

Exposure and Vulnerability to Energy, Water, and Land Hotspots under Different Climate Futures

A Spatially-Explicit, Global Assessment of Vulnerable
Populations and Hydroclimatic Impacts

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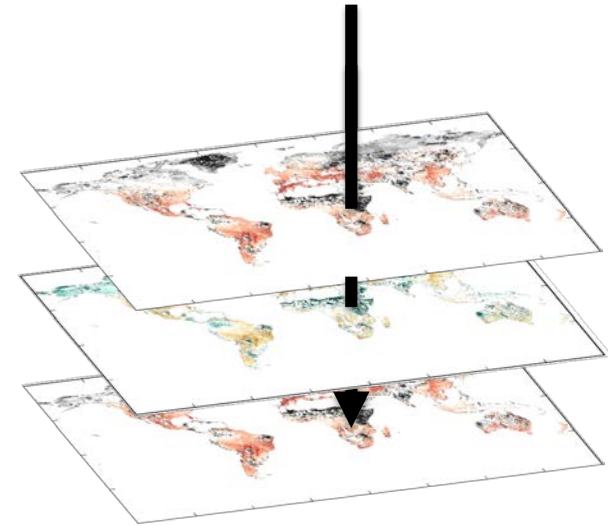
10th Integrated Assessment Modelling Consortium Meeting

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Multi-sectoral risk

- Many studies on climate impacts
- Increased attention on multiple sectors



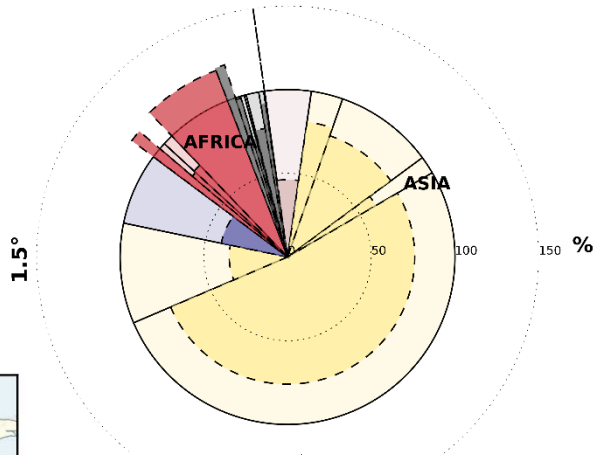
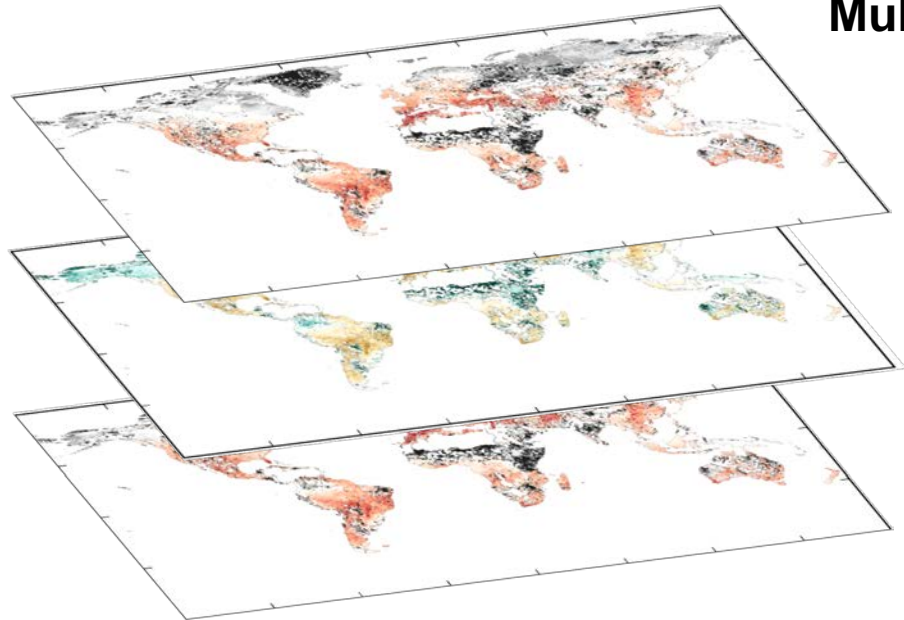
Risk aggregation in climate impacts research

1. Temporal: “*compound events*” ... two or more simultaneous events, e.g. drought + heatwave, with impacts often more severe than their sum
2. Spatial: “*hotspots*”... locations exposed to risks in multiple sectors, although not necessarily at the same time

