



Globally consistent, open-source river flood impact model using open data

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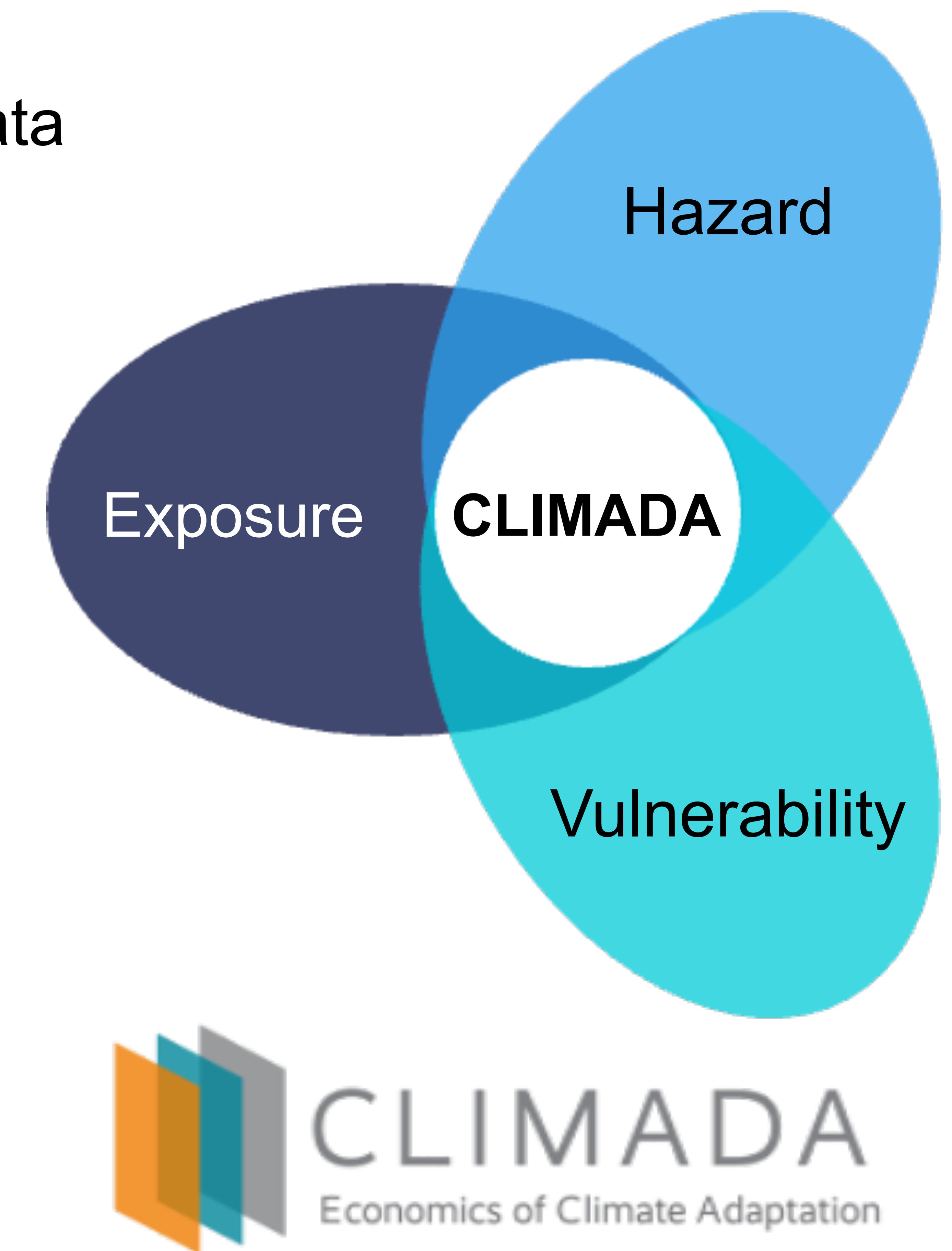
Project Overview

Goals

- Develop open-source **river flood inundation** model based on open data
- Forecast river flood **impacts** on short- to medium-range time scale
- Support **anticipatory action** and disaster relief efforts

