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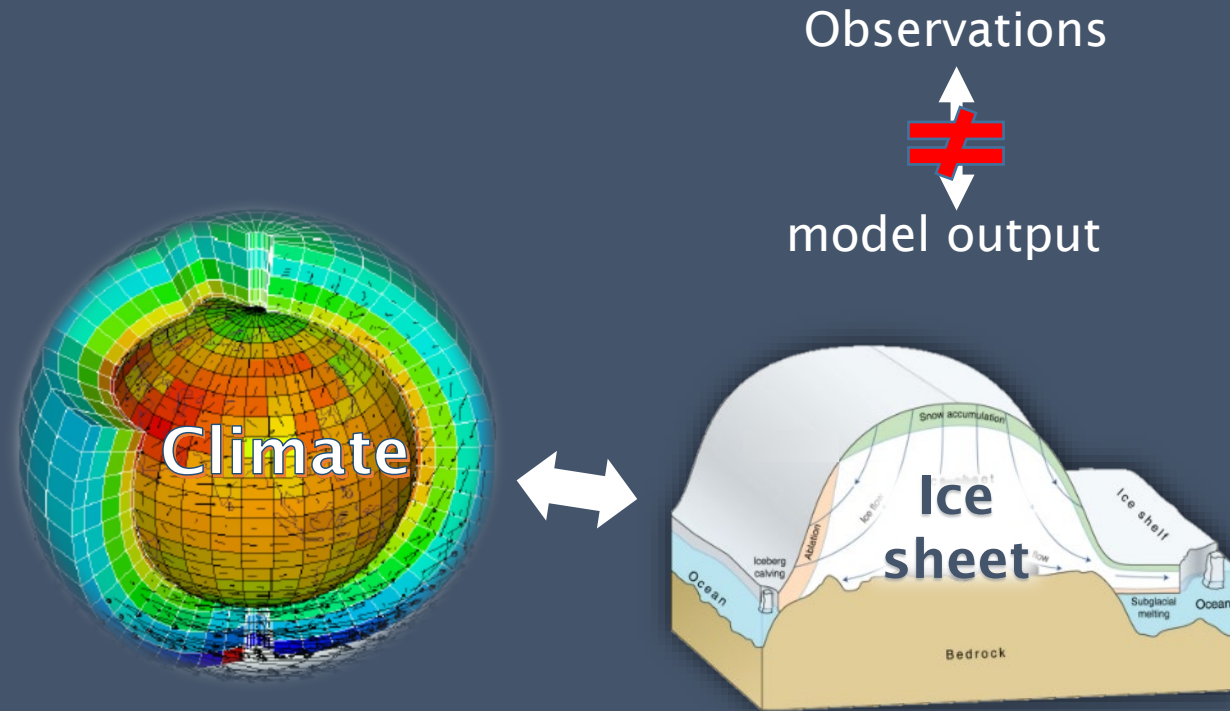
Exploring the complex uncertainties in coupled climate-ice simulations of the Last Glacial Maximum

Lauren Gregoire, UKRI Future Leaders Fellow

Lachlan Astfalk, Niall Gandy, Ruza Ivanovic, Sam Sherriff-Tadano (Leeds), Danny Williamson (Alan Turing Fellow, U of Exeter), Robin Smith, Jonathan Gregory (U of Reading)

