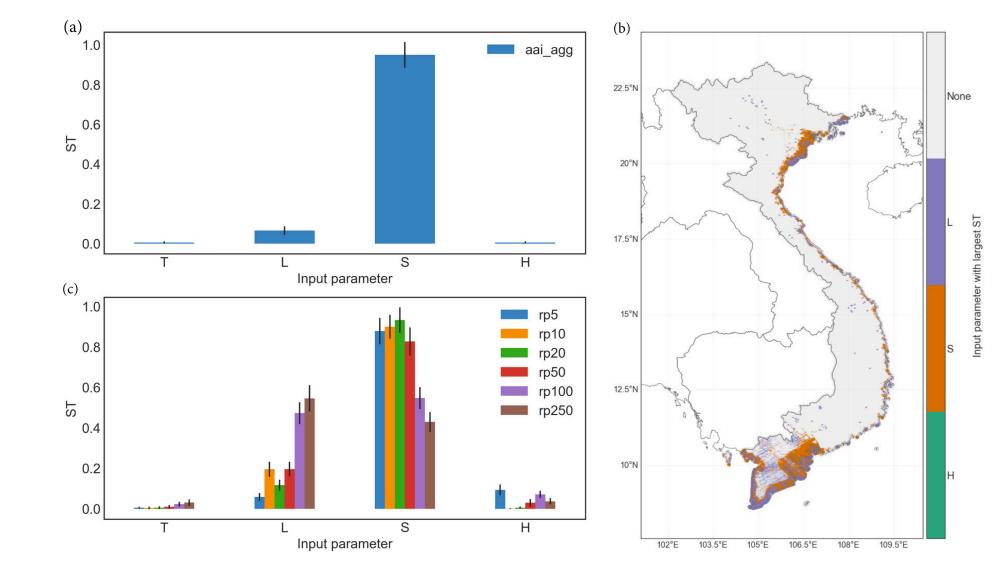
► NOT: attribution of contribution to output metric from input variable/parameter

### **Impact uncertainty**

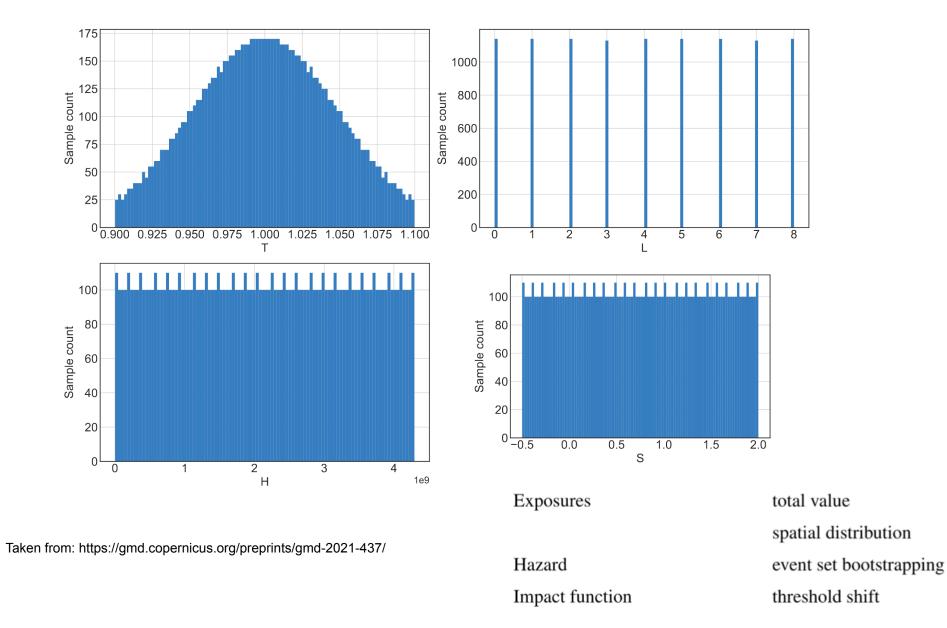


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## Sensitivity



### **Global sampling with Sobol sequence**



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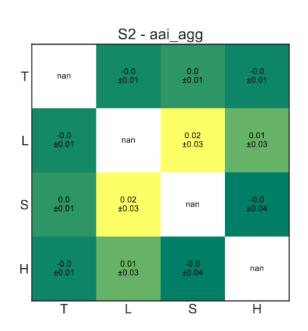
### Weather and Climate Risks

Summary

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- Unsequa module for uncertainty and sensitivity analysis of climate risk
- Fully integrated and flexible
- Several helper methods for quick and easy analysis
- All input variables can have uncertainty
- Quasi-Monte Carlo sampling from SALib library
- Sensitivity analysis from SALib library
- Future development: allow for the use of surrogate model

Kropf, C. M., Ciullo, A., Otth, L., Meiler, S., Rana, A., Schmid, E., McCaughey, J. W., and Bresch, D. N.: Uncertainty and sensitivity analysis for probabilistic weather and climate risk modelling: an implementation in CLIMADA v.3.1.0, 1–32, <u>https://doi.org/10.5194/gmd-2021-437</u>, 2022.





