

Uncertainties in climate economic models – a machine learning approach

Presentation for GDR Workshop "Interfaces dans le système climatique"

Yushan LIU

Supervised by Emmanuel Gobet and Gauthier Vermandel

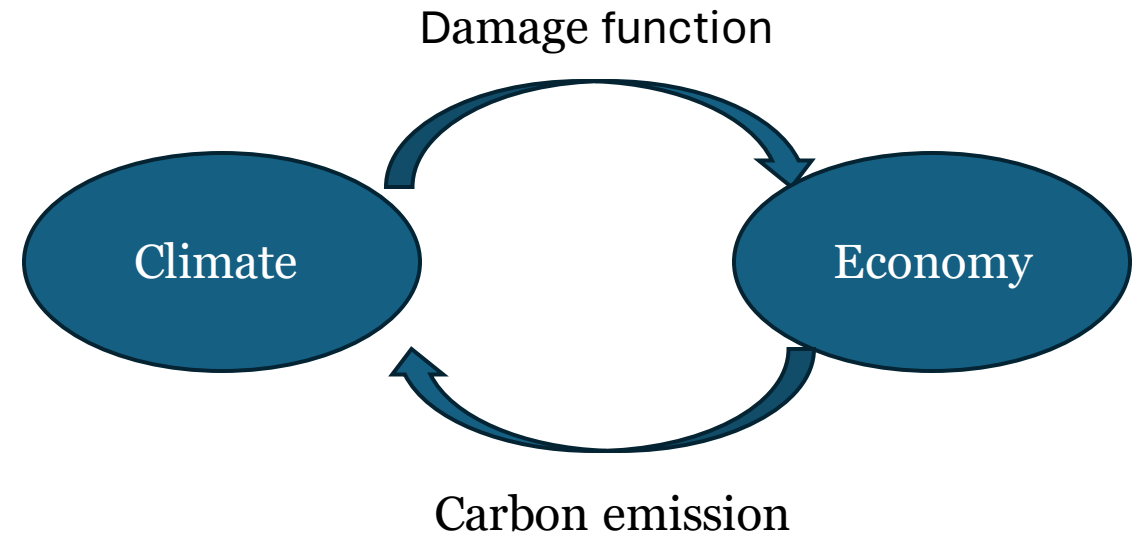
CMAP Ecole Polytechnique and Chair Stress test BNP Paribas



Scopes and Agenda

Context

- Global scale modeling and yearly time scale
- Simple Climate Models (SCMs)
- Optimal control problem