Global mapping of multi-sector climate and vulnerability hotspots

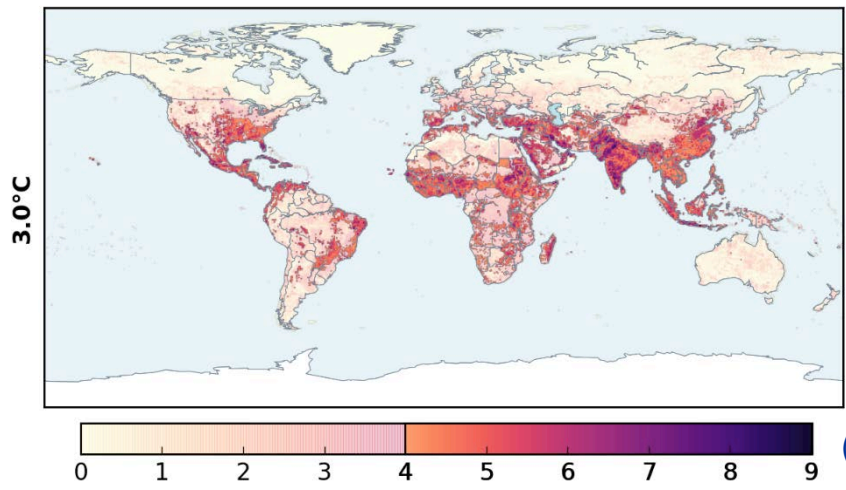
Multiple indicators (14) across 3 sectors

Hotspots of significant nexus vulnerabilities and impacts



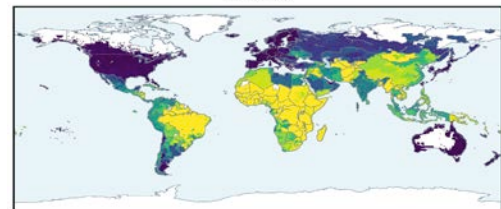
3.0 °C

combined indicators



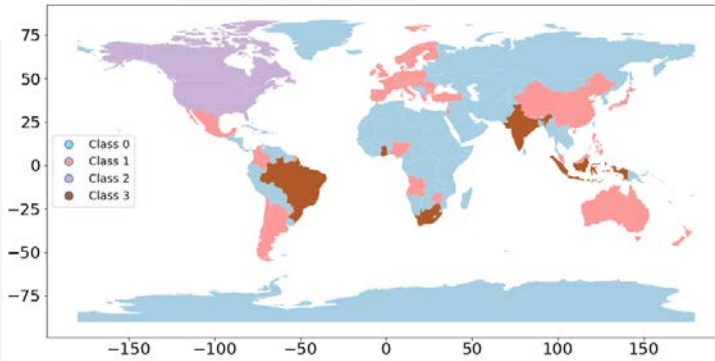
SSP1-3

2010

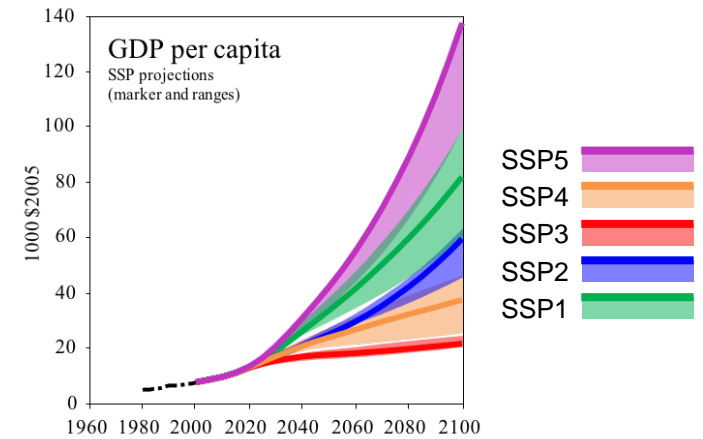


Byers et al. (2017, in review)

Projecting Subnational Income and Inequality

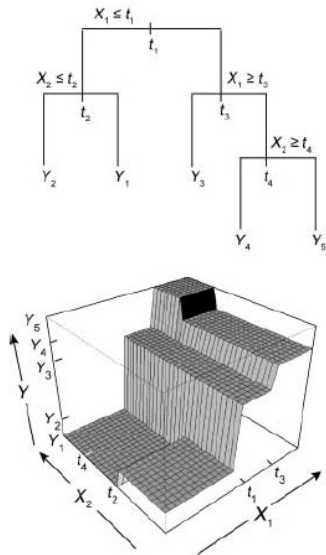


70% of Global Population



BRT Machine Learning

NLP Projection



$$\min_{i_j^r, t_j^r} \left(N^r \Omega \left(\frac{1}{4} I^r, I^r, T^r \right) - \sum_{j \in J} n_j^r \Omega \left(\frac{1}{4} I^r, i_j^r, t_j^r \right) \right)^2$$

$$\text{s.t. } \sum_{j \in J} n_j^r i_j^r = N^r I^r$$

$$\frac{i_j^r - i_j^{r-1}}{\Delta \tau} \leq 1.65 \frac{I^r - I^{r-1}}{\Delta \tau} \quad \forall j \in J \quad (6c)$$

$$\left| 1 - \frac{t_j^r}{t_j^{r-1}} \right| \leq 0.05 \Delta \tau \quad \forall j \in J \quad (6d)$$

$$\left| 1 - \frac{\sum_{j \in J} n_j^r i_j^r t_j^r - \sum_{j \in J} n_j^r i_j^{r-1} t_j^{r-1}}{N^r I^r T^r} \right| \leq 0.05 \quad \forall j \in J \quad (6e)$$

$$t_j^r \in [t_{\min}, t_{\max}] \quad \forall j \in J \quad (6f)$$

$$i_j^r \in [0, \infty) \quad \forall j \in J \quad (6g)$$

(6a)

(6b)

