



# Closing the loop between integrated assessment and climate risk research - Rapid climate risk emulation

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EGU General Assembly  
CL3.2.5 Techniques and advances in climate emulation, economics  
and integrated assessment

25<sup>th</sup> April 2023 – 14:00-17:55

# Emulating climate impacts and exposure from IAM scenarios

Aim: Exogenously assess the climate impacts and exposure from IAM emissions scenarios

**IAM scenarios  
+ SCM (FaIR, MAGICC etc)  
= rapid climate risk  
emulation**

Background:

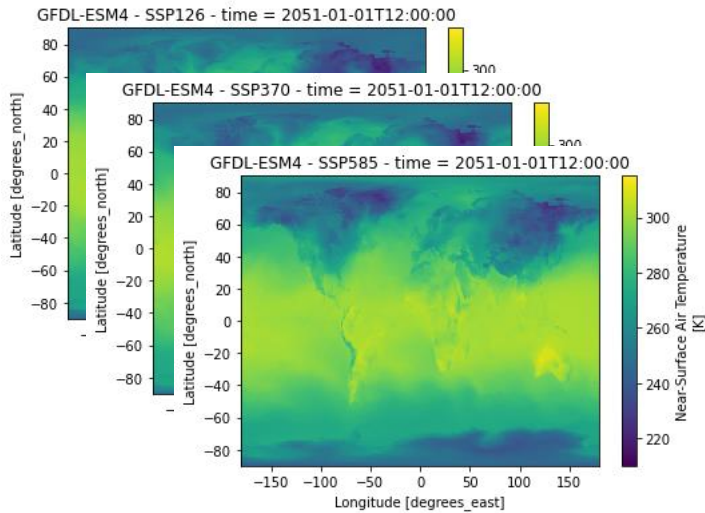
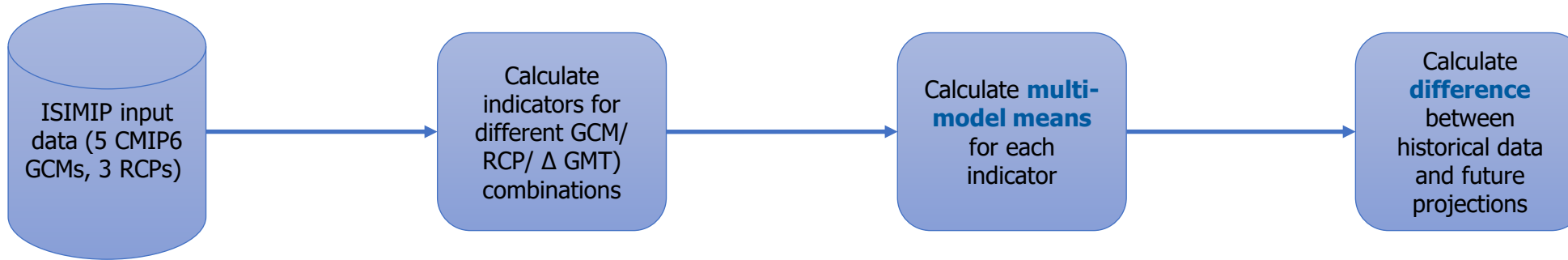
Expanding number of emulations (e.g. SCMs) that estimate global warming from an input emissions scenario (e.g. FaIR, MAGICC, OSCAR, etc) – annual timeseries, global and macro-region variables.

- More recently, new approaches to extend this with grid-level assessment, and more climate-related variables (e.g. temperatures, precipitation (MESMER))

Here:

- Extend approaches for climate impacts and risk assessment – e.g. heatwaves, drought => population exposed
  - Pre-process: Climate impacts & exposure data (e.g. from ISIMIP impact models)
  - Input: Global mean temperature projection (+IAM scenario), e.g. from AR6 Scenarios database
  - Output: Maps & table data of land/population exposure to impacts

# Workflow: Pre-processing impacts data



## Precipitation

- Heavy precipitation days
- Wet & very wet days
- Precipitation intensity index
- Consecutive dry days