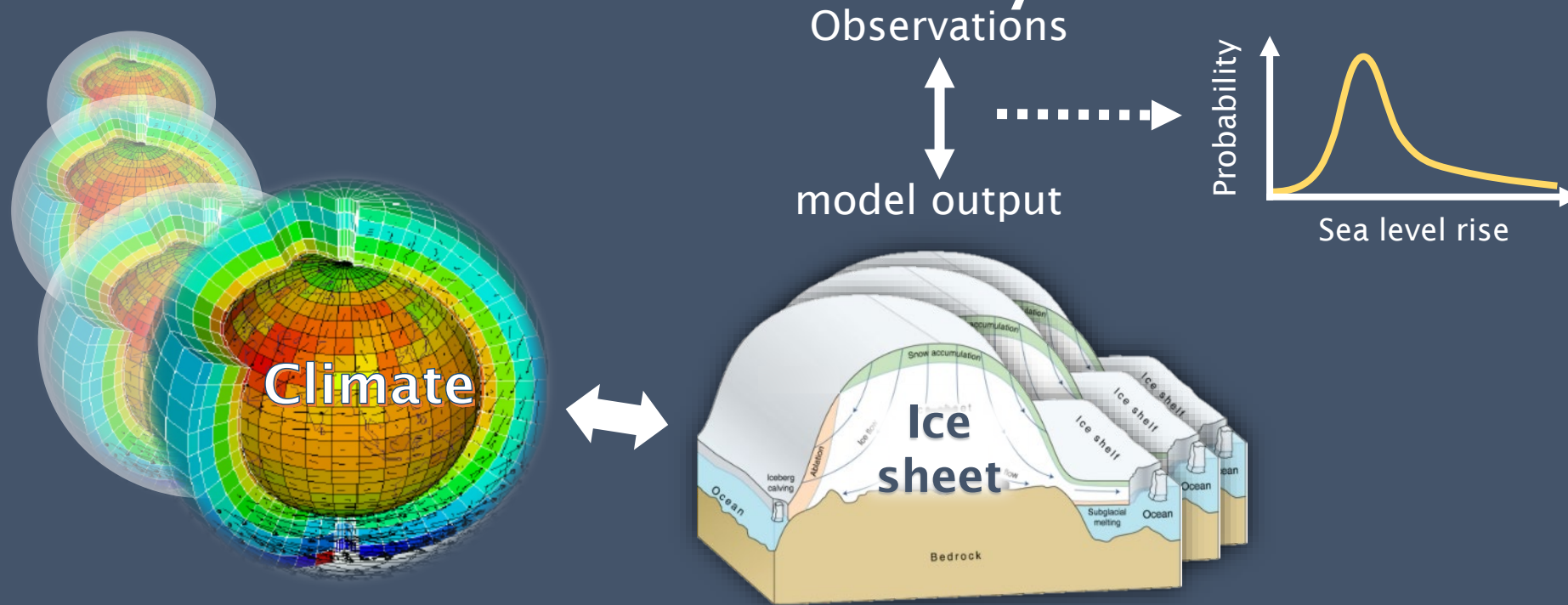
**UK Research
and Innovation**

The Challenge: Biases of simulating co-evolution of Climate and ice sheets



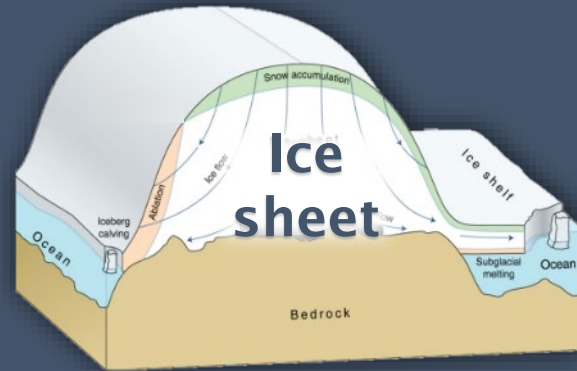
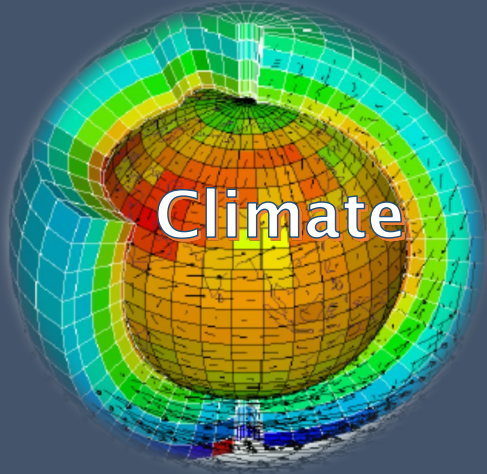
- Small biases amplified by strong ice-climate feedbacks

The Challenge: Tackling climate uncertainty



- Climate is the largest source of uncertainty
- Complex spatio-temporal problem
- Requires new Artificial Intelligence techniques

Model: FAMOUS-Ice



FAMOUS:

Low resolution
Ocean-Atmosphere
Dynamic vegetation
General circulation model
~500 model years per day

- Atmosphere only
- Slab ocean

Ice sheets

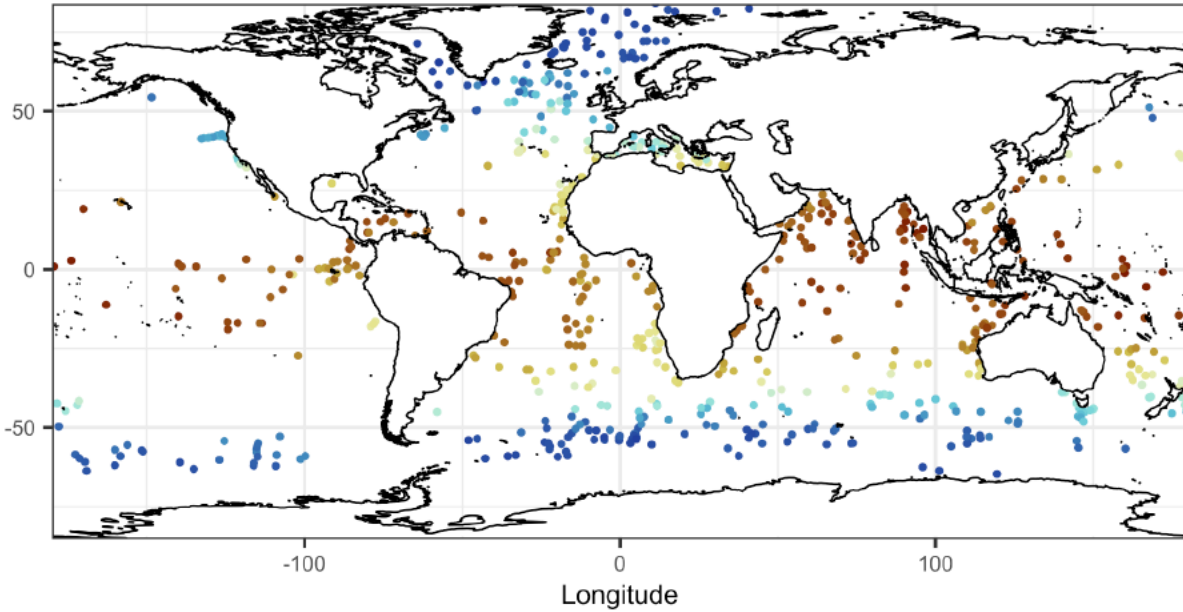
Glimmer: fast (Shallow ice approx.)
BISICLES: efficient
intermediate complexity
marine ice sheet model.

- Energy Mass balance calculation within each grid cell with land ice at 10 different elevations
- SMB downscaled onto ice sheet model surface by linear interpolation.
- Snow/firn layers ; albedo as function of grain size.

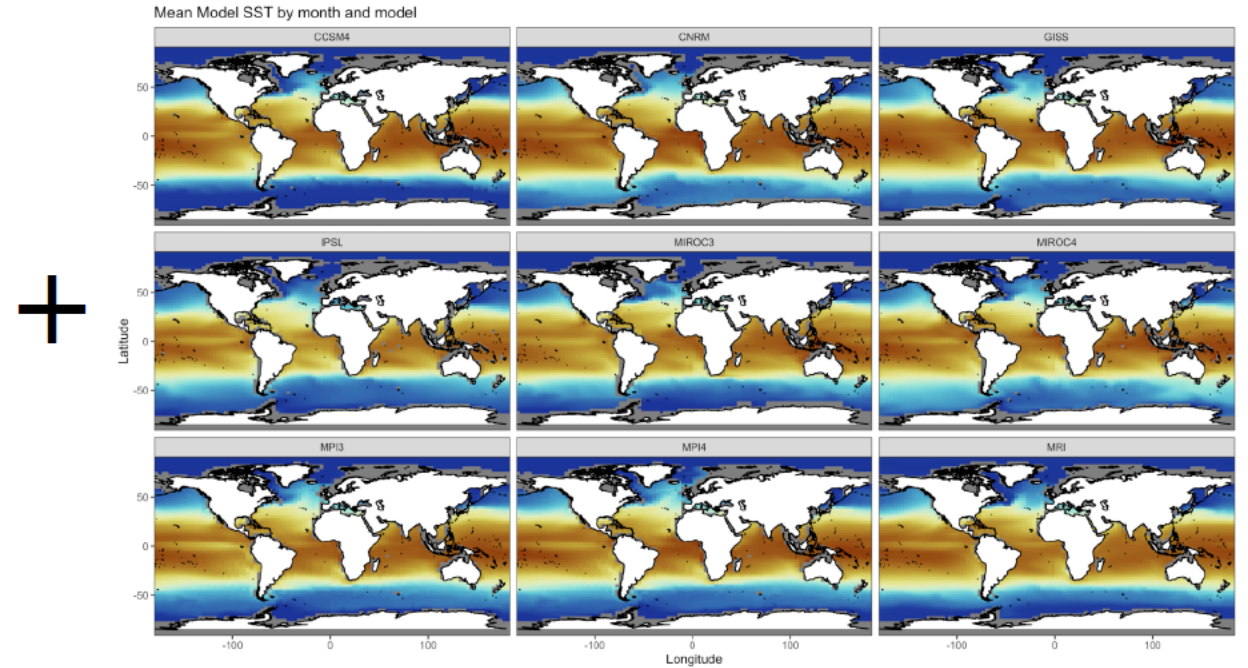
Smith et al. GMD discuss
Gregory et al. The Cryosphere
2020