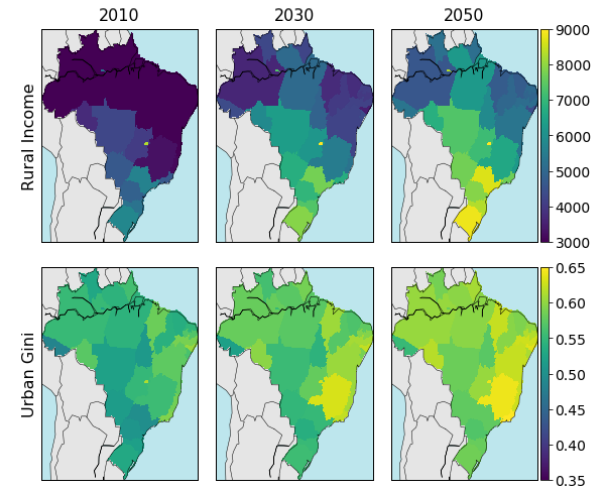
(6c)

(6d)

(6e)

(6f)

(6g)



Indicators

Urban and Rural Income and Inequality

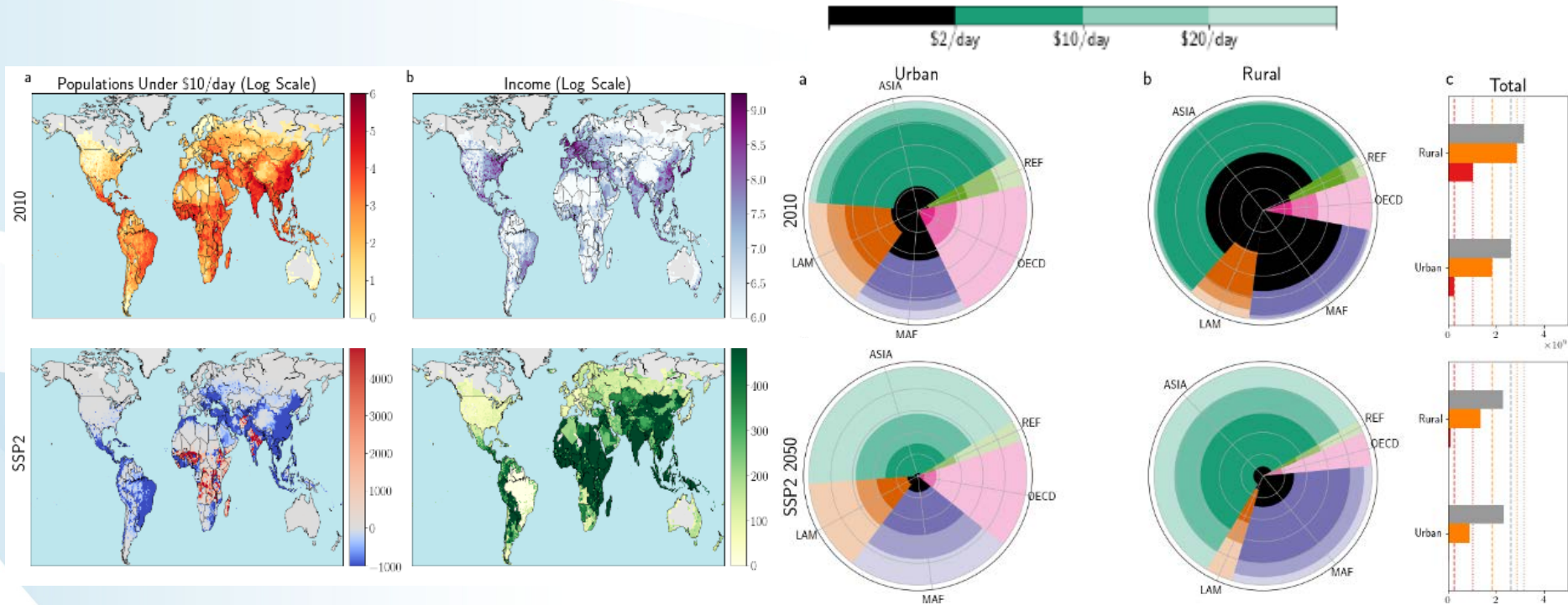
Scenarios

All 5 SSPs

Resolution

Global, States, 0.125 Grids

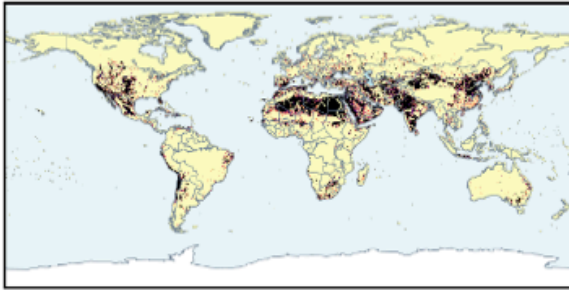
Populations Vulnerable to Poverty



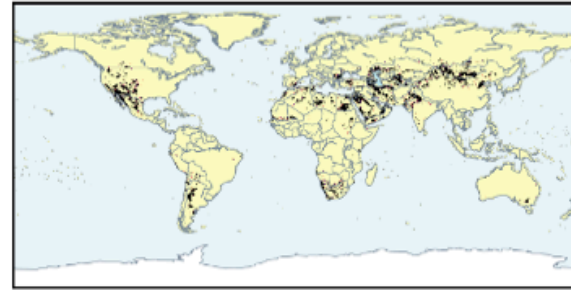
Multi-sector risk indicators

| Indicator | Description | Models |
|---|---|-------------------------|
| <u>Water</u> | | |
| Water stress index | Water stress index: as a proportion of human demands divided by renewable surface water resources | 5 GCMs, 3 GHMs |
| Non-renewable GW abstraction index | Fraction of groundwater abstraction that is non-renewable | HadGEM2-ES + PCR-GLOBWB |
| Drought intensity | % change in drought intensity (deficit / duration) | 5 GCMs, 4 GHMs |
| Peak flows risk | Substantial change in flood risk (doubling) is expected | 5 GCMs, 4 GHMs |
| Seasonality | % change for the index of mean seasonality | 5 GCMs, 4 GHMs |
| Inter-annual variability | % change for the index of mean inter-annual variability | 5 GCMs, 4 GHMs |

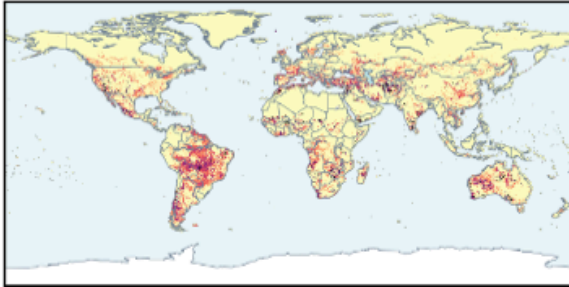
w1: Water stress index



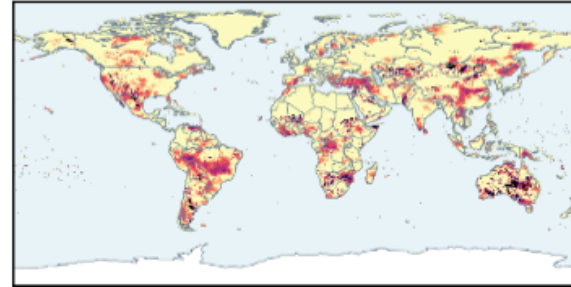
w2: Non-renew GW abstr.