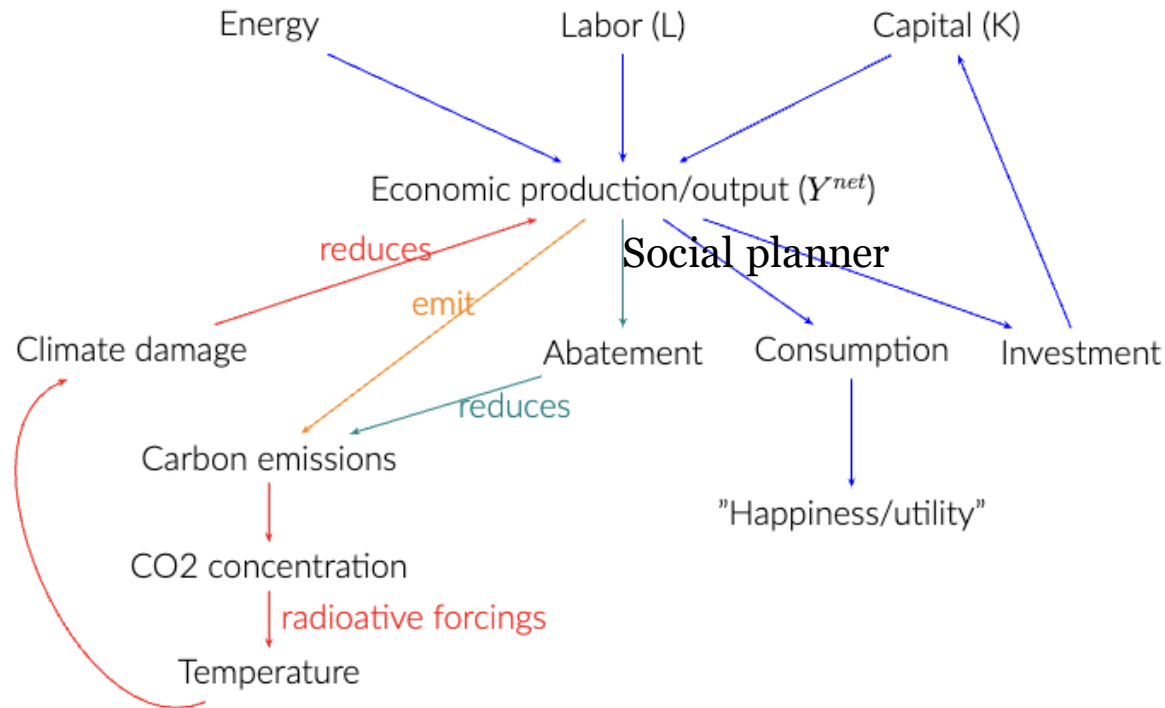
Content

- DICE (Dynamic Integrated Climate Economic) Model by William Nordhaus
- Update in DICE 2024: From Energy Balance Model to FAIR model
- Bayesian FAIR model
- Conclusion and perspectives

DICE (Dynamic Integrated Climate Economic) Model

- 3 boxes for carbon Masses $\mathbf{M}(t)$
- 2 layers for temperature $\mathbf{T}(t)$

$$\mathbf{M} = \begin{bmatrix} M^{AT} \\ M^{UO} \\ M^{LO} \end{bmatrix}, \mathbf{T} = \begin{bmatrix} T^{AT} \\ T^{OC} \end{bmatrix}$$



Continuous-time DICE formulation

$$\max_{C(t), \mu(t)} \int_{t=t_0}^{\infty} \frac{U(t)}{(1+\rho)^t} dt, \quad \text{where } U(t) = L(t) \left(\frac{\left(\frac{C(t)}{L(t)}\right)^{1-\alpha} - 1}{1-\alpha} \right)$$

subject to

$$\frac{\partial \mathbf{M}(t)}{\partial t} = \mathbf{B}\mathbf{M}(t) + \mathbf{B}_E \left(\underbrace{\Gamma(t)(1-\mu(t))Y^{net}(t)}_{\text{Industrial emission}} + E^{LAND}(t) \right), \quad (1)$$

$$\frac{\partial \mathbf{T}(t)}{\partial t} = \mathbf{C}_T \mathbf{T}(t) + \left(F_{2x} \frac{\log\left(\frac{M^{AT}(t)}{M_{EQ}^{AT}}\right)}{\log(2)} + F^{EX}(t) \right), \quad (2)$$

$$\frac{\partial K(t)}{\partial t} = \log(1-\delta_K)K(t) + \underbrace{\left(1 - \Lambda_t \mu(t)^{\alpha_{abate}} - \psi_2 (T^{AT}(t))^2 \right) Y^{net}(t) - C(t)}_{\text{Investment}} \quad (3)$$

...

(4)

DICE (Dynamic Integrated Climate Economic) Model

- 3 boxes for carbon Masses $\mathbf{M}(t)$
- 2 layers for temperature $\mathbf{T}(t)$

$[M^{AT}]$

Journal of the Association of Environmental and Resource Economists > Volume 8, Number 5

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Are Economists Getting Climate Dynamics Right and Does It Matter?