### **Global vs local sampling**

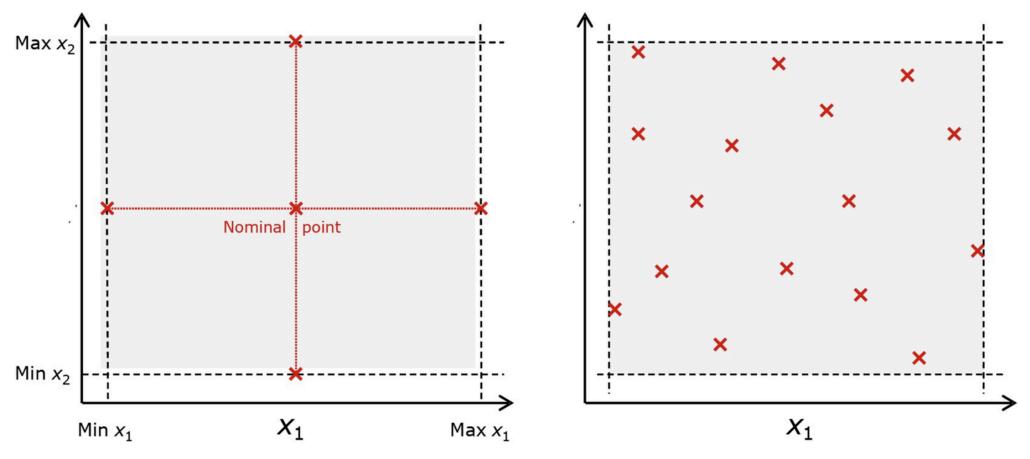
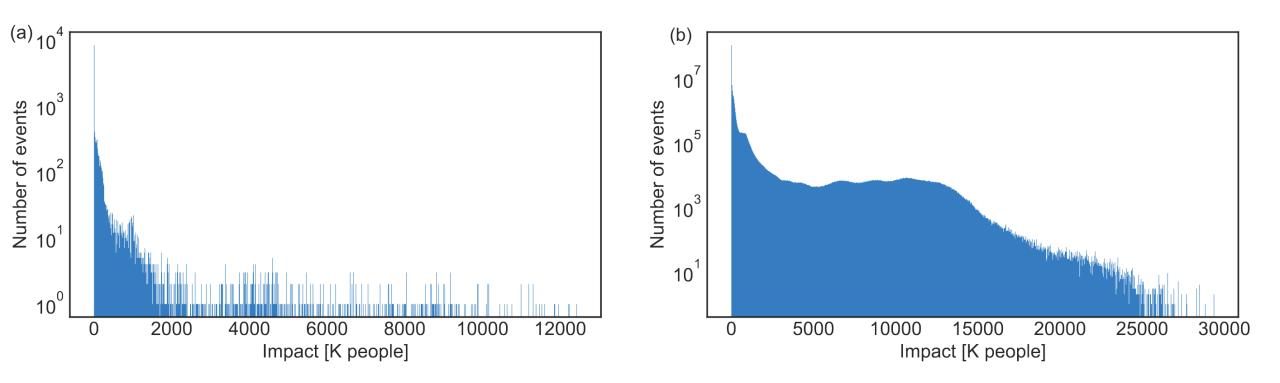


Fig. 2. OAT design (left) contrasted against global design (right).

• For non-linear models (like CLIMADA), always use Global sampling

Taken from: DOI:10.1016/j.envsoft.2019.01.012

### Impact over all events form sampling



### **Core impact computation**

- Exposures value at given location : E(x)
- Intensity of hazard event epsilon at modelled location  $\hat{\mathcal{X}}$  closest to  $\, x : \, h_\epsilon( ilde{x}) \,$
- Frequency (probability) of event :  $\, {\cal V}_{\epsilon} \,$
- Impact function of exposures at location  $\mathcal{X}$  :  $f(E(x)) = f_x$
- Impact matrix:

$$I_{\epsilon,x} = f_x(h_\epsilon(\tilde{x}))E(x)$$



### Impact and risk metrics

• Impact at event

$$I_{\epsilon} = \sum_{x} I_{\epsilon,x}$$

• Expected average impact at exposures

$$\overline{I}_x = \sum_{\epsilon} I_{\epsilon,x} \nu_{\epsilon}$$

• Average impact over all exposures and all events (total annual expected risk)

$$\overline{R} = \sum_{\epsilon, x} I_{\epsilon, x} \cdot \nu_{\epsilon}$$