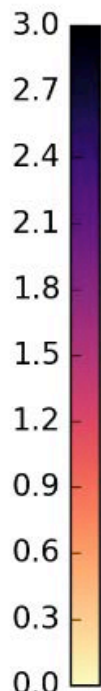
w3: Drought intensity



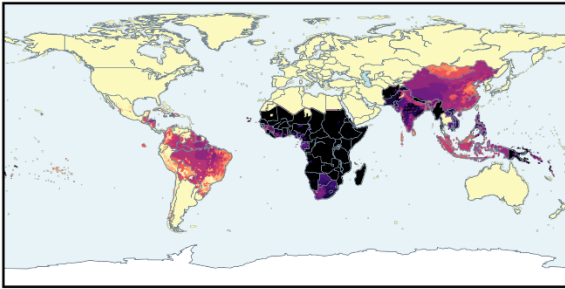
w6: Inter-annual variability



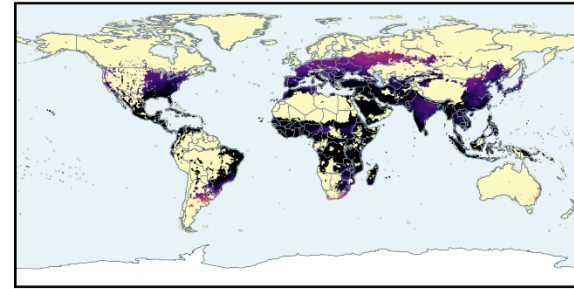
Water impacts: 2.0° SSP2



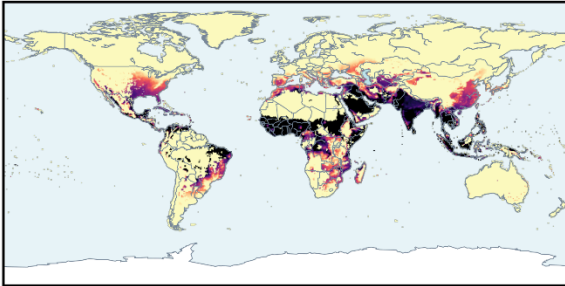
e1: Lack of clean cooking access



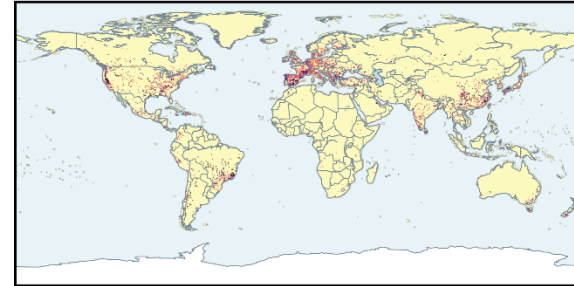
e2: Heat events



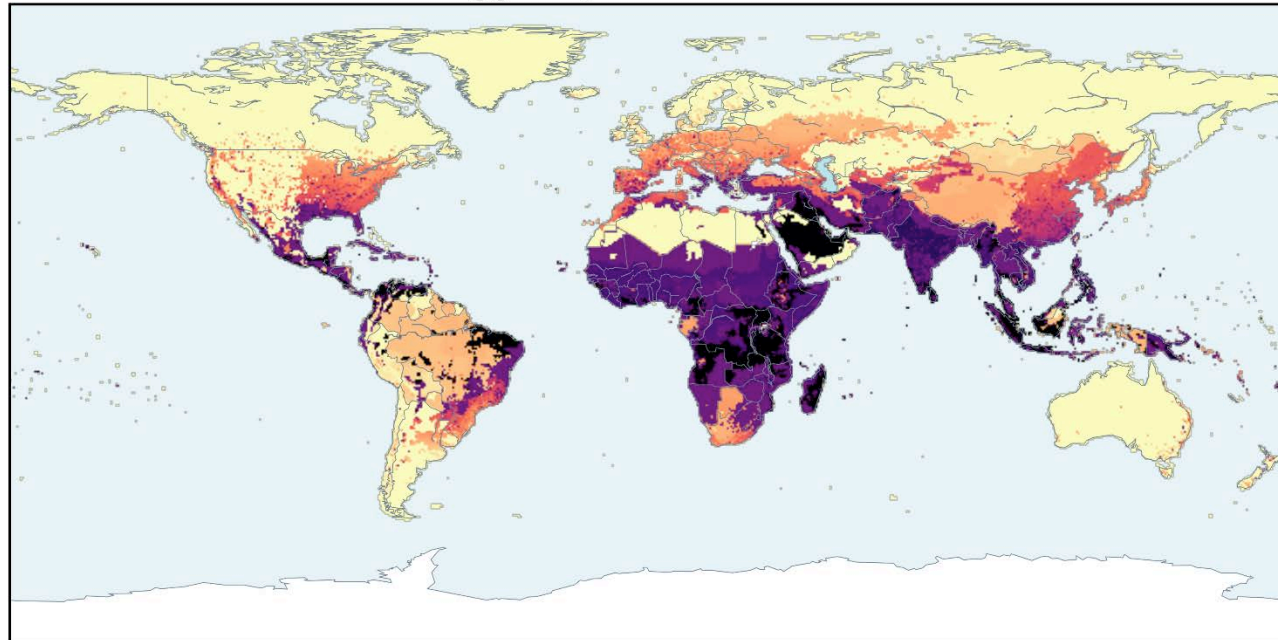
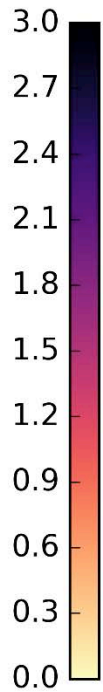
e3: Cooling degree days