Simon Dietz, Frederick van der Ploeg, Armon Rezai, and Frank Venmans

CO2 concentration

↓ radioactive forcings

Temperature

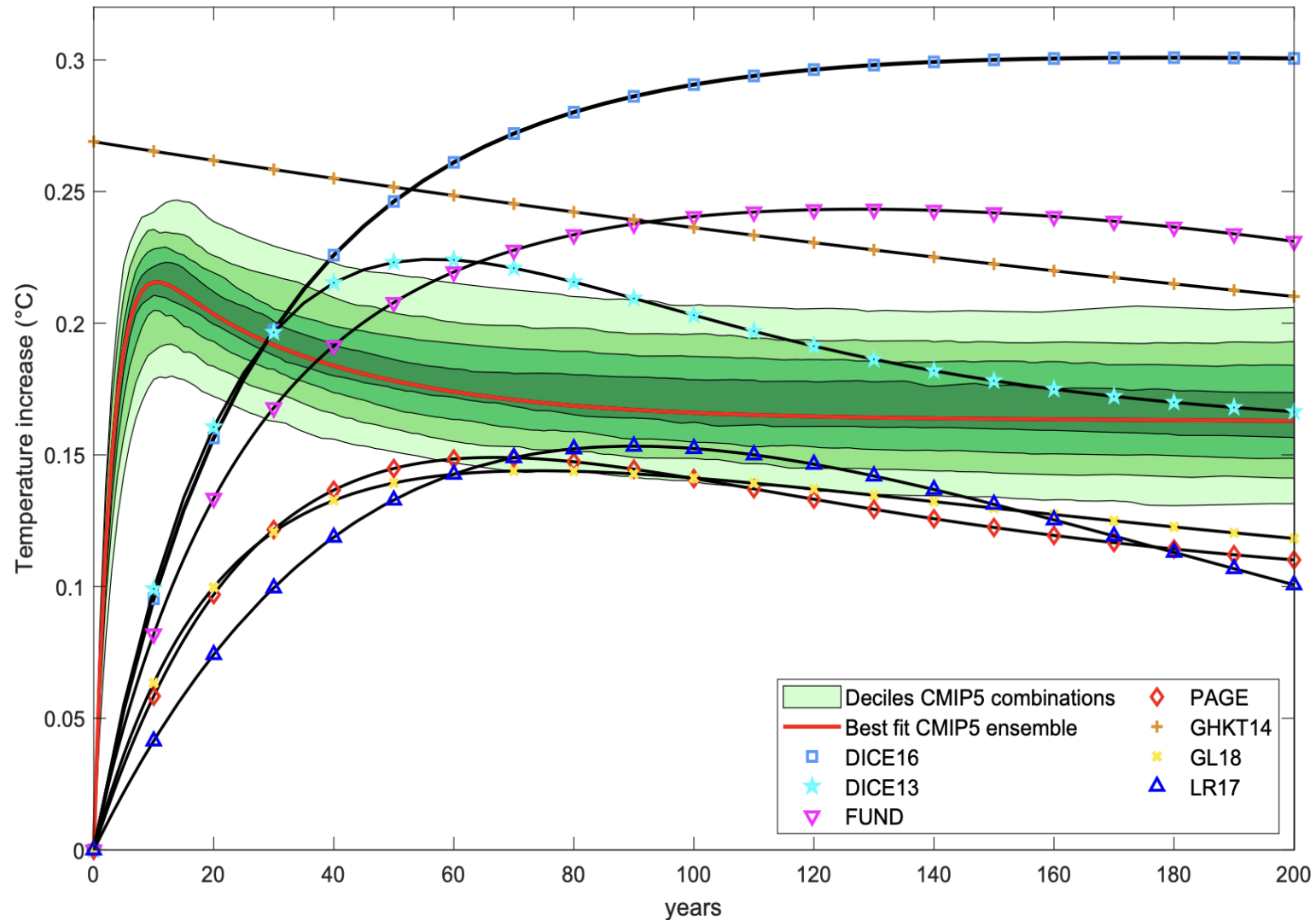
$$\frac{\partial \mathbf{T}(t)}{\partial t} = \mathbf{C}_T \mathbf{T}(t) + \left(F_{2x} \frac{\log \left(\frac{M^{AT}(t)}{M_{EQ}^{AT}} \right)}{\log(2)} + F^{EX}(t) \right), \quad (2)$$