Data Sources

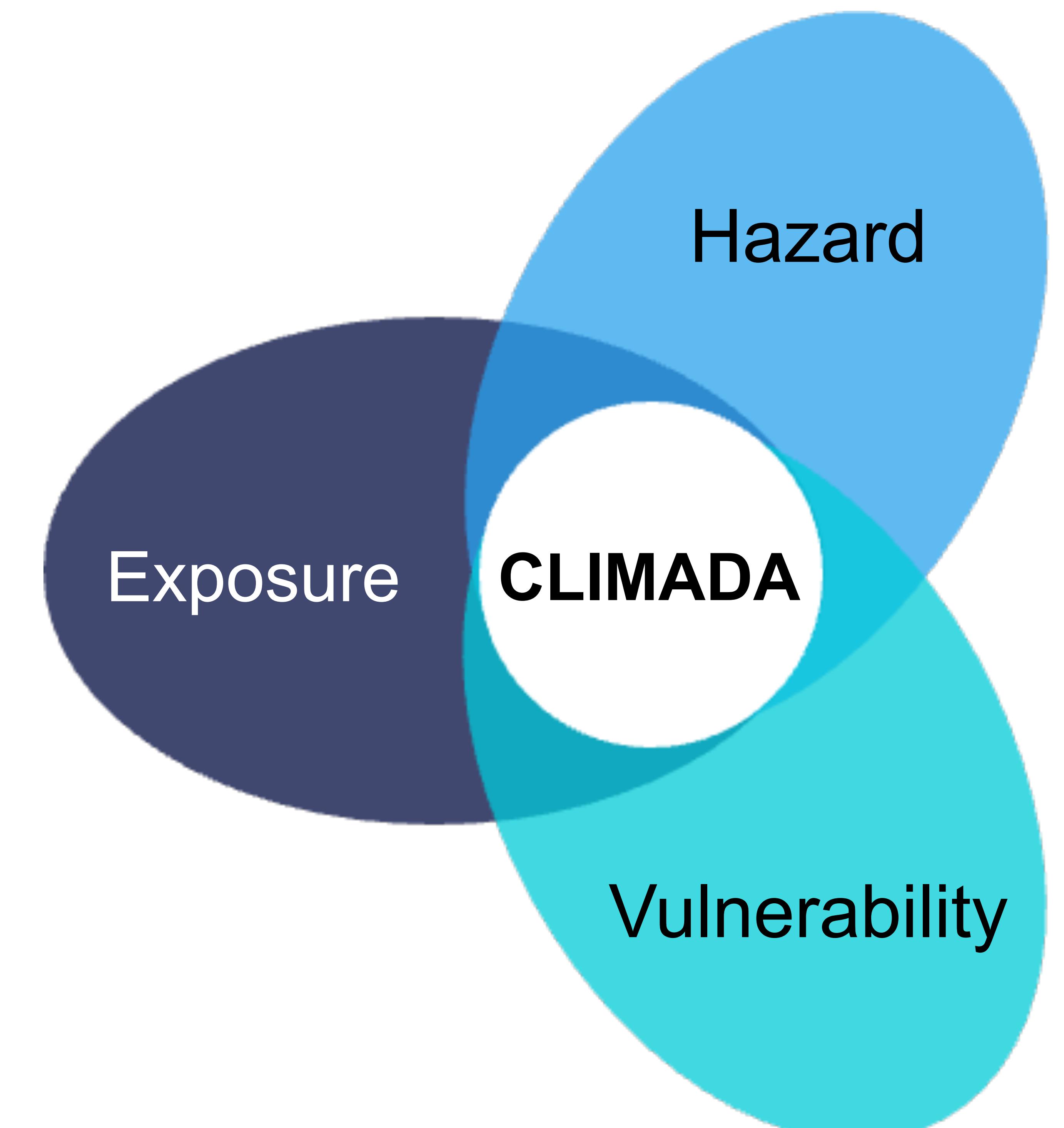
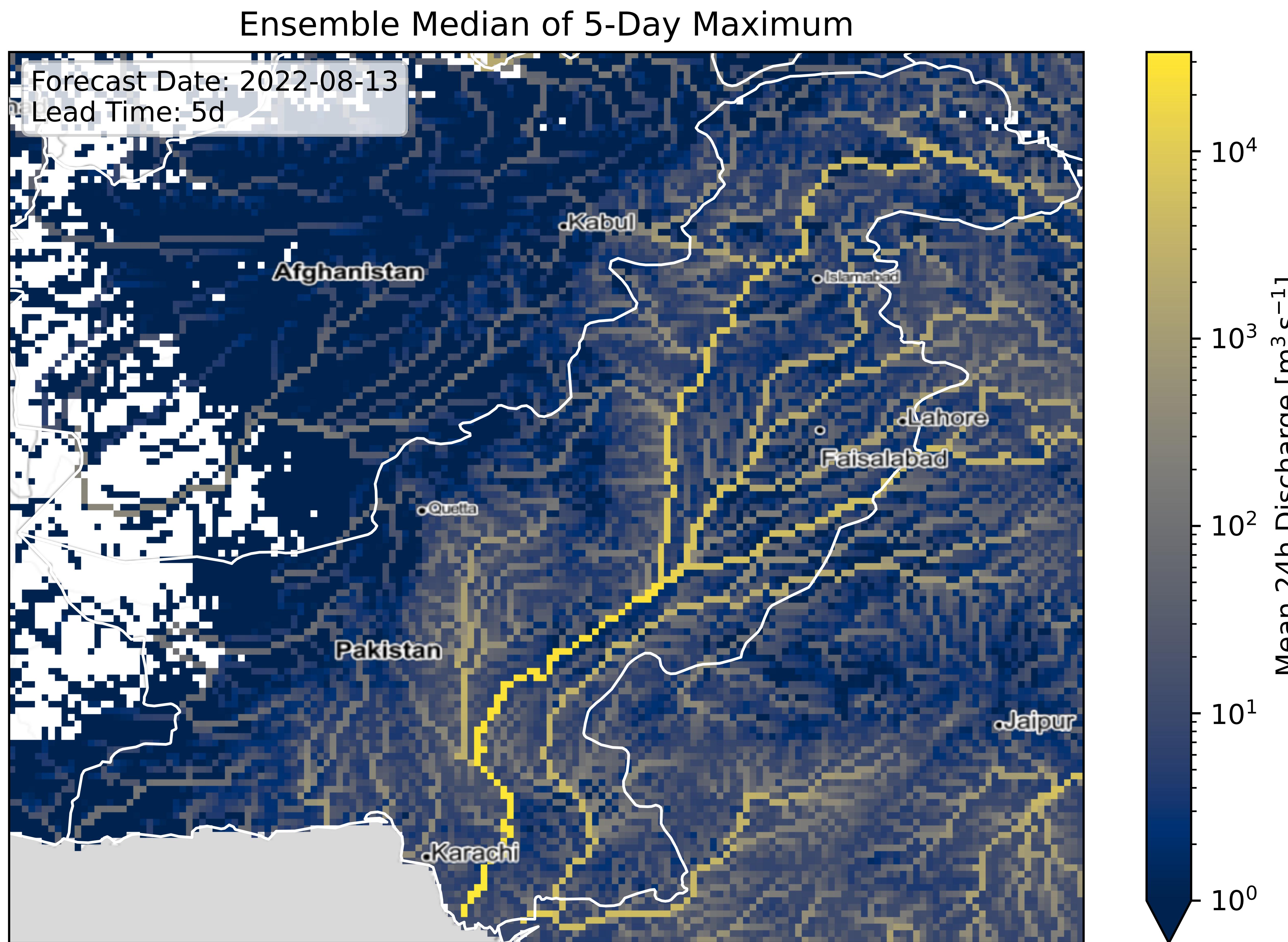
- River discharge (GloFAS, Copernicus Climate Data Store)
 - Daily ensemble forecasts
 - Daily reanalysis data from 1980 onwards
- Flood hazard maps (EC JRC Data Catalogue)