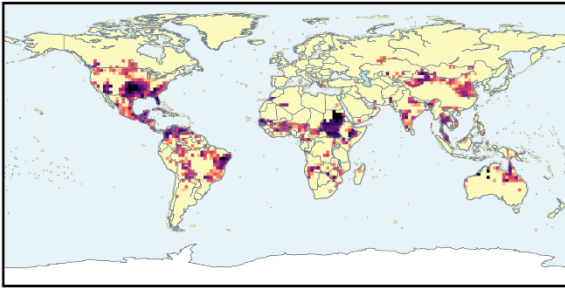
e4: Hydroclimate risk to power plants



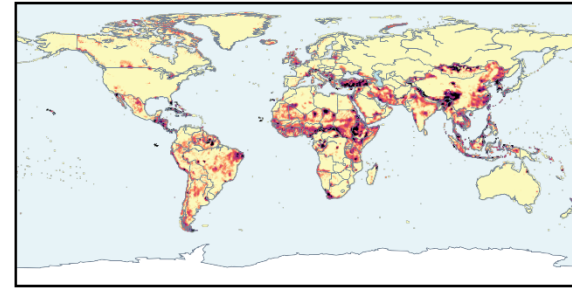
Energy impacts: 2.0° SSP2



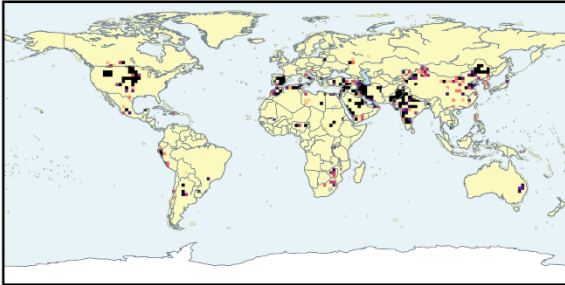
I1: Crop yield change



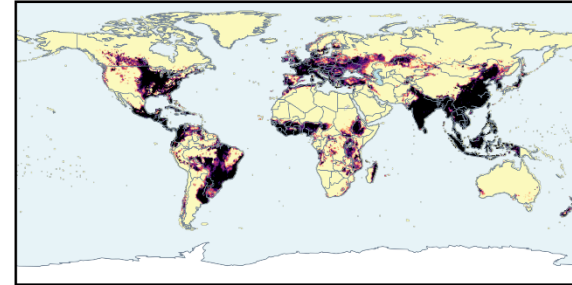
I3: Habitat degradation



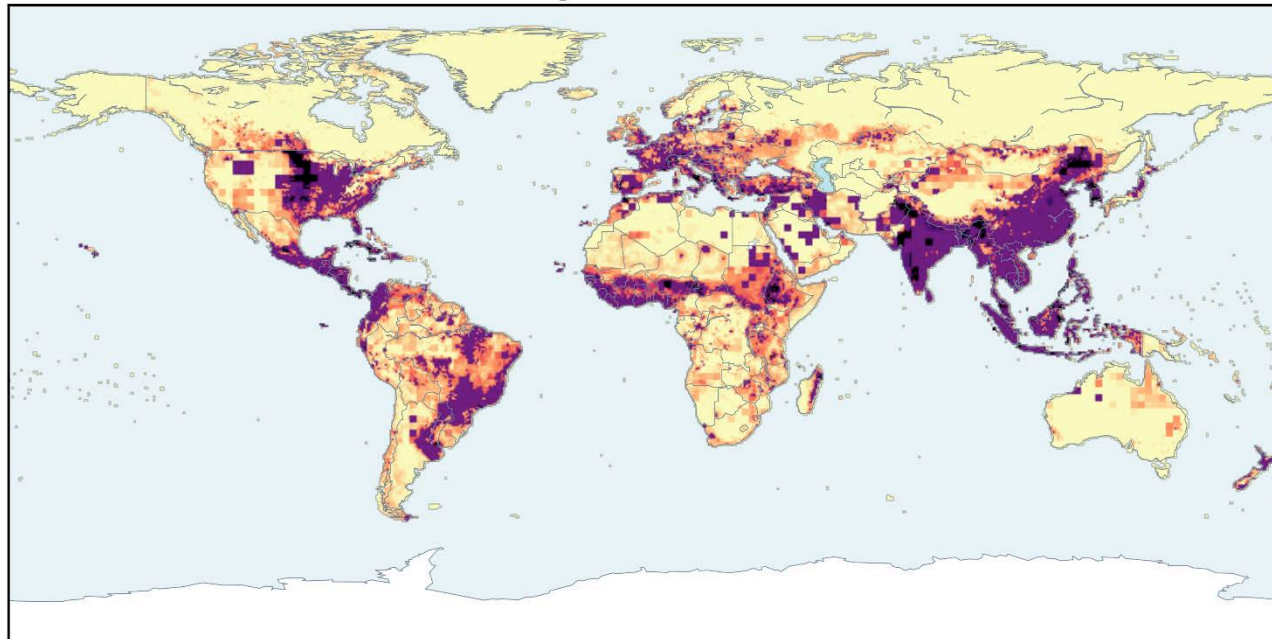
I2: Ag. water stress index



I4: Nitrate leaching



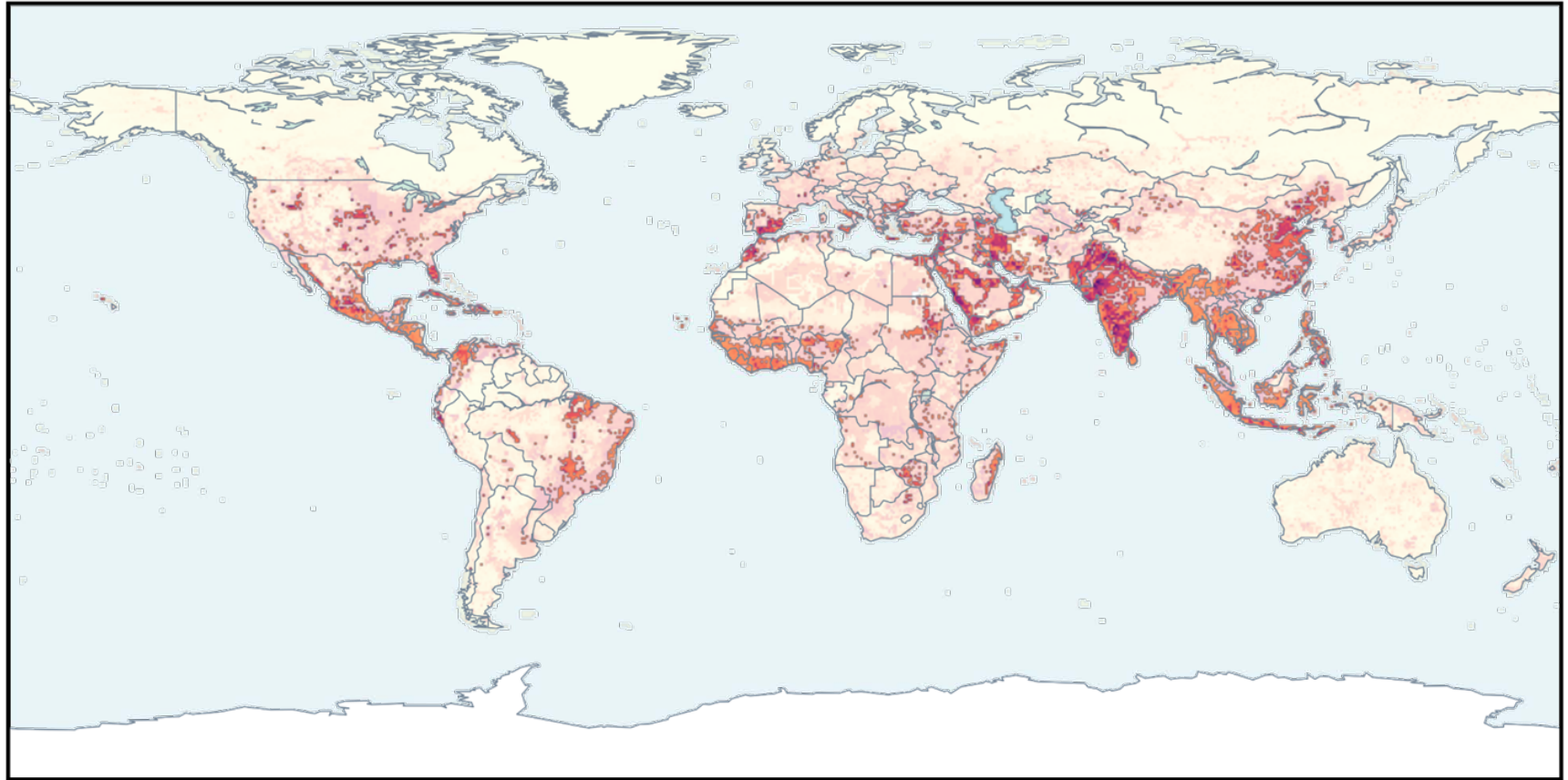