Generating a range of plausible Sea surface conditions for atmosphere only runs

Data

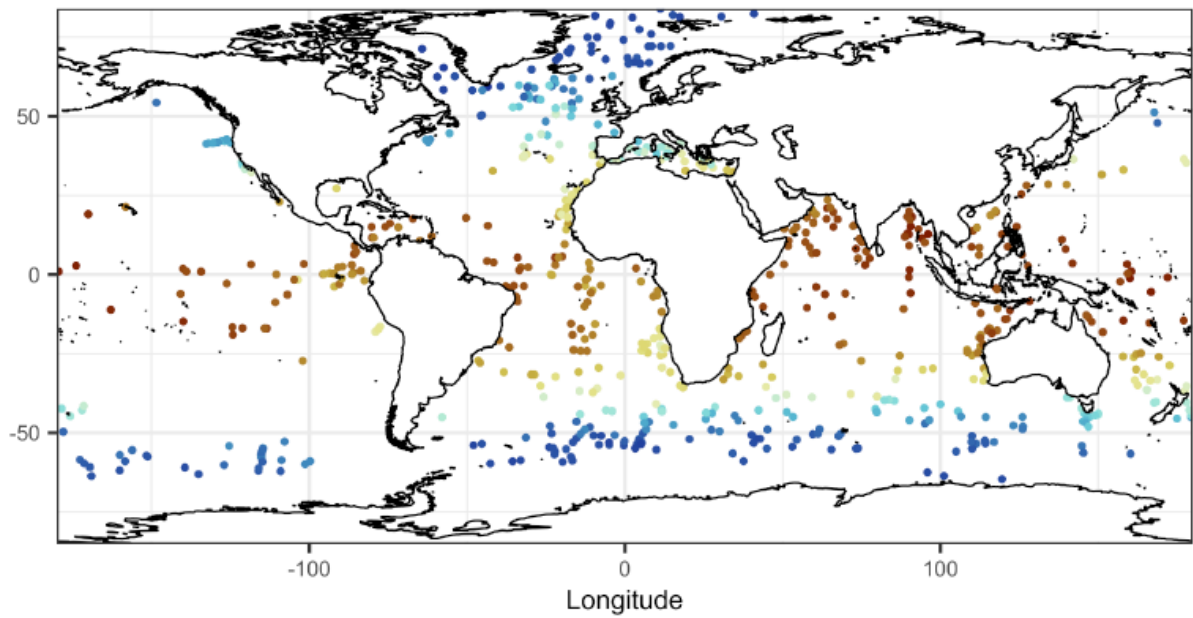


Model Runs

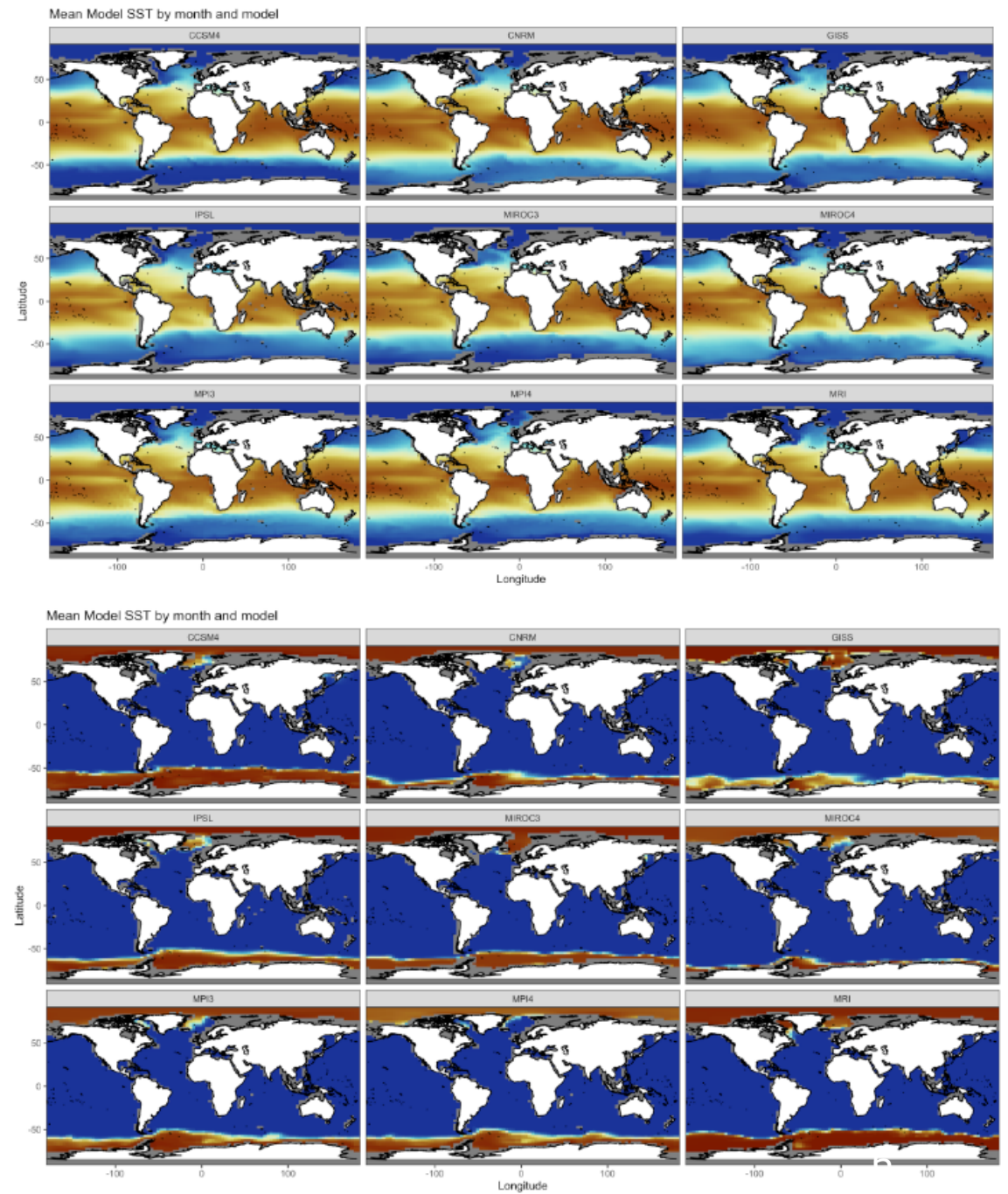


Margo Sea surface Temperatures + Southern Ocean

PMIP3 and PMIP 4 LGM simulations



+

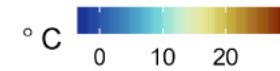
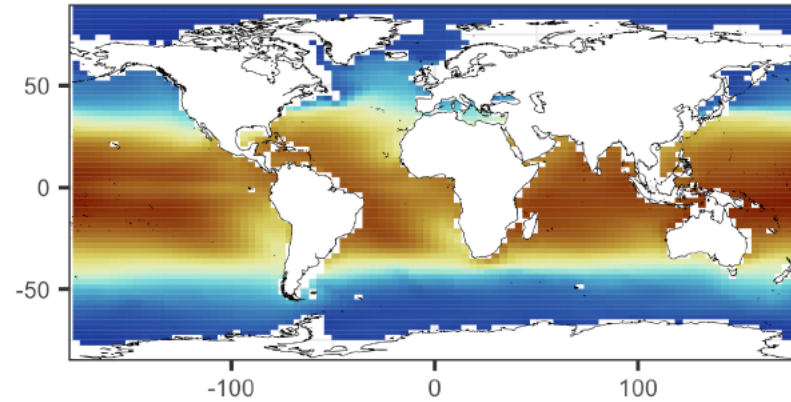