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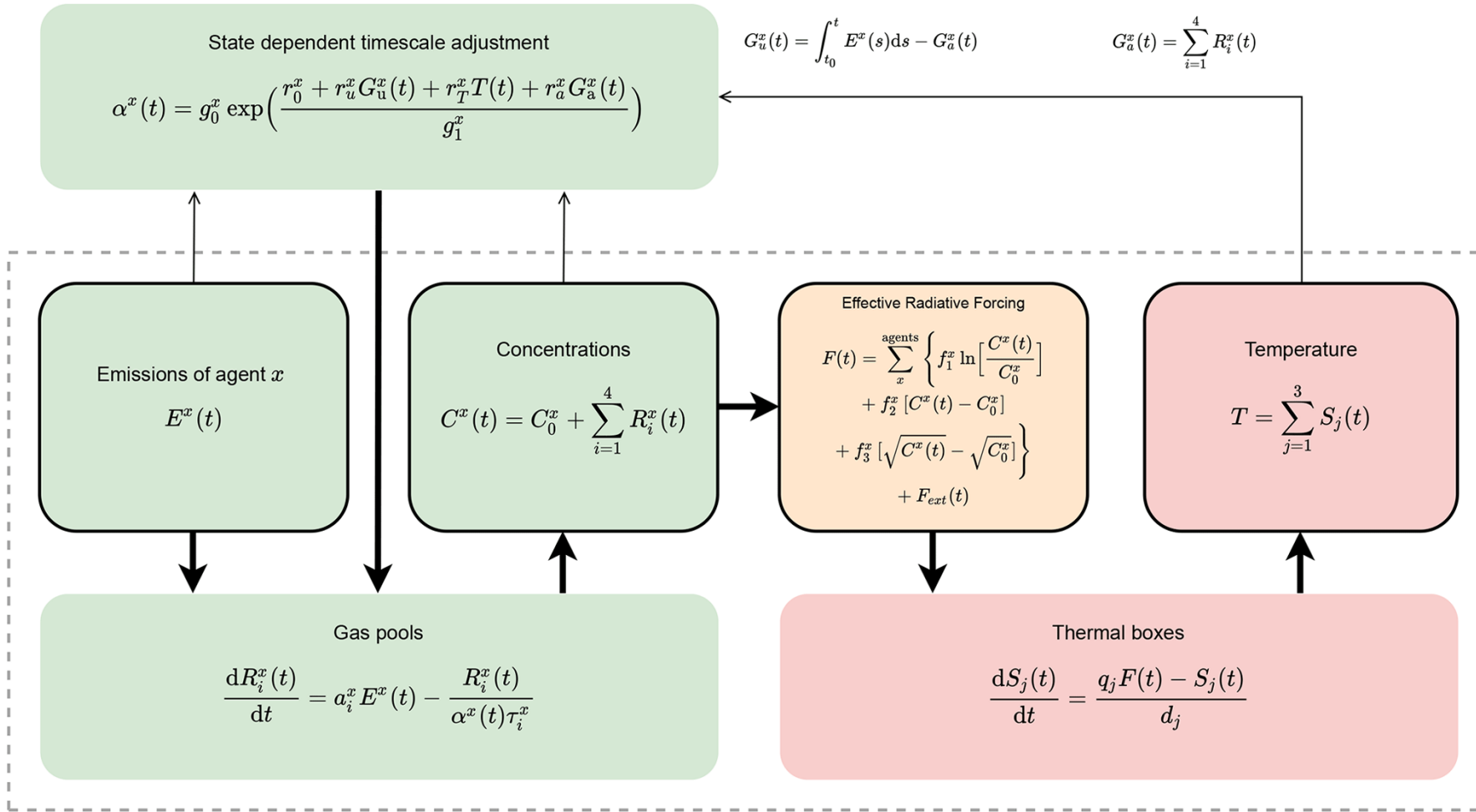
$$\dots \quad (4)$$

Dietz, S., van der Ploeg, F., Rezai, A., & Venmans, F. (2021). Are economists getting climate dynamics right and does it matter?. *Journal of the Association of Environmental and Resource Economists*, 8(5), 895-921.