Land impacts: 2.0° SSP2



Global hotspot exposure

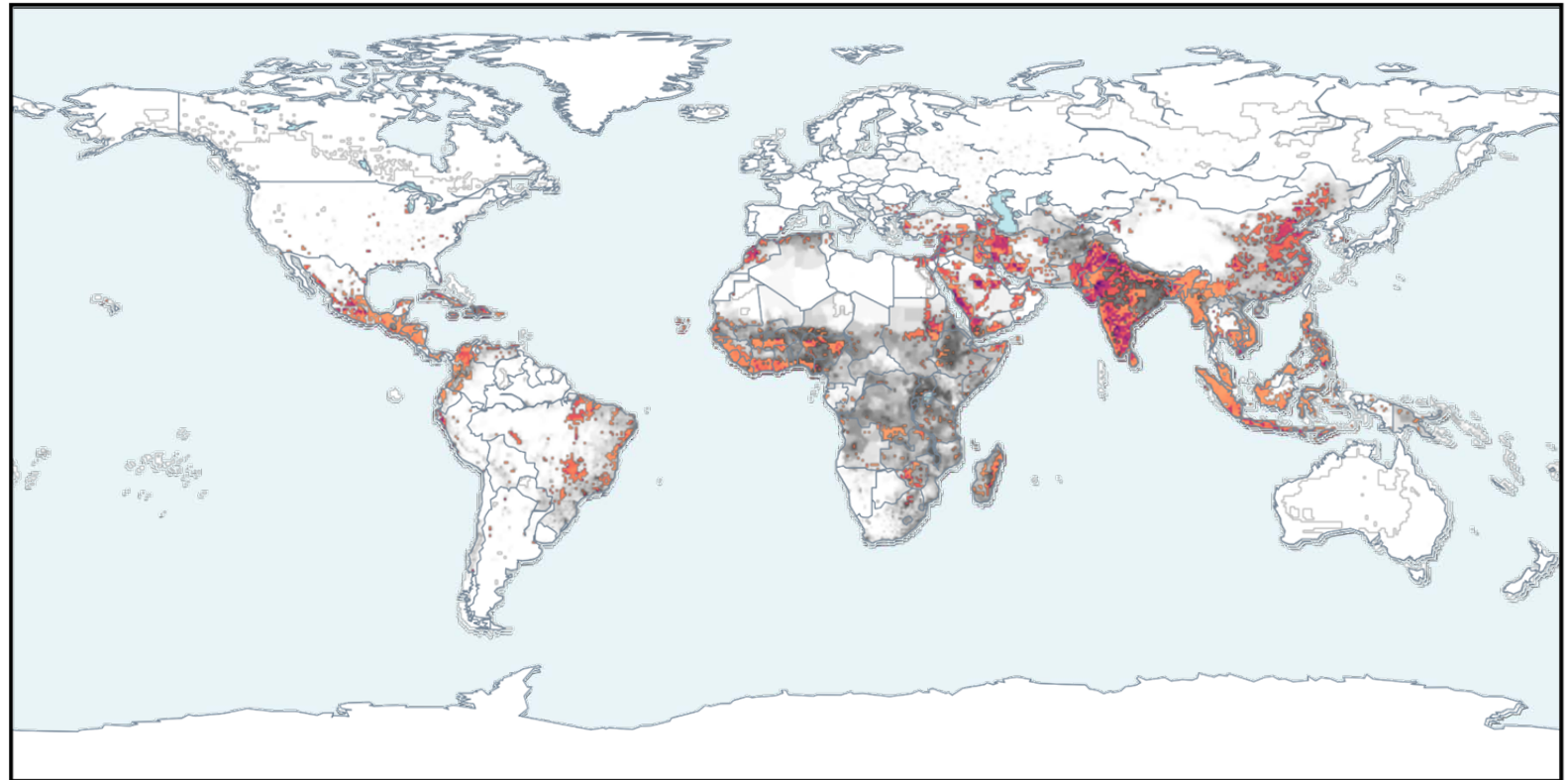
3.0 °C

1.5 °C



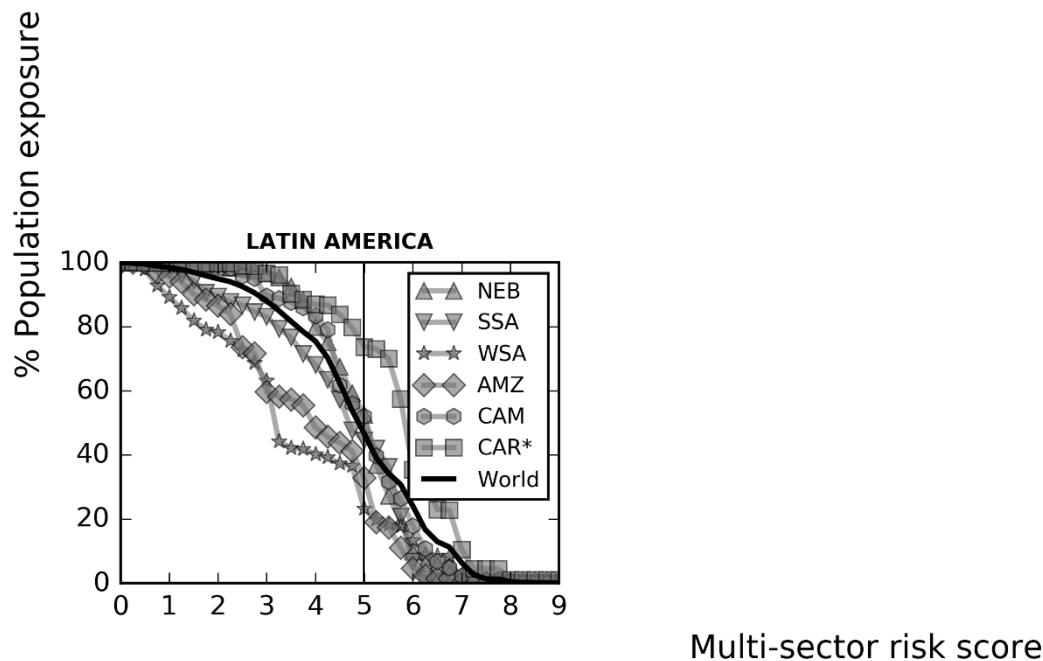
Hot and vulnerable

3.0 °C



Byers et al.
(2017, in review)

MSR >4.0, income < \$10 /day



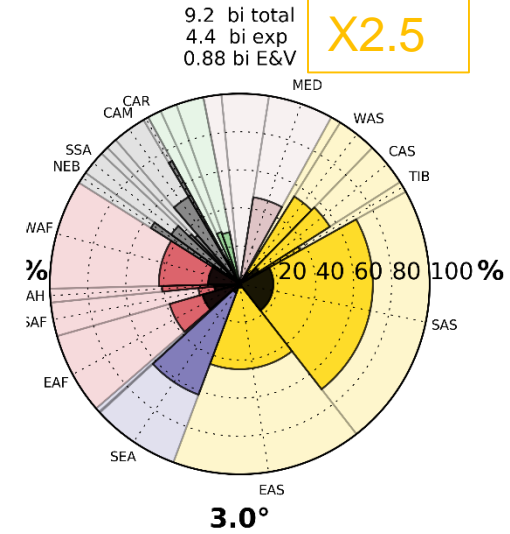
- Northern hemisphere regions have better than average impacts
- Most Asian and southern regions are on/worse than average

Impacted Populations