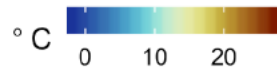
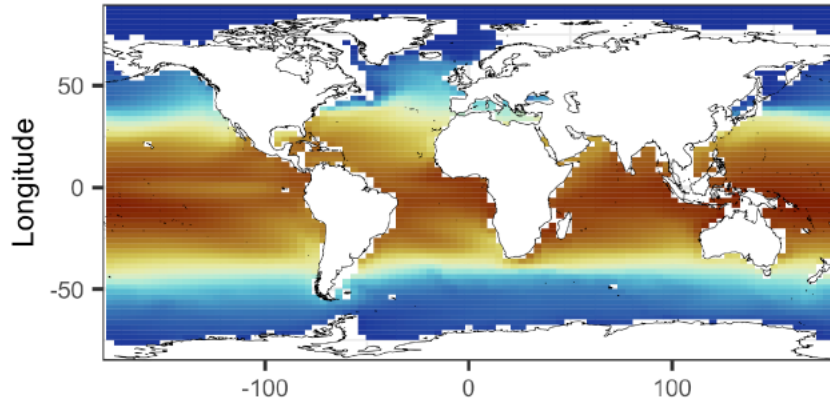


We want a method to update BOTH sea-surface temperature and sea-ice concentration.

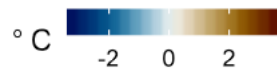
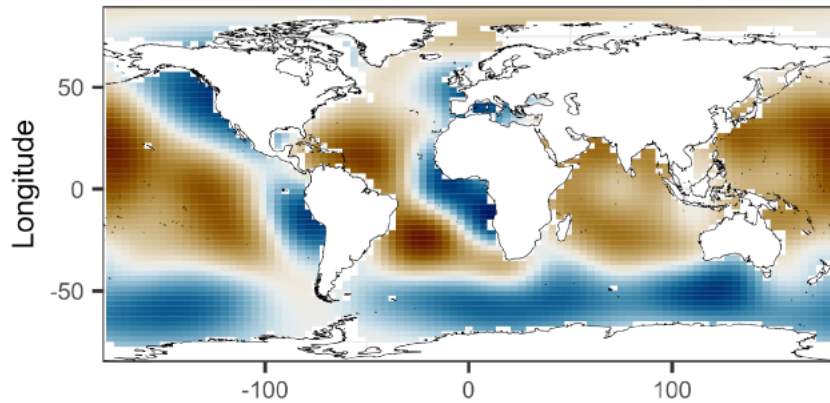


Exchangeable regressions

Multi-model
mean

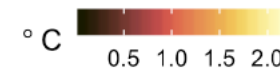
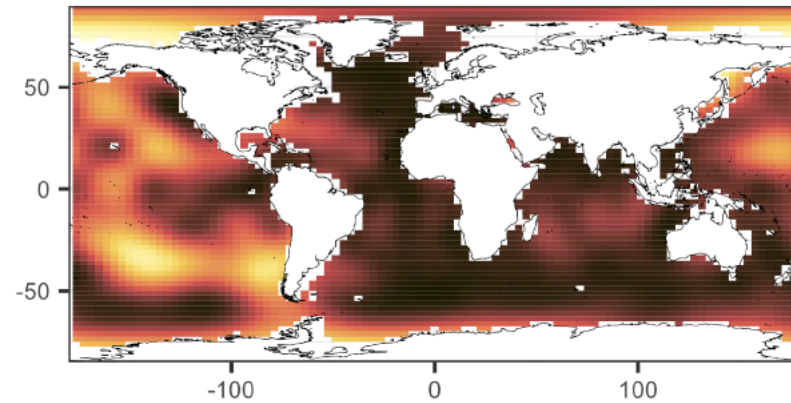