- Gridded Population of the World (GPW), NASA Black Marble
- FLOPROS flood protection database



Flood Impact: Potential Displacement of Population

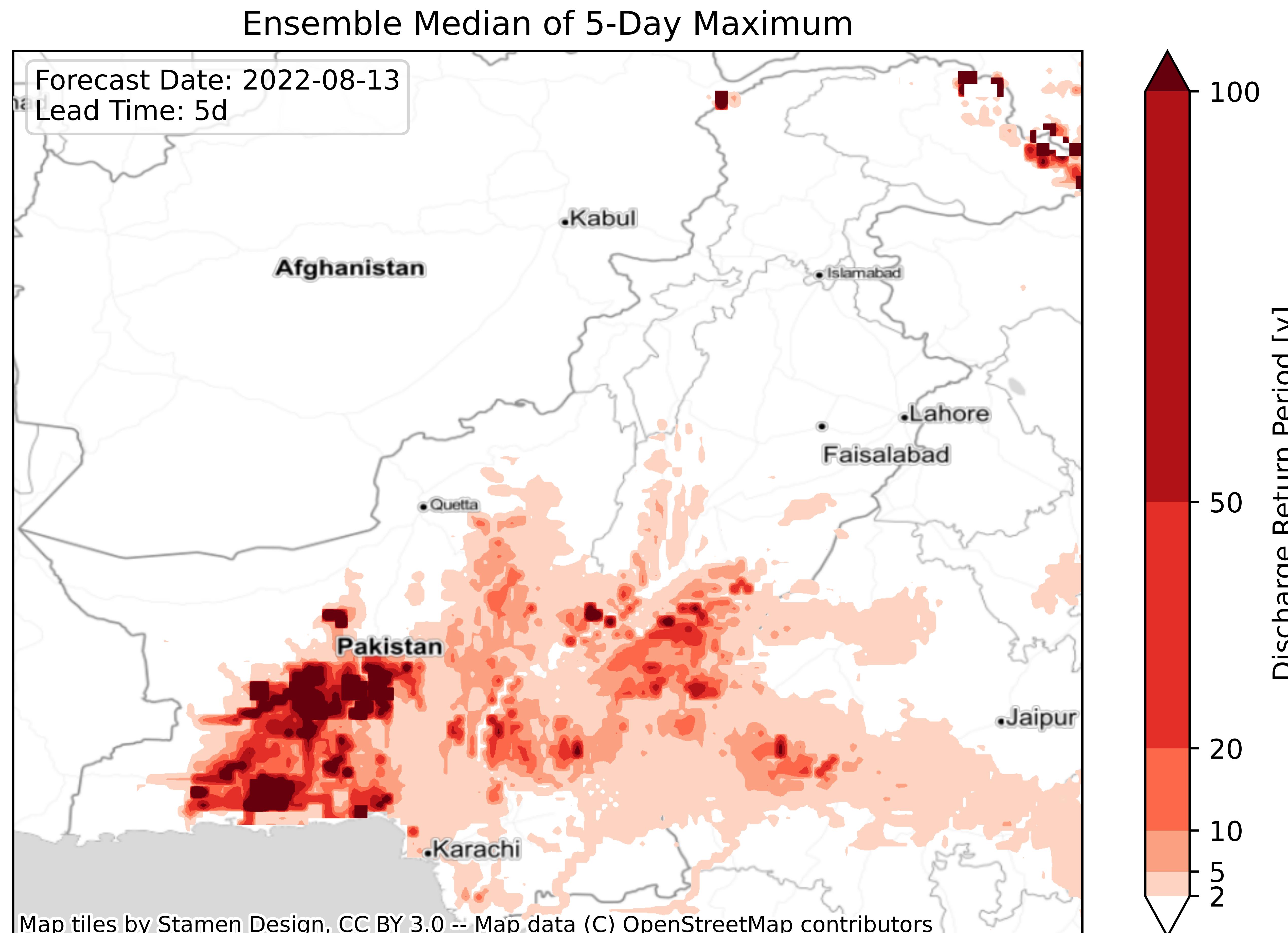
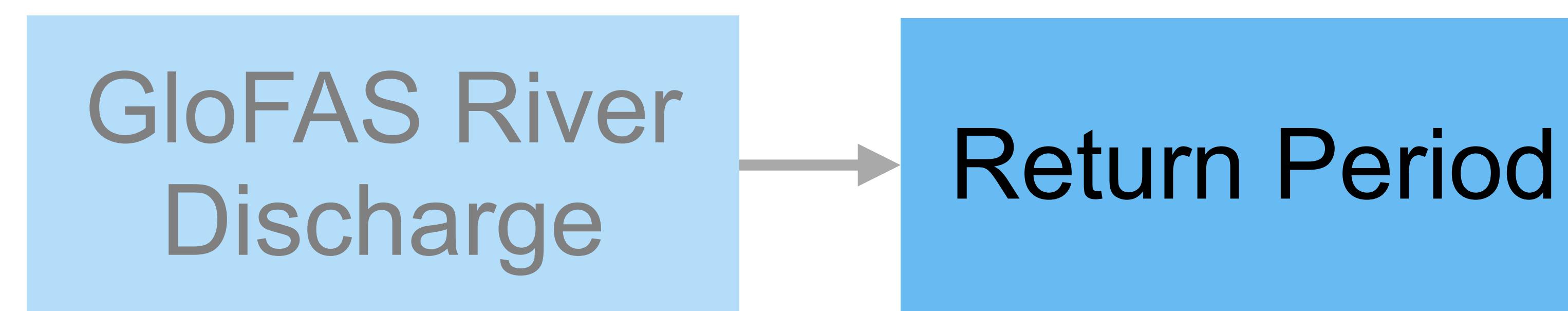
Pakistan 2022 Flood Showcase