Exposed

x2

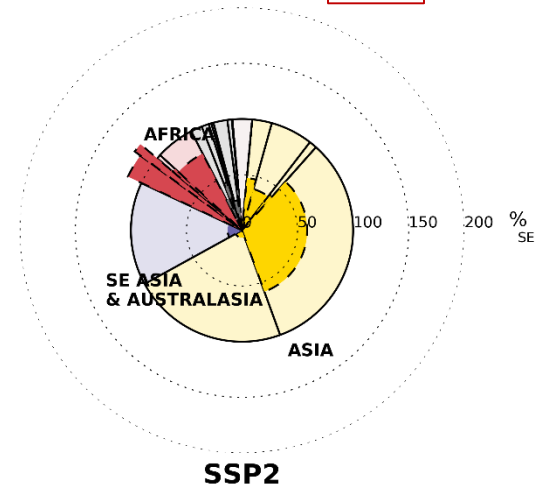
X2.5



Exposed & Vulnerable

x5

x15



Byers et al.
(2017, in review)

Sustainability

Middle of the road

Rocky road

Conclusions and findings

- More than double exposure between 1.5-3.0°C
- Asia already faces severe exposure and vulnerability at 1.5°C
- African exposure emerges intensely at 2-3°C
- As many E&V in SSP3 as today, but more concentrated in Asia and Africa
- Difference in SSPs results in order of magnitude reductions in E&V

Thank you very much for your attention!

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Backup Slides

Climate change index scoring under uncertainty

2.0°C climate example: Drought intensity change

Continuous scale (0 to 3) with intermediate ranges determined

- 0. Negligible risk
- 1. Low risk
- 2. Moderate risk
- 3. High risk