Mean
Correction



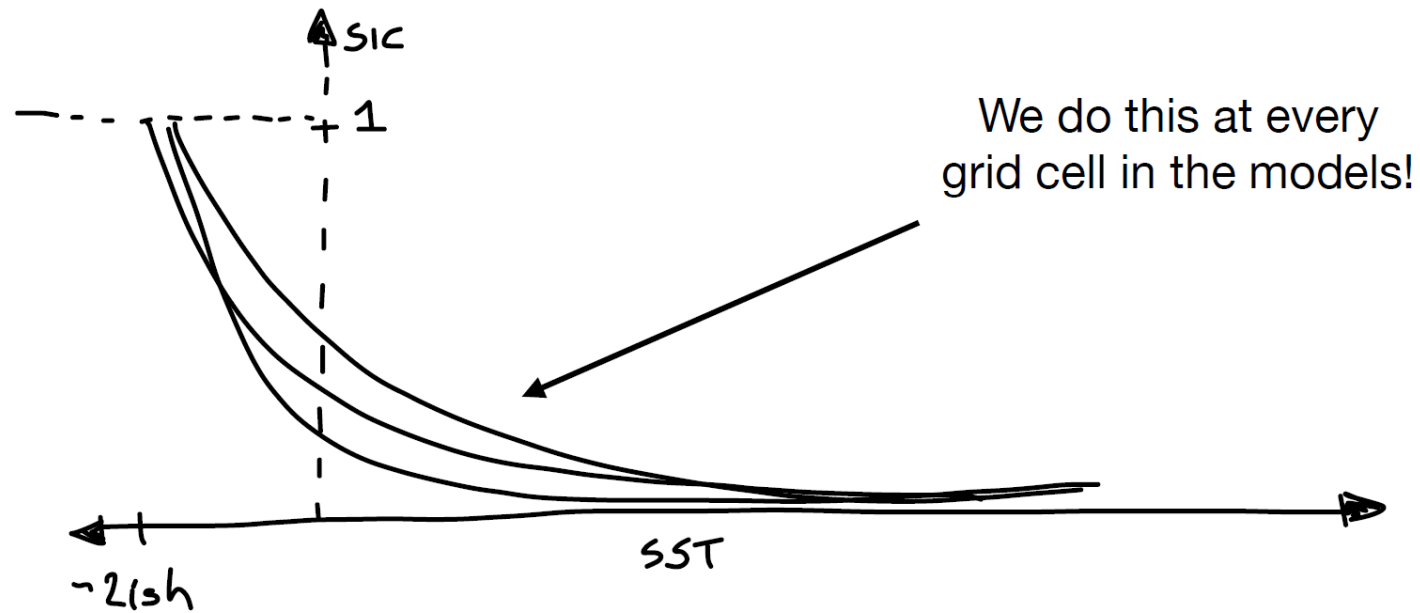
Reconstructed
Mean

Reconstructed
Variance



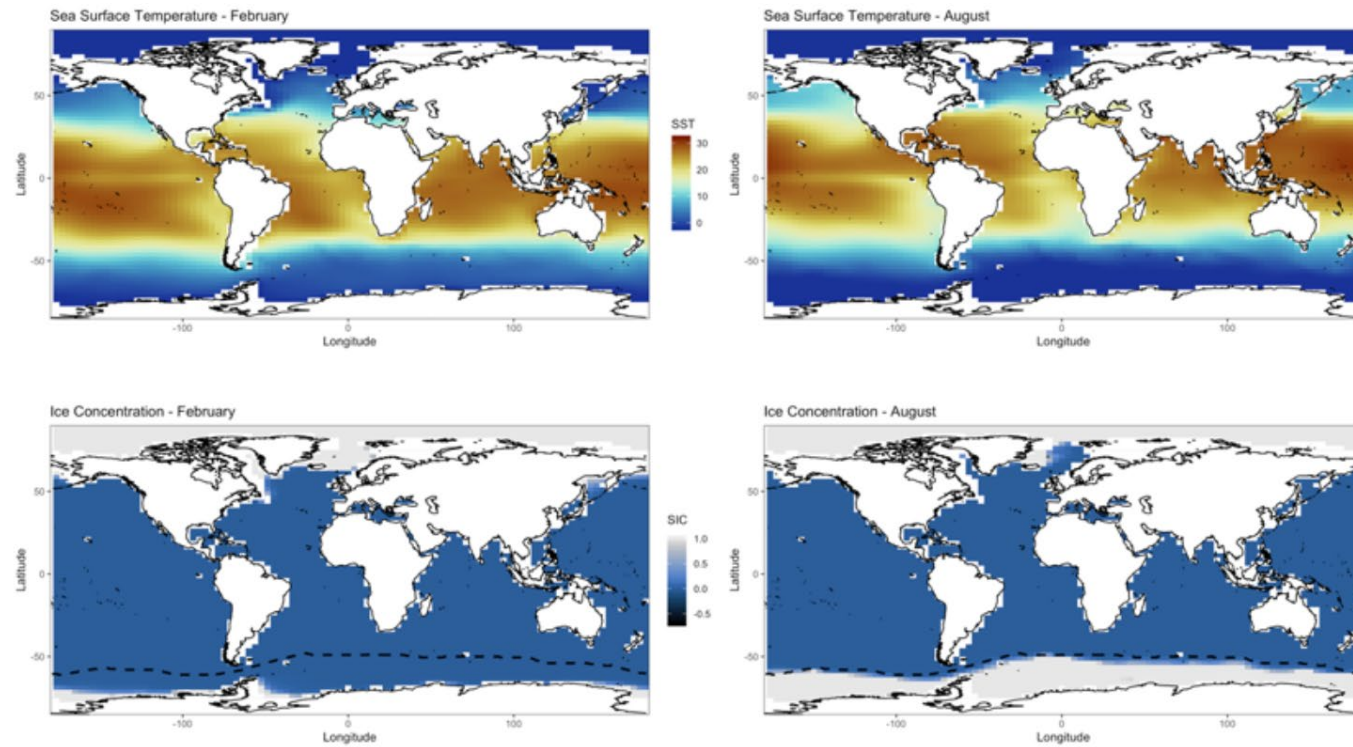
Learning plausible sea ice - SST relationship

But what about sea-ice??



Co-reconstruction of sea surface temperature and Sea ice cover

We use the empirical relationship between SST and SIC to also update SIC in a physically coherent manner