GloFAS River
Discharge



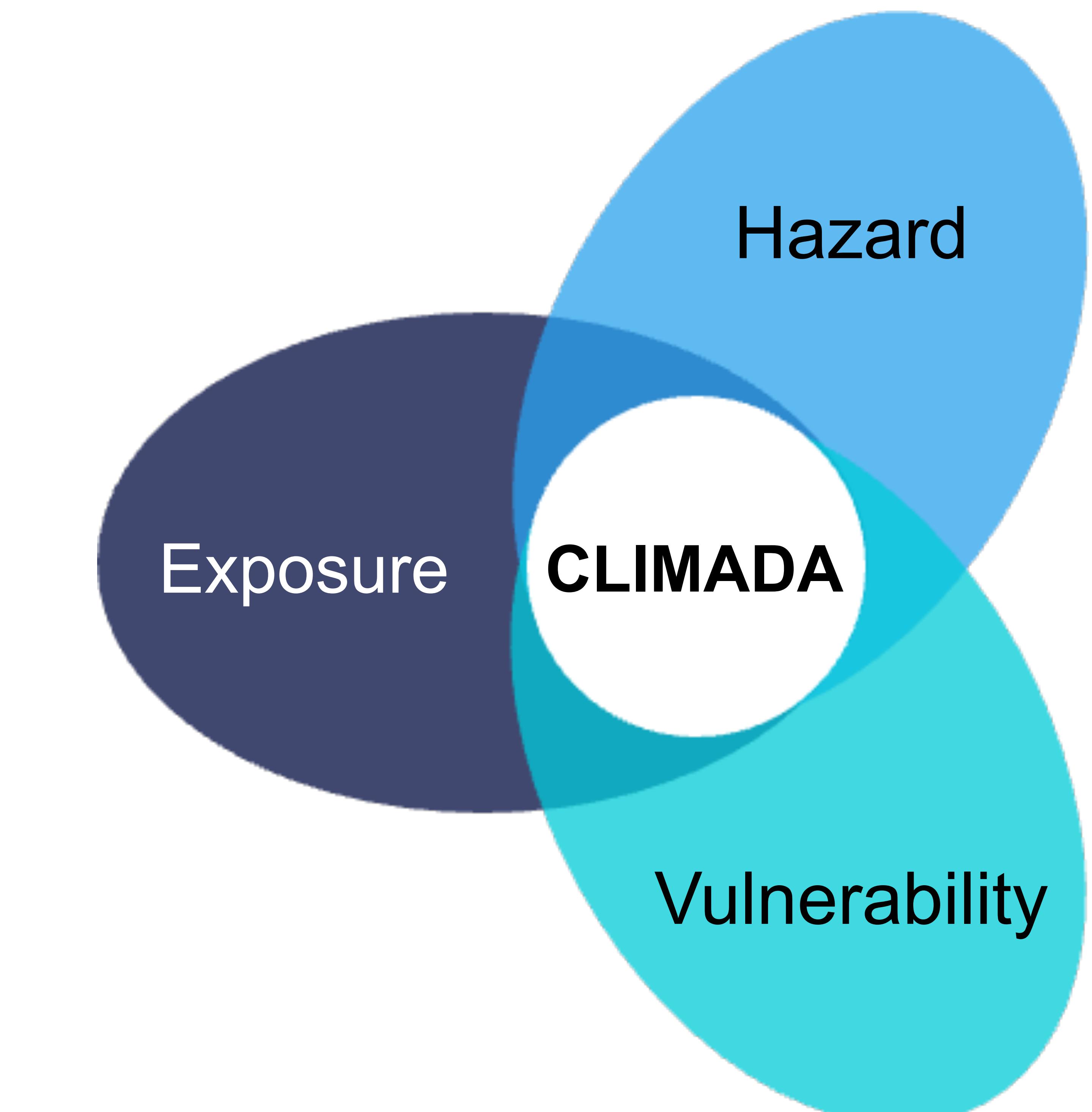
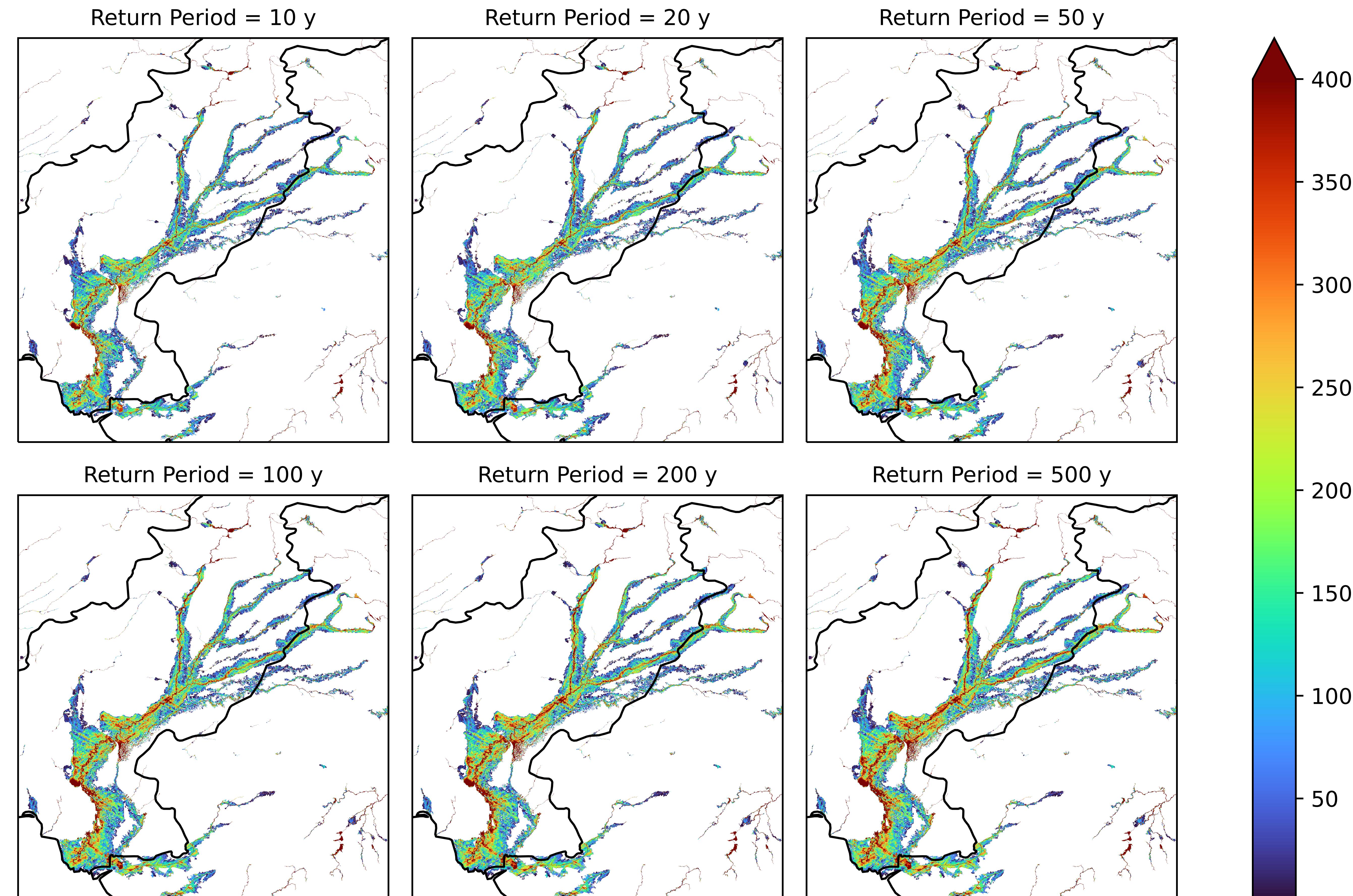
Flood Impact: Potential Displacement of Population