## Hydrology

- Drought intensity
- Seasonality
- Interannual variability
- Water stress

## Temperature

- Heat wave intensity
- Tropical nights

## Land

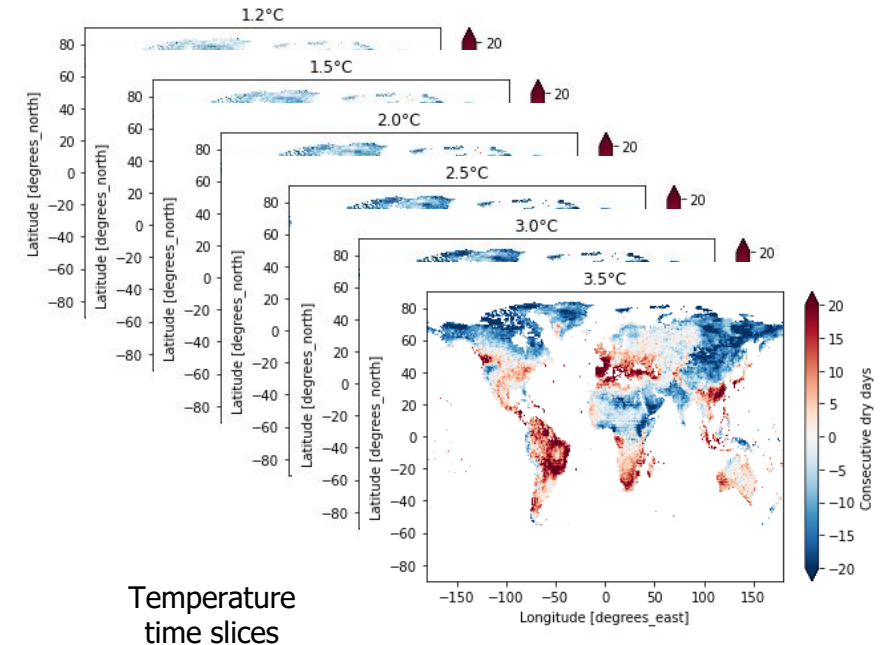
- Low intensity land cover
- Crop yields

## Energy demands

- Cooling degree days

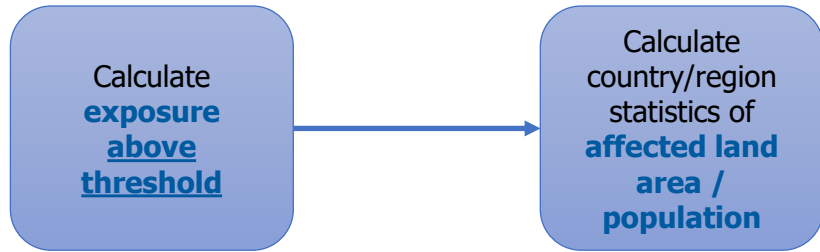
## Air pollution

- Ambient PM2.5



Temperature time slices

# Workflow: Calculate population & land exposure



## Available statistics:

Hazard/Difference value for country/region

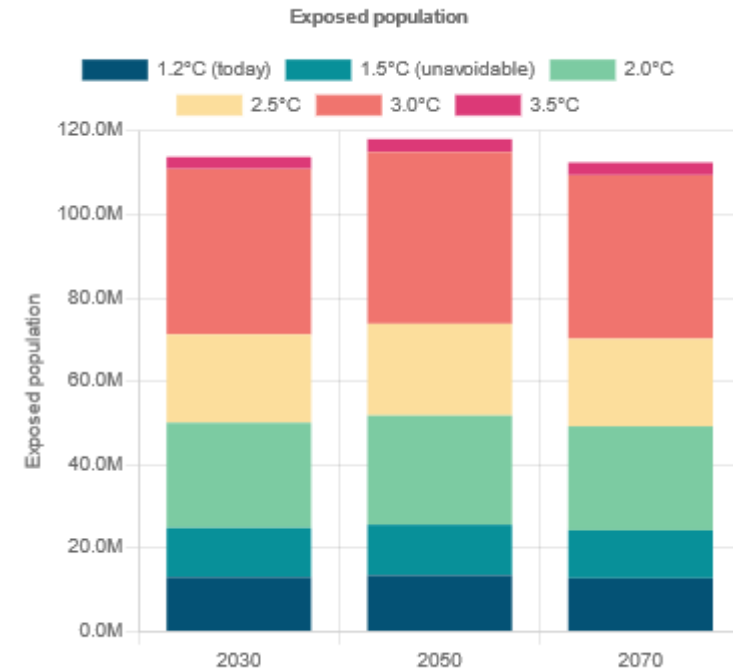
- weighted by population
- weighted by land area