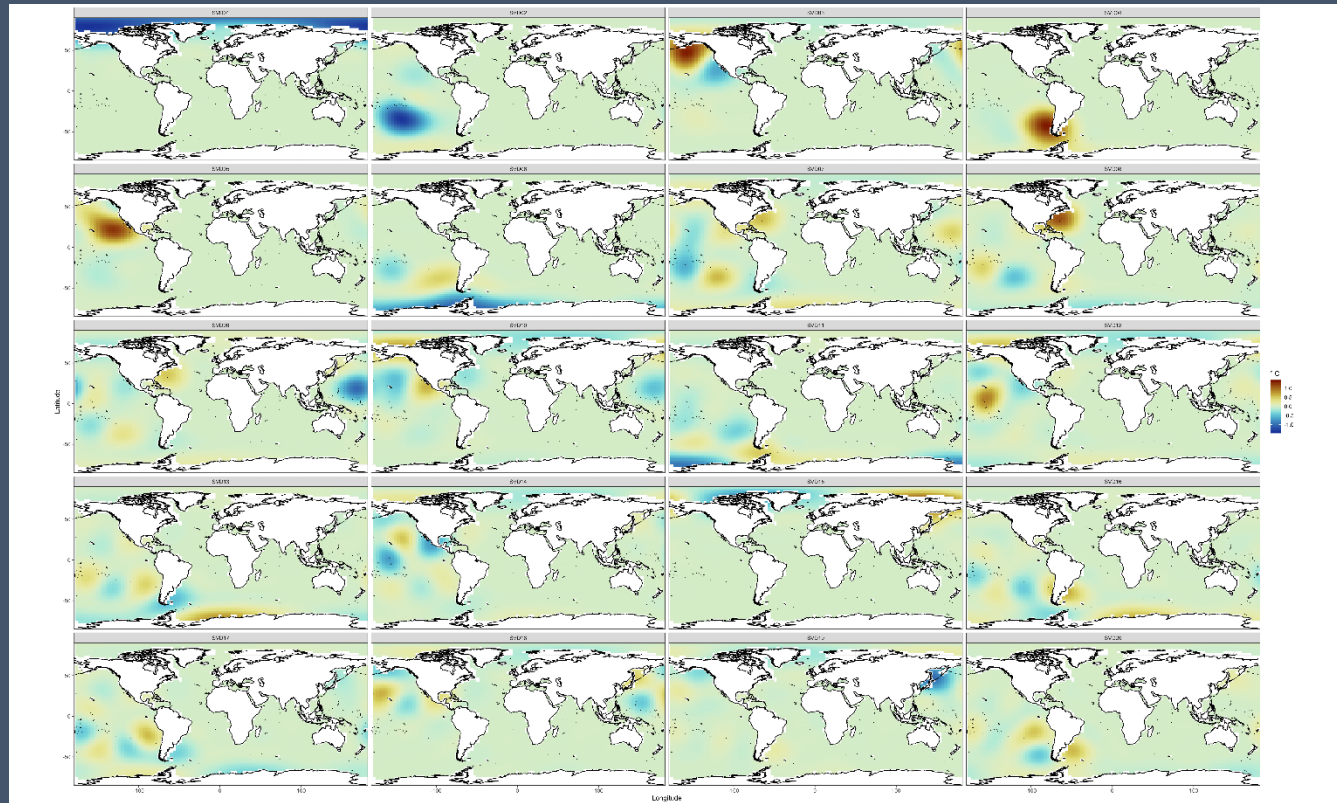
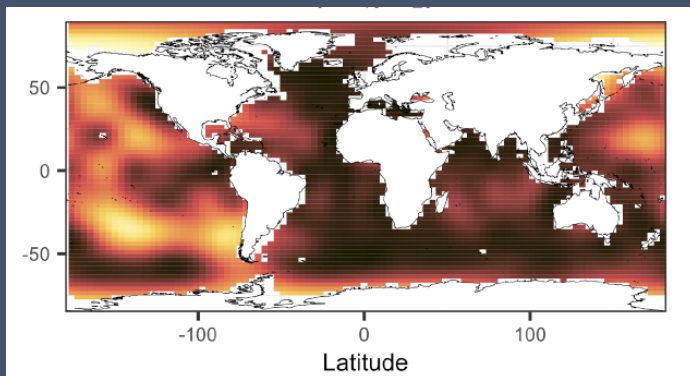


Sampling from an uncertain reconstruction

1. Decomposition of Covariance matrix to select direction of maximum variance and remove small scale patterns.

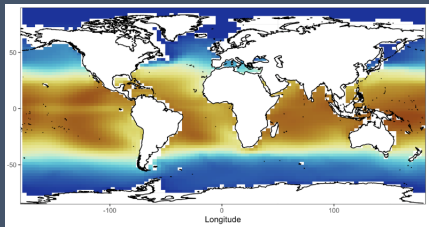
Variance-Covariance Matrix					
	1	2	3	4	5
1	σ^2	COV	COV	COV	COV
2	COV	σ^2	COV	COV	COV
3	COV	COV	σ^2	COV	COV
4	COV	COV	COV	σ^2	COV
5	COV	COV	COV	COV	σ^2