Pakistan 2022 Flood Showcase




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
Flood Impact: Potential Displacement of Population

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
Flood Impact: Potential Displacement of Population

Pakistan 2022 Flood Showcase



Flood Impact: Potential Displacement of Population

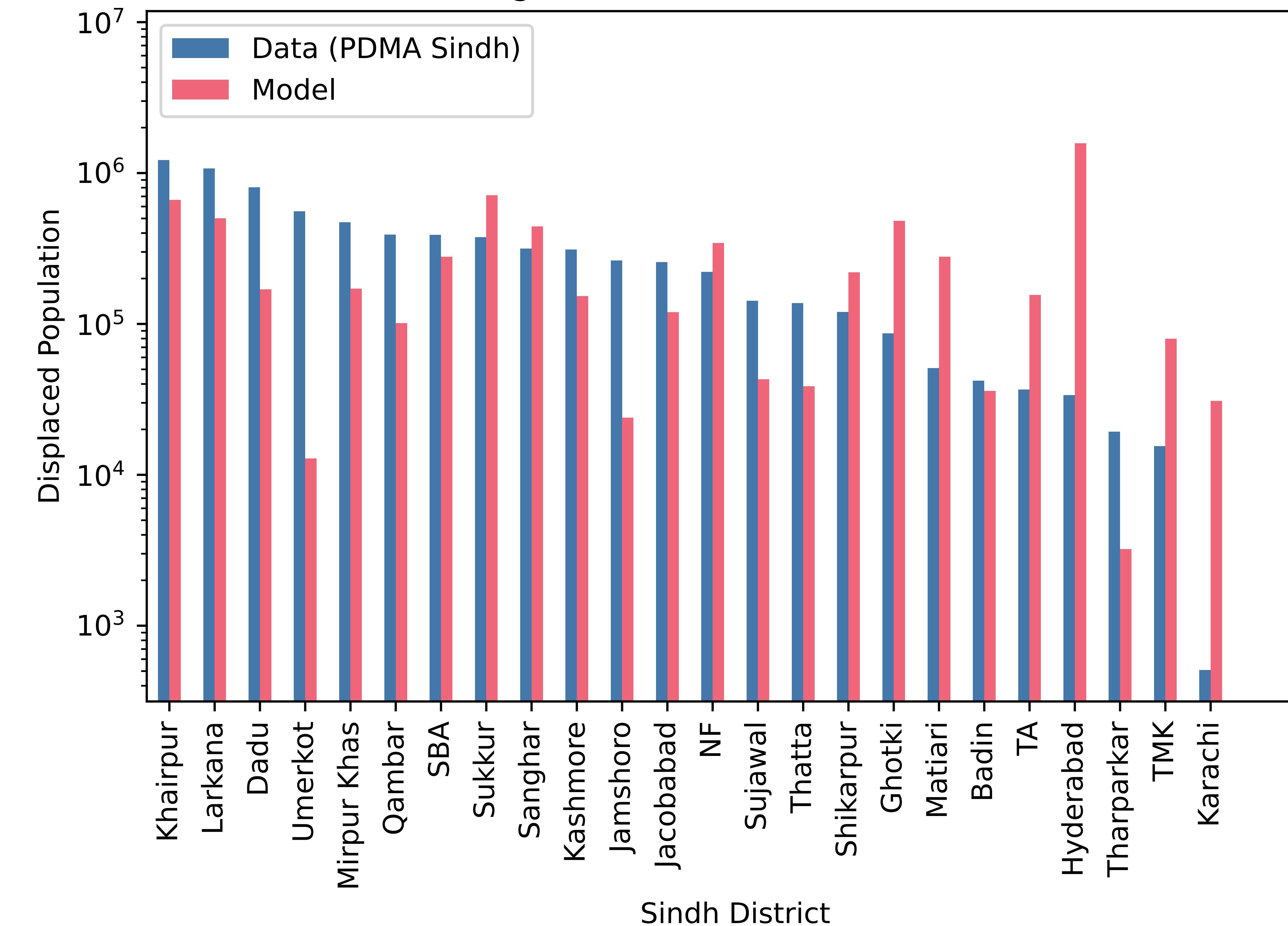
Pakistan 2022 Flood Showcase



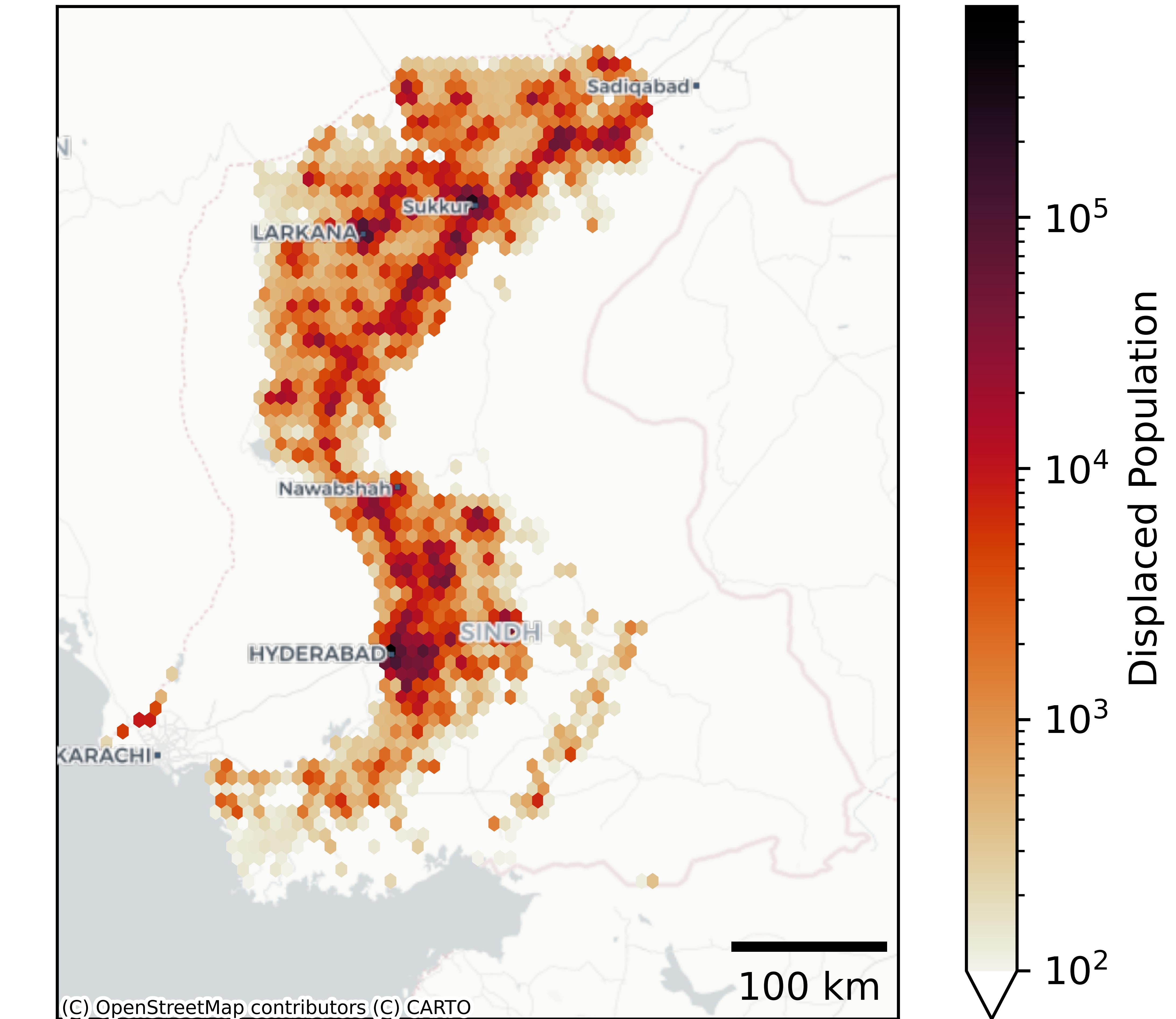