## **Above threshold:** e.g.

- Water stress index >0.3
- x% increase in Tropical Nights
- x% reduction in crop yields

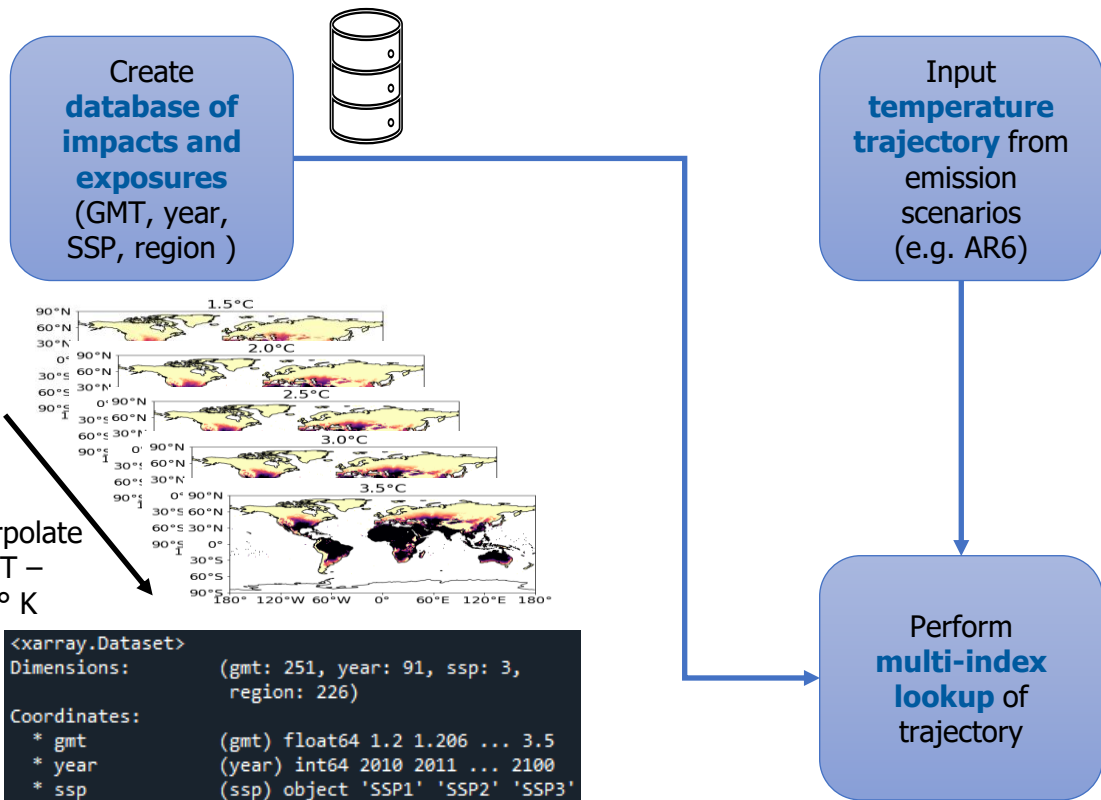
## Exposure statistics:

- Exposed land area
- Exposed land area %
- Exposed population
- Exposed population %
- + High/Low population ranges

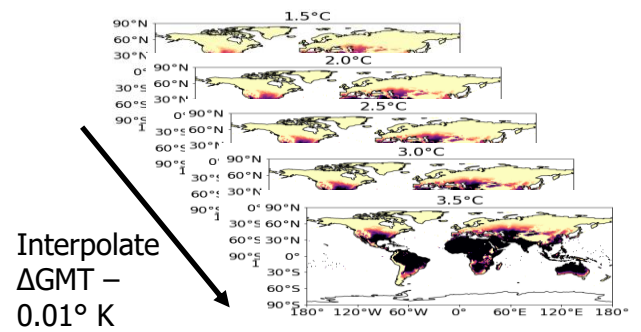
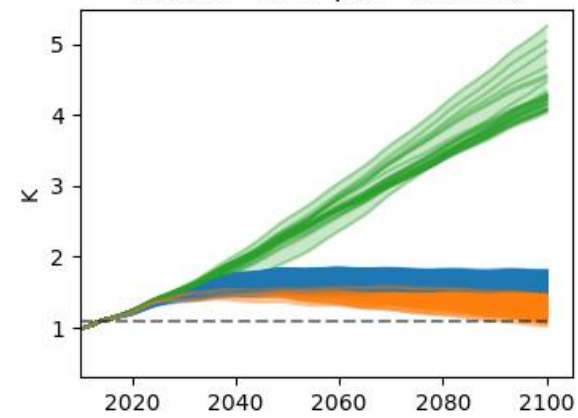