Variance



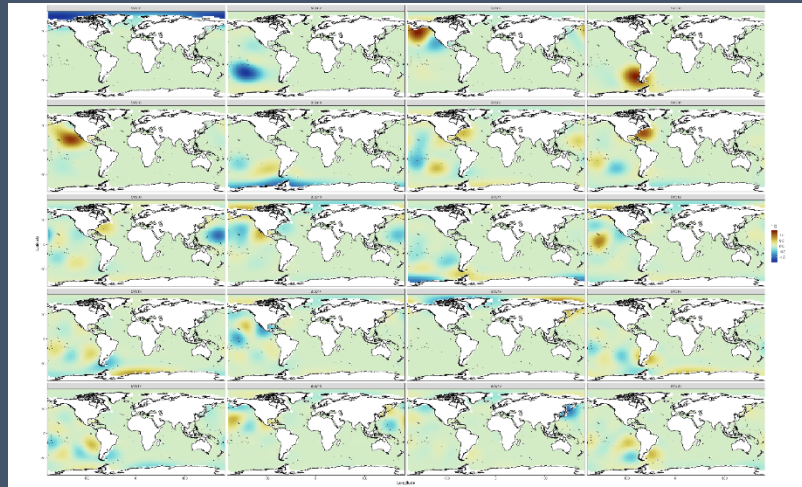
Sampling plausible SST and sea ice

Reconstructed Mean



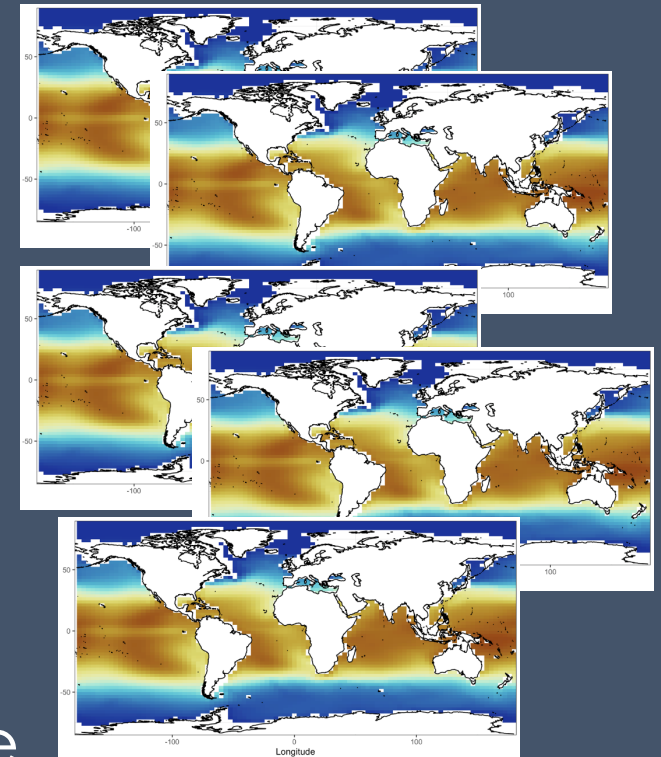
+ a_i

Linear combination of principle components



=

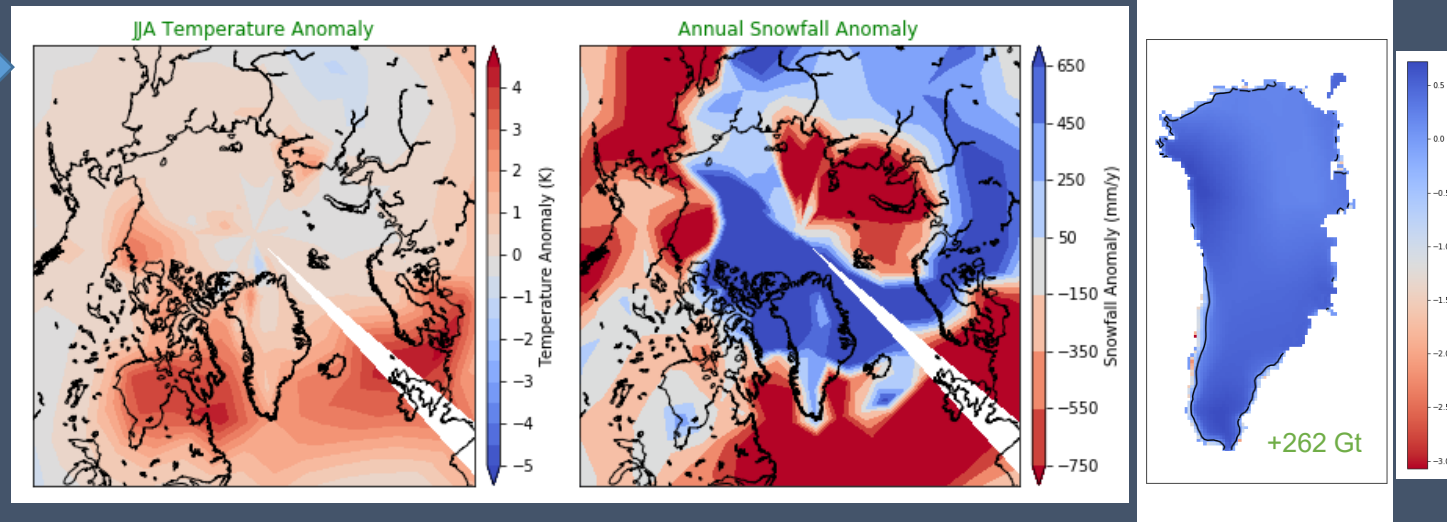
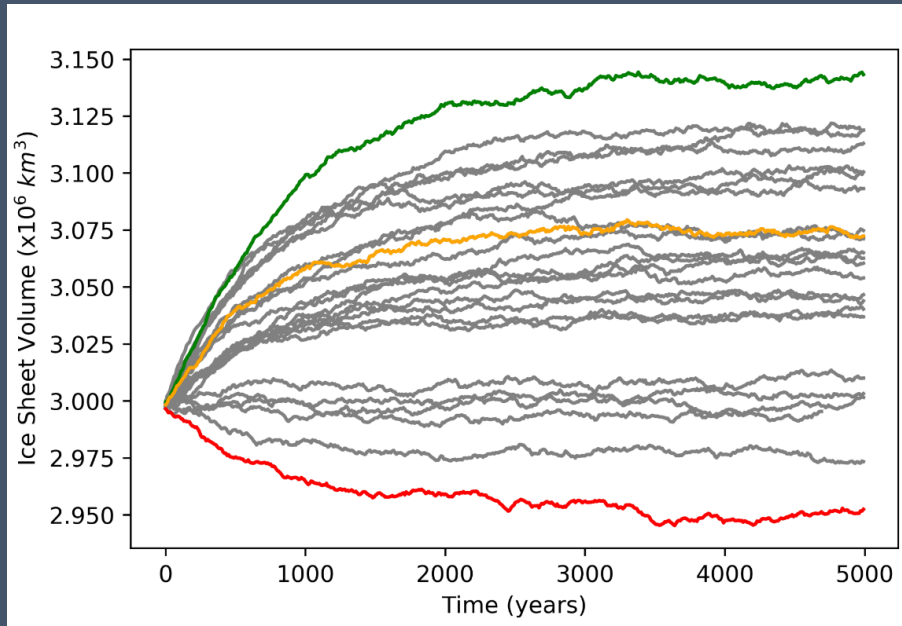
Ensemble of plausible SSTs and Sea ice



We randomly sample parameters a_i to generate ensembles of plausible boundary conditions for FAMOUS