Calibrated Model Output

Data Source: Provincial Disaster Management Authority (PDMA), Sindh Province

Flooding from 2022-07-01 to 2022-09-30



Model Impact



Model Sensitivity Analysis

Based on hazard reanalysis data

Evaluation

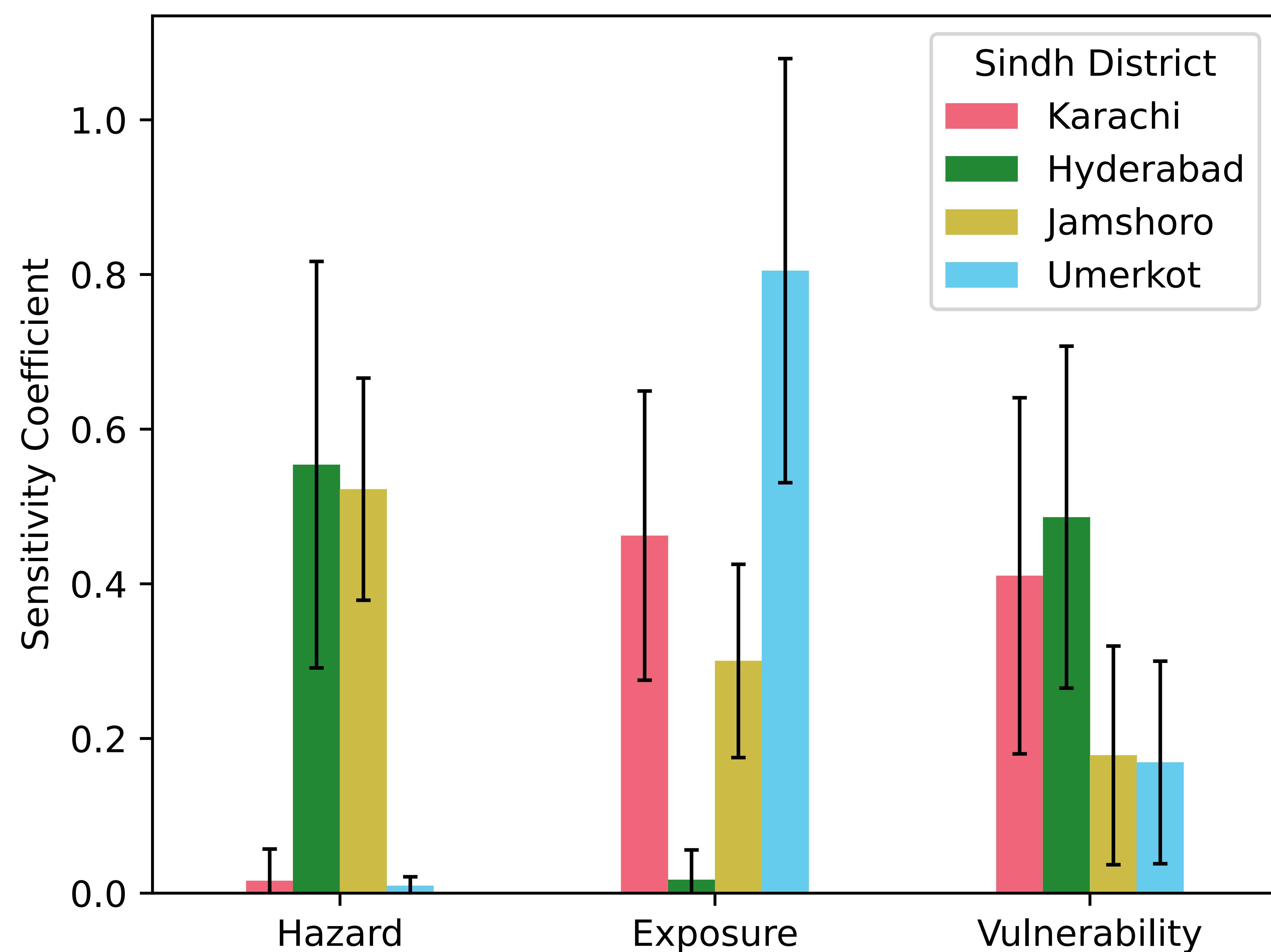
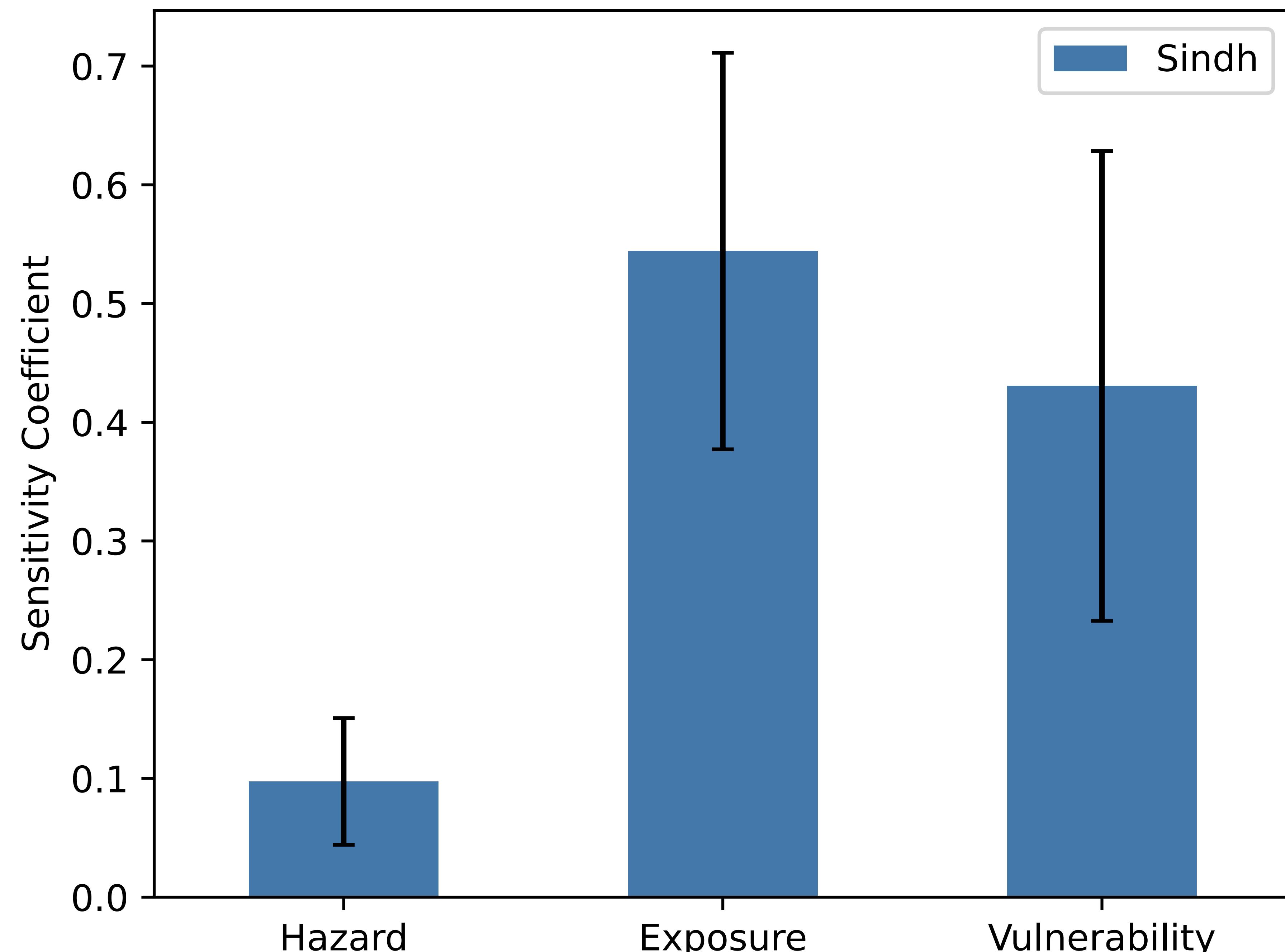
- Sensitivity Coefficient S_i relates variance w.r.t. parameter X_i of expected model output Y to overall model output variance