Population exposed to consecutive dry days in Brazil

# Workflow: Map impacts



AR6DB - GSAT p50 - MAGICC



Interpolate  $\Delta$ GMT – 0.01° K

```
<xarray.Dataset>
Dimensions:   (gmt: 251, year: 91, ssp: 3, region: 226)
Coordinates:
  * gmt       (gmt) float64 1.2 1.206 ... 3.5
  * year     (year) int64 2010 2011 ... 2100
  * ssp      (ssp) object 'SSP1' 'SSP2' 'SSP3'
  * region   (region) object 'AFG' ... 'world'
Data variables: (12/189)
  iavar|Exposure|Land area
  iavar|Exposure|Land area|%
  iavar|Exposure|Land area|High
  iavar|Exposure|Land area|High|%
  iavar|Exposure|Land area|Low
  iavar|Exposure|Land area|Low|%
  ...
  sdii|Hazard|Difference
  sdii|Hazard|Difference|Land area weighted
  sdii|Hazard|Difference|Population weighted
  sdii|Hazard|Risk score
  sdii|Hazard|Risk score|Land area weighted
  sdii|Hazard|Risk score|Population weighted
```

↓ 3.5 °C

Million people exposed to X

→ 2100

gmt	2015	2016	2017	2018	2019	2020	2021	2022	2023	2024	2025	2026	2027	2028	2029	2030
1.20	82.0	82.7	83.4	84.0	84.7	85.4	85.9	86.5	87.0	87.5	88.0	88.6	89.1	89.6	90.1	90.6
1.21	83.3	84.0	84.7	85.4	86.1	86.8	87.3	87.9	88.4	88.9	89.5	90.0	90.5	91.0	91.6	92.1
1.22	84.7	85.4	86.1	86.8	87.5	88.2	88.7	89.3	89.8	90.3	90.9	91.4	91.9	92.5	93.0	93.6
1.23	86.0	86.7	87.4	88.1	88.9	89.6	90.1	90.7	91.2	91.7	92.3	92.8	93.4	93.9	94.5	95.0
1.24	87.3	88.1	88.8	89.5	90.2	91.0	91.5	92.1	92.6	93.2	93.7	94.3	94.8	95.4	95.9	96.5
1.25	88.7	89.4	90.1	90.9	91.6	92.3	92.9	93.5	94.0	94.6	95.1	95.7	96.2	96.8	97.4	97.9
1.26	90.0	90.7	91.5	92.2	93.0	93.7	94.3	94.9	95.4	96.0	96.6	97.1	97.7	98.2	98.8	99.4
1.27	91.3	92.1	92.8	93.6	94.4	95.1	95.7	96.3	96.8	97.4	98.0	98.5	99.1	99.7	100.3	100.8
1.28	92.7	93.4	94.2	95.0	95.7	96.5	97.1	97.7	98.2	98.8	99.4	100.0	100.6	101.1	101.7	102.3
1.29	94.0	94.8	95.6	96.3	97.1	97.9	98.5	99.1	99.6	100.2	100.8	101.4	102.0	102.6	103.2	103.7
1.30	95.3	96.1	96.9	97.7	98.5	99.3	99.9	100.5	101.1	101.6	102.2	102.8	103.4	104.0	104.6	105.2
1.31	96.7	97.5	98.3	99.1	99.9	100.7	101.3	101.9	102.5	103.1	103.7	104.3	104.9	105.5	106.1	106.7
1.32	98.0	98.8	99.6	100.4	101.2	102.0	102.6	103.3	103.9	104.5	105.1	105.7	106.3	106.9	107.5	108.1
1.33	99.3	100.2	101.0	101.8	102.6	103.4	104.0	104.7	105.3	105.9	106.5	107.1	107.7	108.3	109.0	109.6
1.34	100.7	101.5	102.3	103.2	104.0	104.8	105.4	106.1	106.7	107.3	107.9	108.5	109.2	109.8	110.4	111.0
1.35	102.0	102.9	103.7	104.5	105.4	106.2	106.8	107.5	108.1	108.7	109.3	110.0	110.6	111.2	111.9	112.5
1.36	103.4	104.2	105.0	105.9	106.7	107.6	108.2	108.9	109.5	110.1	110.8	111.4	112.0	112.7	113.3	113.9