Initial results: Greenland Ice Sheet

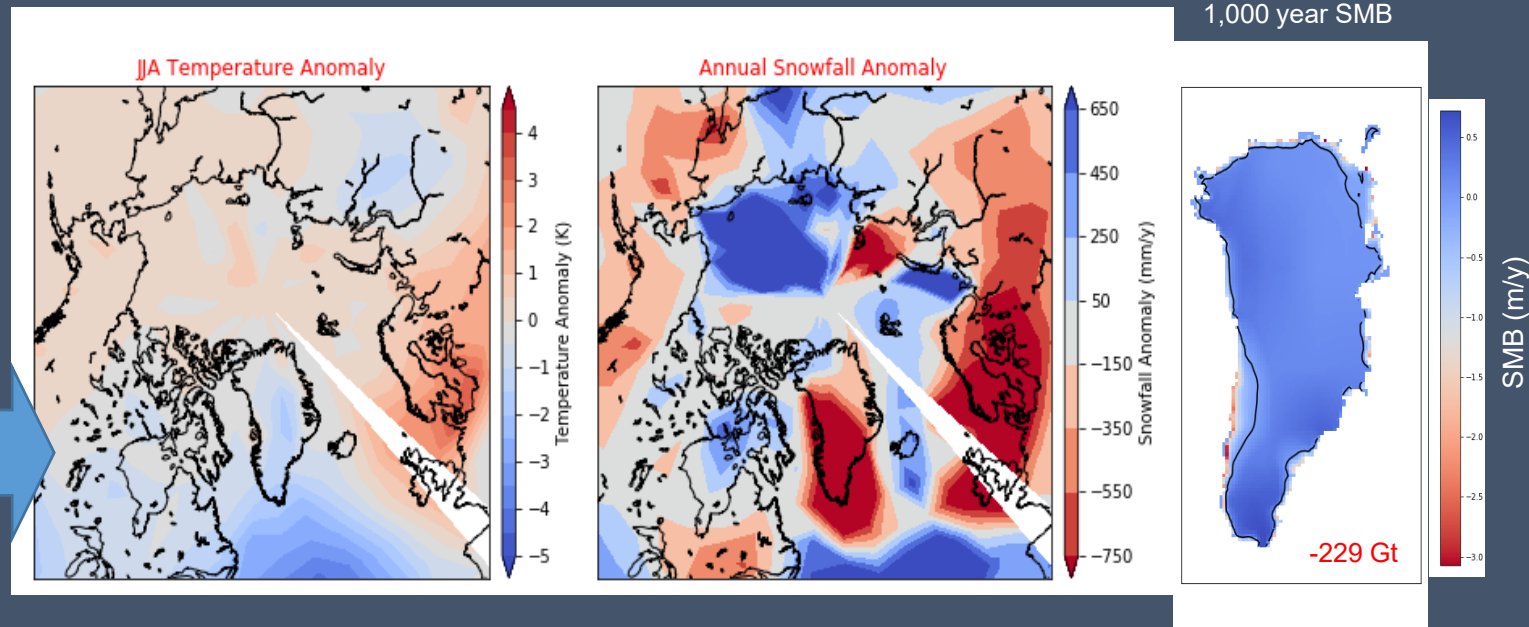
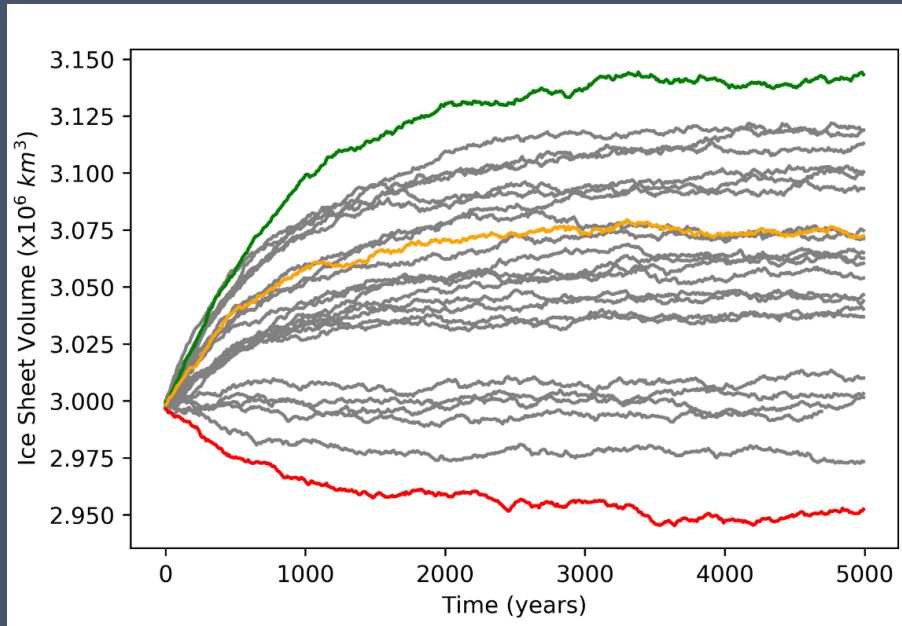
- Test ensemble; Greenland Ice Volumes varying due to SST+ sea ice concentration fields



- High ice sheet volume due to higher snowfall

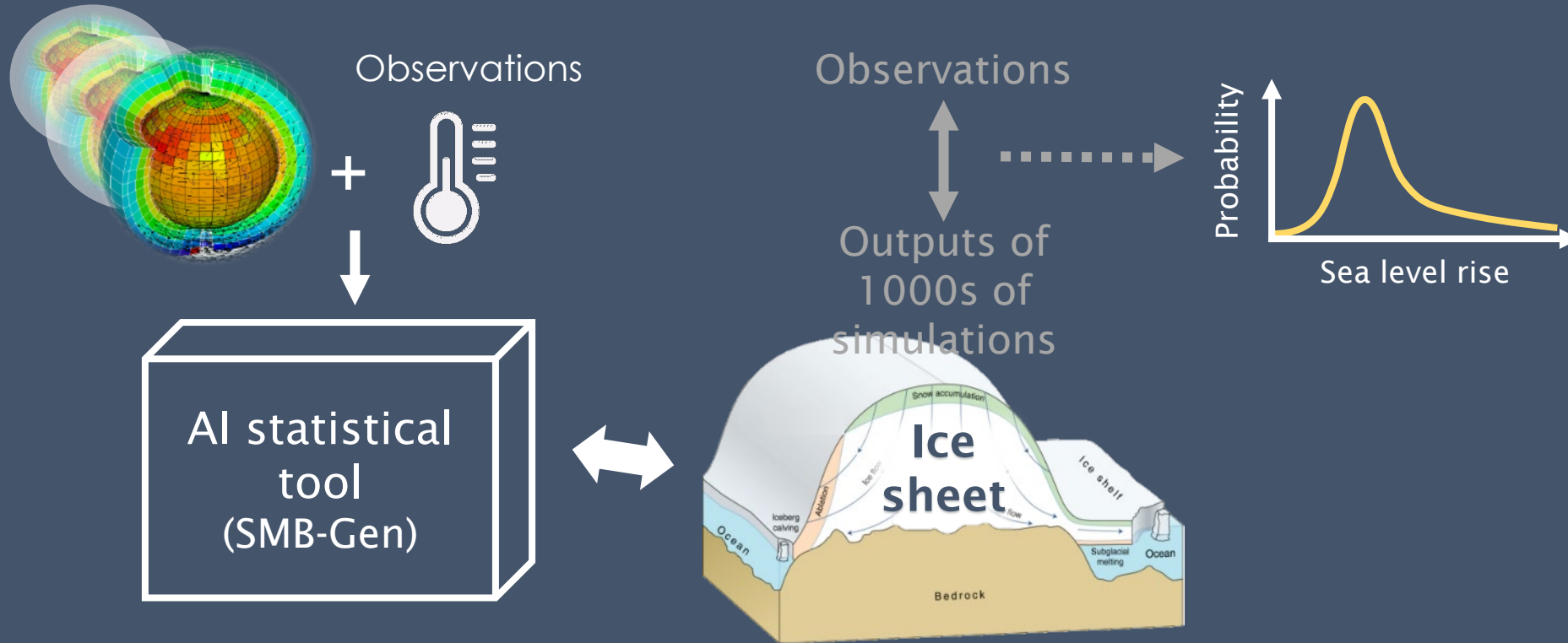
Initial results: Greenland Ice Sheet

- Test ensemble; Greenland Ice Volumes varying due to SST+ sea ice concentration fields



- Low ice sheet volume due to low snowfall

Developing a Surface mass balance generator (SMB-Gen)



- Novel Artificial intelligence tool
- Use ensembles of simulations to train statistical emulator of surface mass balance

Summary and perspective

- Bias reduction by assimilating data on sea surface temperature (SST) and sea ice concentrations (SIC)
- Bayes Linear method to jointly reconstruct SST and SIC and sample ensemble of plausible states.
- Next steps:
 - Ensemble of FAMOUS-Ice simulations with Northern Hemisphere ice sheets
 - Simulations of the deglaciation and 8.2 kyr event with fixed and slab ocean
 - Gaussian Process emulation of ice sheet volume and 3D SMB.
- **Quantitatively relate past and future ice sheet evolution.**