$$Y = f(X_1, \dots, X_k)$$

$$S_i = \frac{\text{Var}_{X_i}(\mathbb{E}[Y|X_i])}{\text{Var}(Y)} \in [0, 1]$$

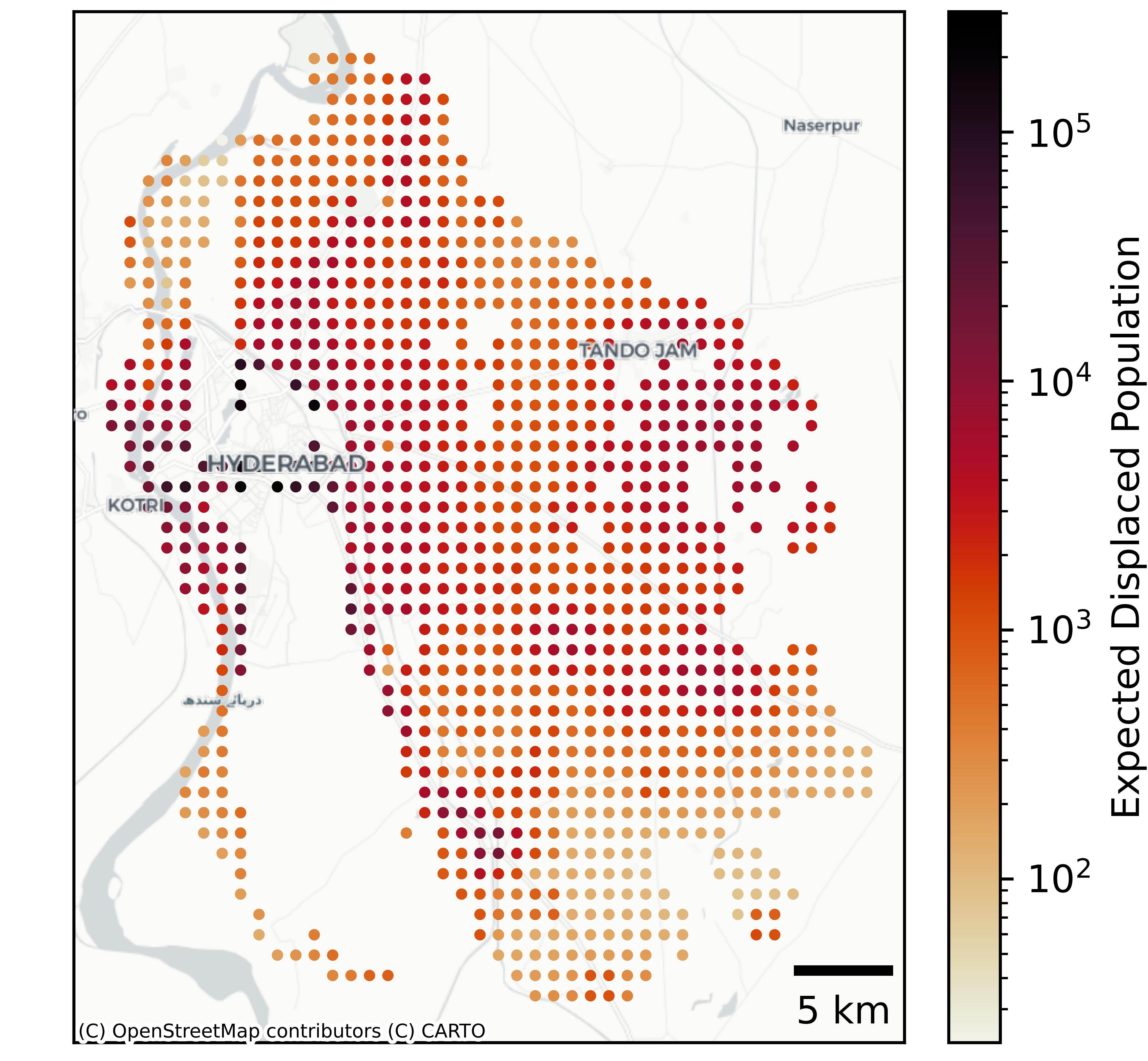
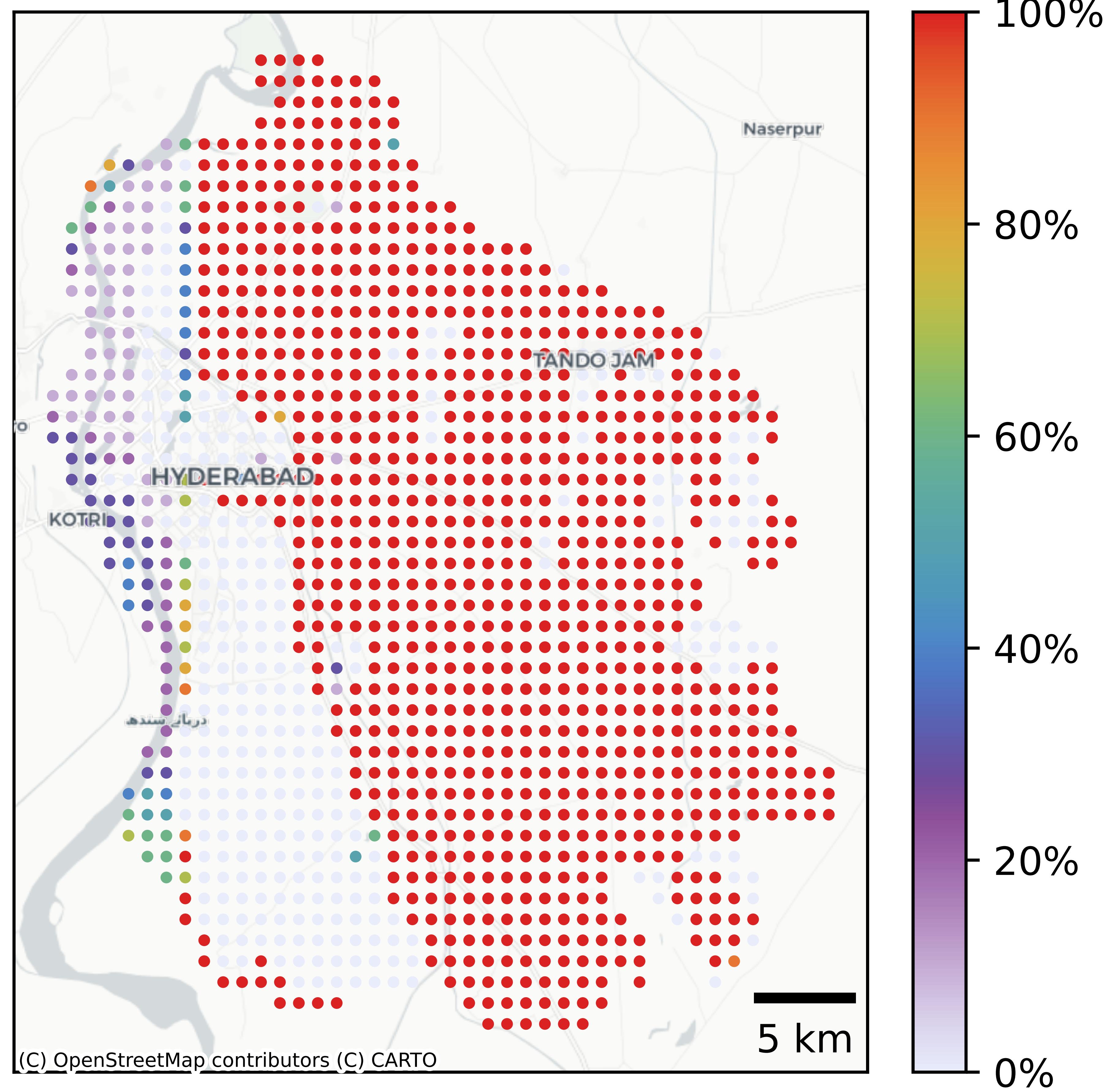
Results