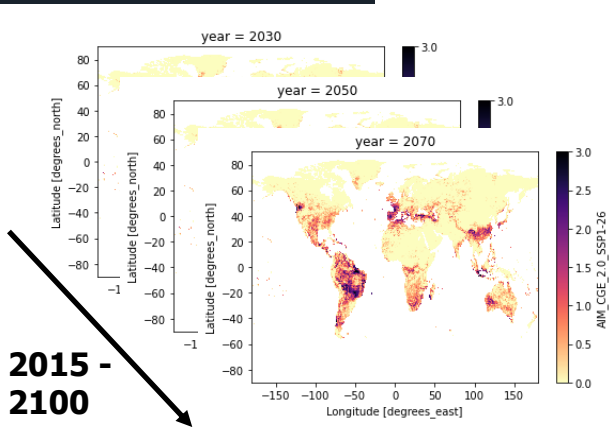
# Community consistent output formats

## Spatial gridded netCDF format



Multiple indicators, one IAM scenario

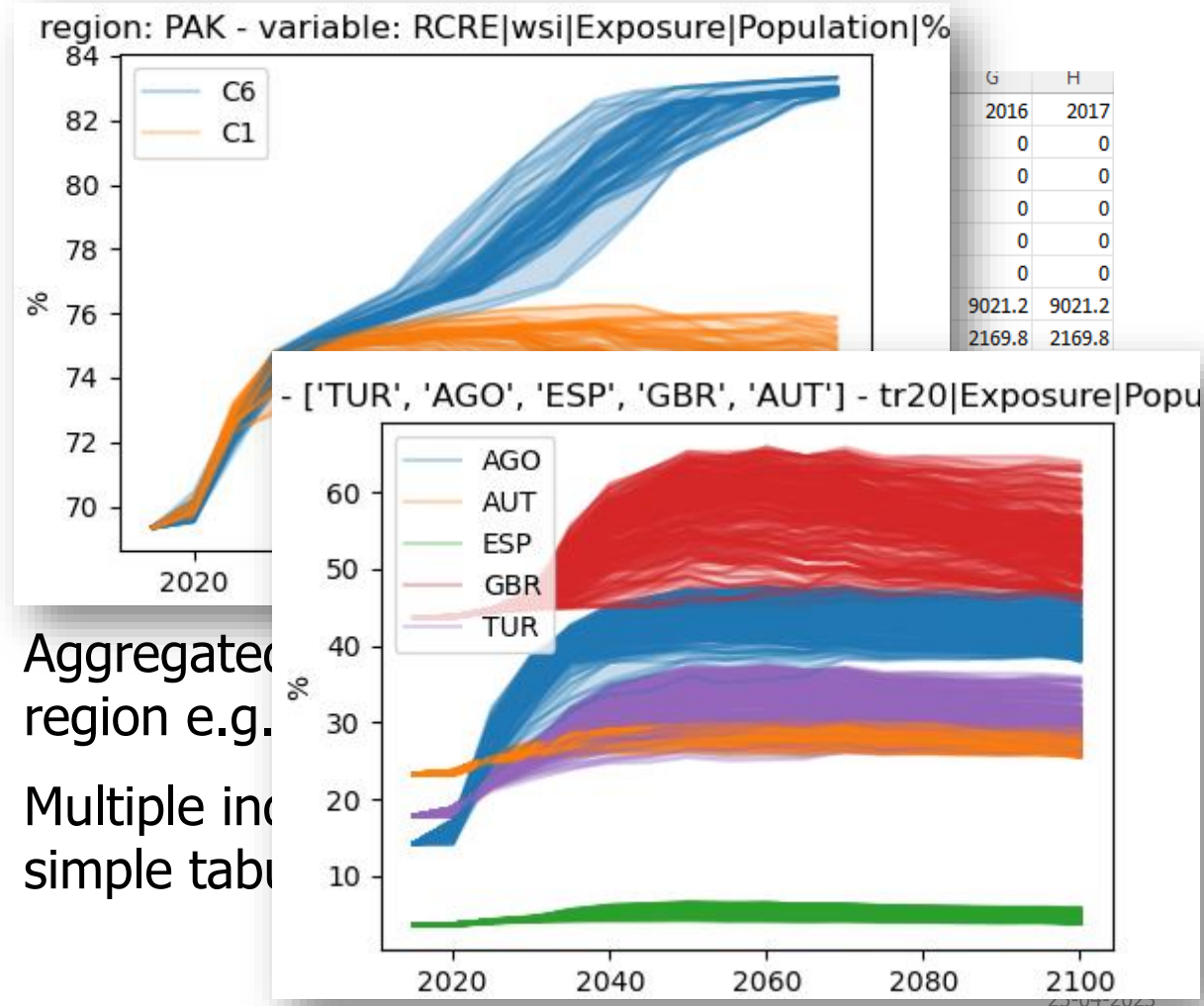
```
<xarray.Dataset>
Dimensions:   (lat: 360, lon: 720, year: 86)
Coordinates:
  * lon       (lon) float64 -179.8 -179.2 -178.8 -178.2 ... 178.8 179.2 179.8
  * lat       (lat) float64 89.75 89.25 88.75 88.25 ... -88.75 -89.25 -89.75
  * year      (year) int32 2015 2016 2017 2018 2019 ... 2097 2098 2099 2100
Data variables: (12/18)
  cdd        (lat, lon, year) float64 ...
  dri        (lat, lon, year) float64 ...
  dri_qtot   (lat, lon, year) float64 ...
  iavar      (lat, lon, year) float64 ...
  iavar_qtot (lat, lon, year) float64 ...
  pr_r10     (lat, lon, year) float64 ...
  ...
  sdd_c      (lat, lon, year) float64 ...
  sdd_c_24p0 (lat, lon, year) float64 ...
  sdd_c_20p0 (lat, lon, year) float64 ...
  sdd_c_18p3 (lat, lon, year) float64 ...
  tr20       (lat, lon, year) float64 ...
  wsi        (lat, lon, year) float64 ...
```



One indicator, multiple IAM scenarios

```
<xarray.Dataset>
Dimensions:   (lat: 360, lon: 720, year: 86)
Coordinates:
  * lon       (lon) float64 -179.8 -179.2 -178.8 ... 179.2 179.8
  * lat       (lat) float64 89.75 89.25 88.75 ... -89.25 -89.75
  * year      (year) int32 2015 2016 2017 2018 ... 2098 2099 2100
Data variables:
  AIM_CGE_2.0_SSP1-26 (lat, lon, year) float64 ...
  GCAM_5.3_SSP_SSP5   (lat, lon, year) float64 ...
```

## IAMC tabular format



Aggregated region e.g. Multiple indicators in simple table

# Conclusions

- Various indicators prepared to develop database of gridded and country-level impacts and exposure
- Scripts for interpolation and re-indexing of datasets
  - Input: GMT trajectory by year (.csv)
  - Output: Impact indicators by year (.csv, netCDF)
- Developed in Python: Xarray + Dask parallelized processing
- Fast for single scenarios – large ensembles like AR6 more difficult
- Uncertainties
  - Socioeconomic uncertainty covered
  - Climate model uncertainty possible but not yet
  - Threshold approach

# Next steps

- Extend to more indicators + vulnerability
- Launch open source
- Facilitate batch processing of IAM scenarios for online data processing and model intercomparison
- Support assessment of unknown emissions scenarios, including overshoot
- Support IPCC WG1-WG2-WG3 integration

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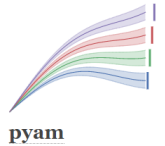
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## Implementation



## Acknowledged support