- Sensitivity to hazard is low when only regarding statistical uncertainty
- Local sensitivity varies strongly



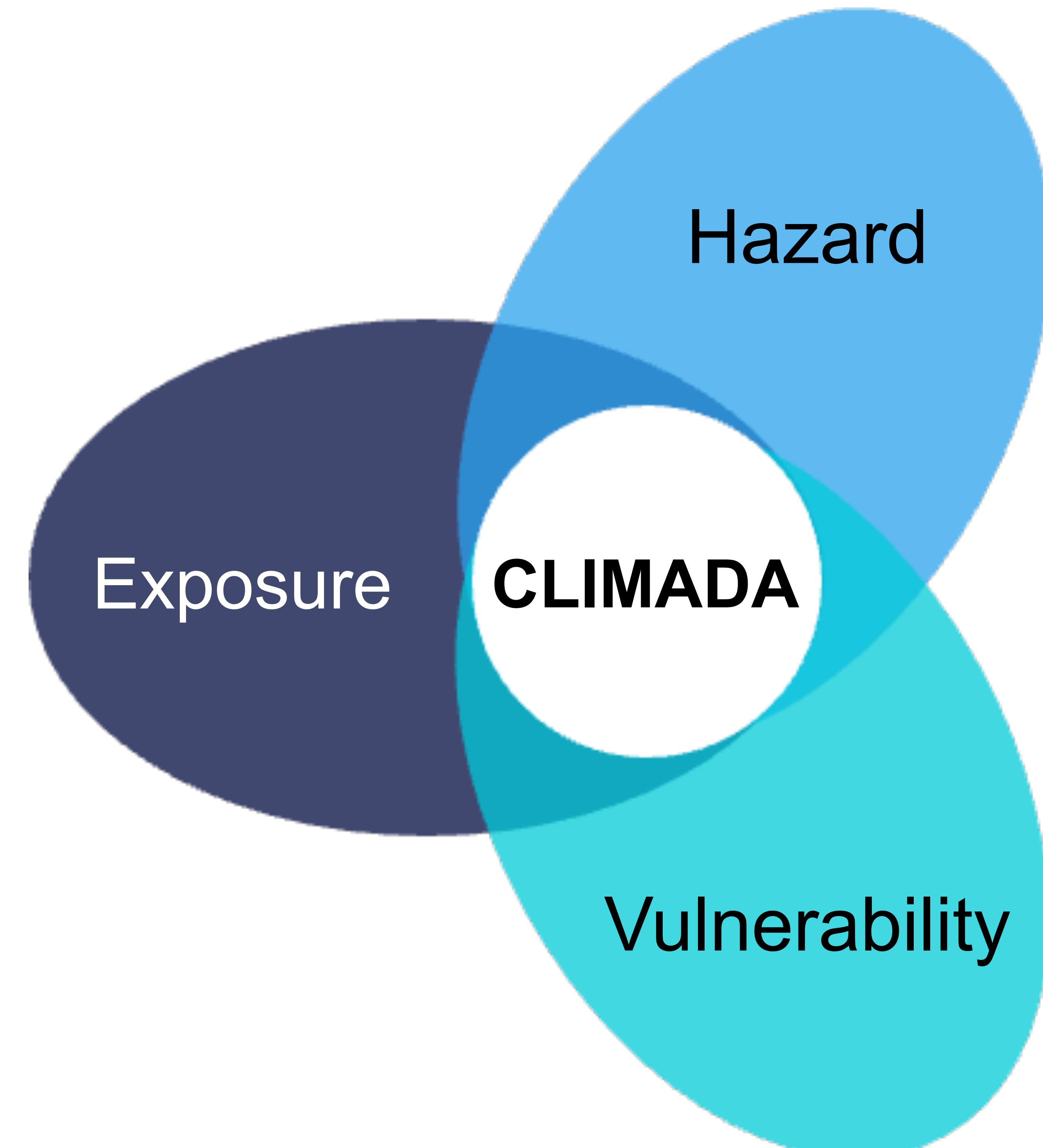
Model Sensitivity Analysis

Hyderabad District, Sindh

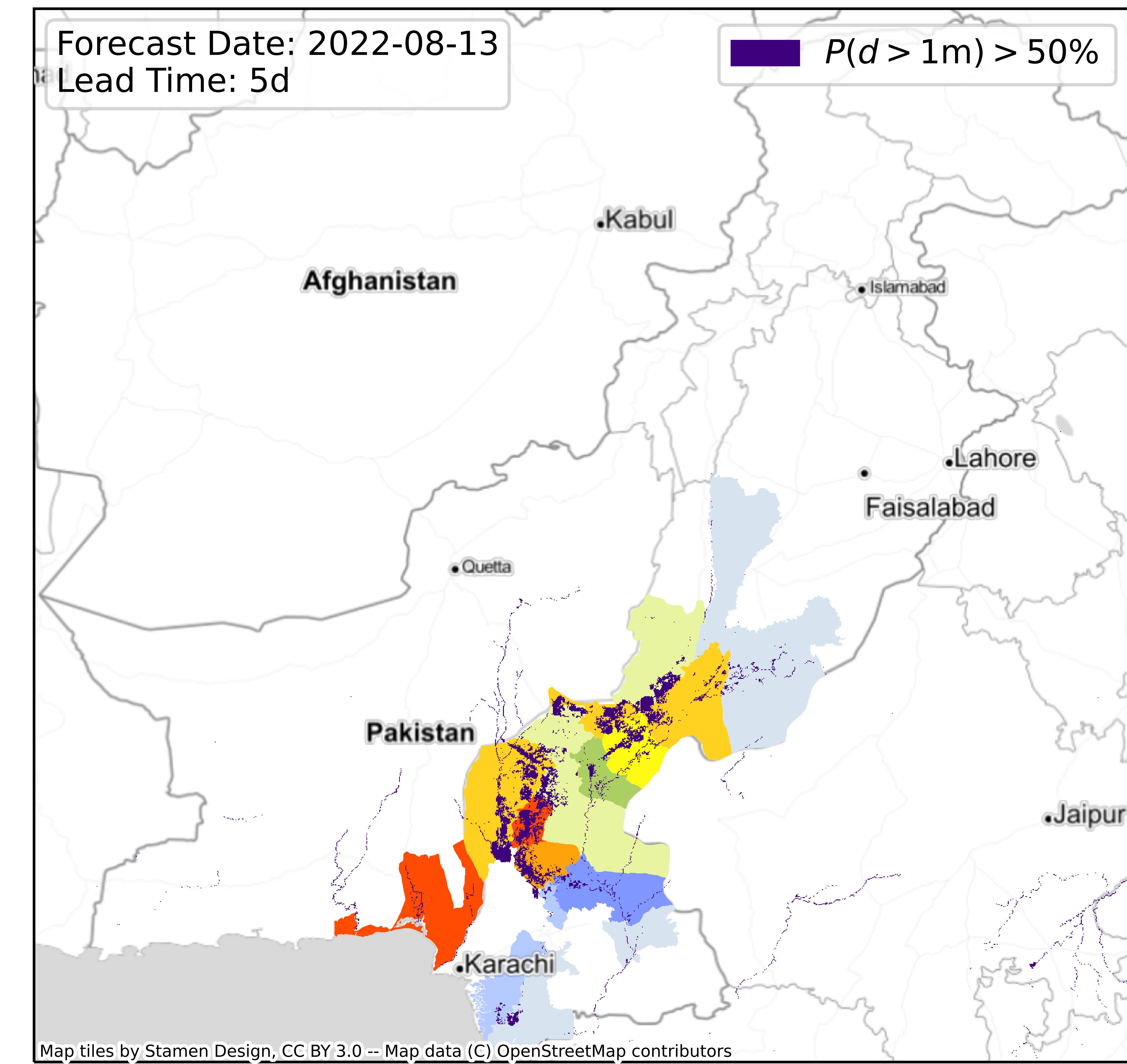


Outlook: Impact Forecasts

Considering Forecast Uncertainty



>10% of Population in Danger of Displacement



1.0

0.8

0.6

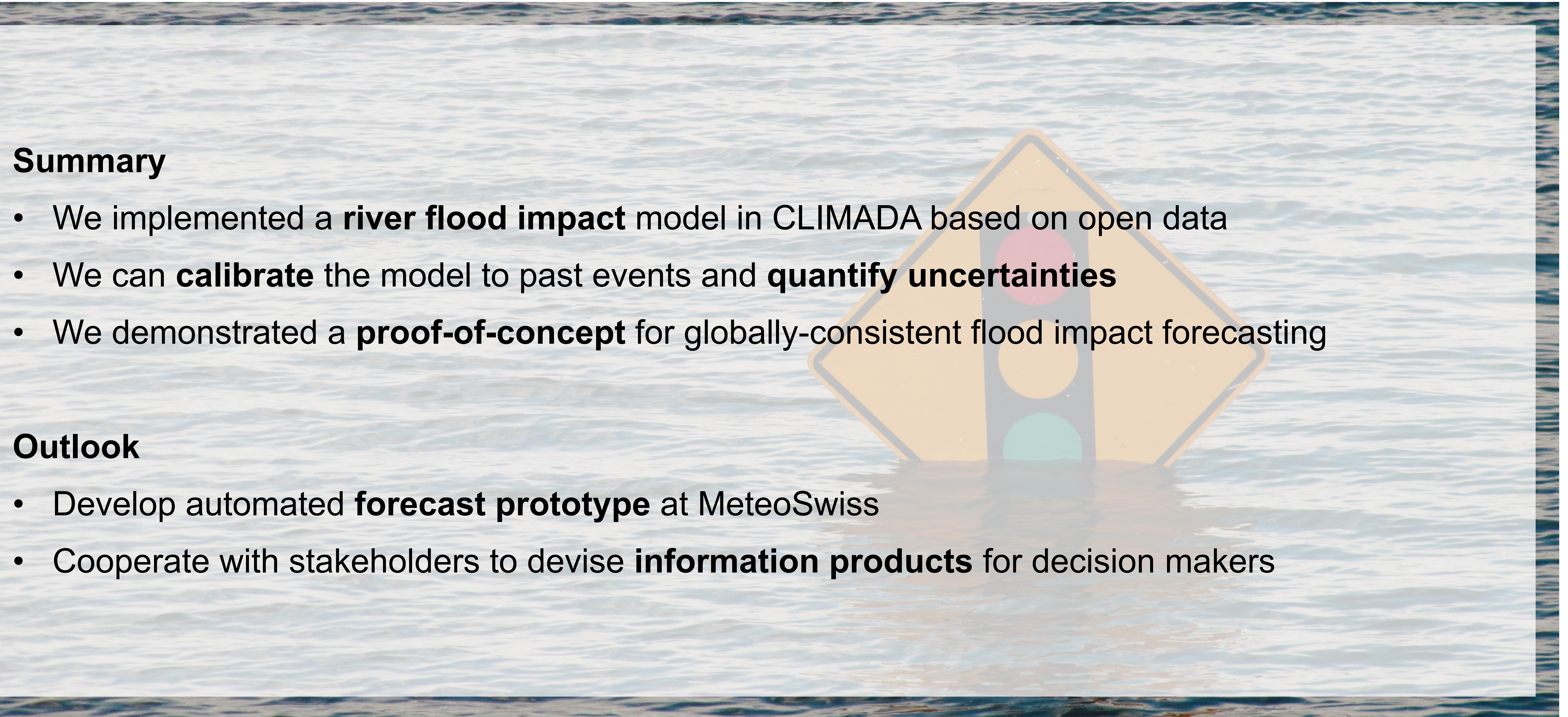
0.4

0.2

0.0

Probability

Conclusion



Summary

- We implemented a **river flood impact** model in CLIMADA based on open data
- We can **calibrate** the model to past events and **quantify uncertainties**
- We demonstrated a **proof-of-concept** for globally-consistent flood impact forecasting

Outlook

- Develop automated **forecast prototype** at MeteoSwiss
- Cooperate with stakeholders to devise **information products** for decision makers

Photo by [Kelly Sikkema](#) on [Unsplash](#)

References

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- Gridded Population of the World: <https://sedac.ciesin.columbia.edu/data/collection/gpw-v4>
- NASA Black Marble: <https://blackmarble.gsfc.nasa.gov/>