Are Economists Getting Climate Dynamics Right?

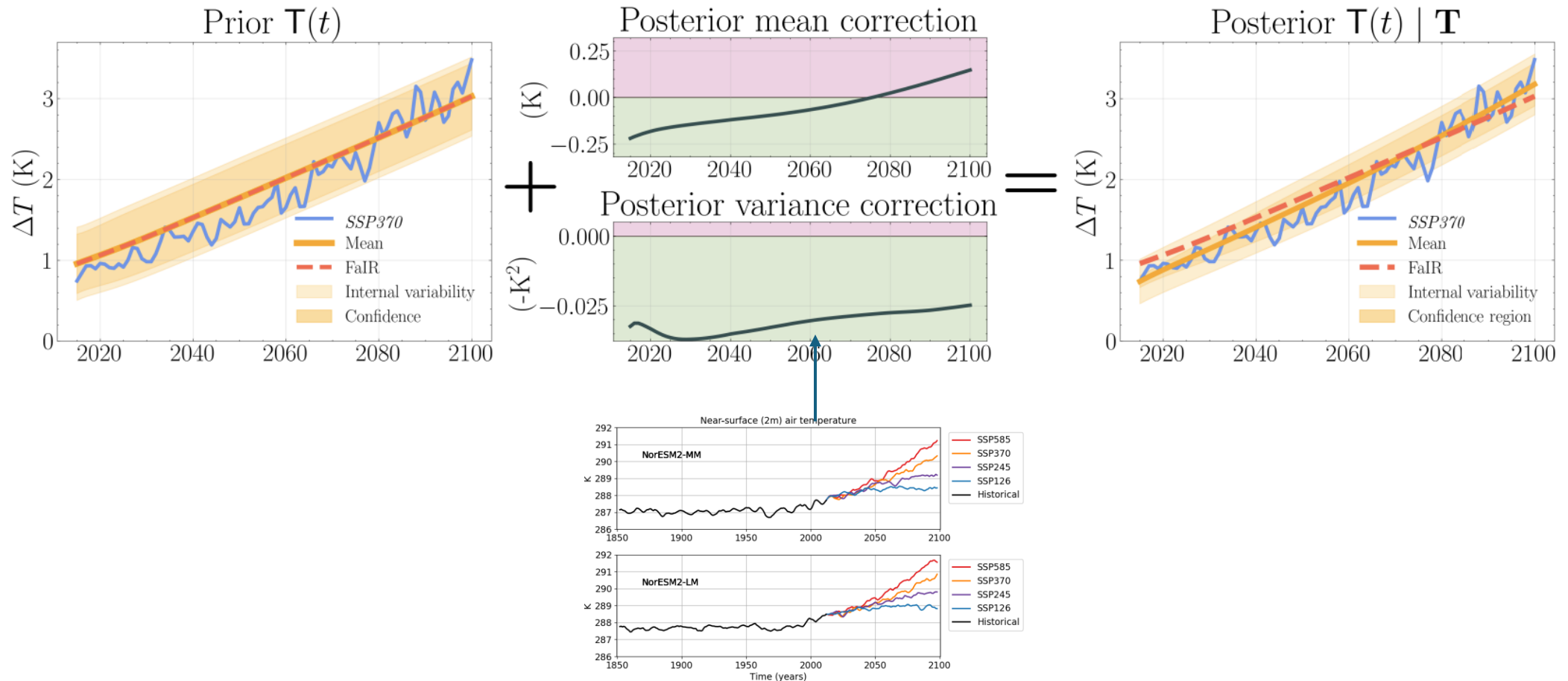


Update in DICE 2024: From Energy Balance Model to FAIR model



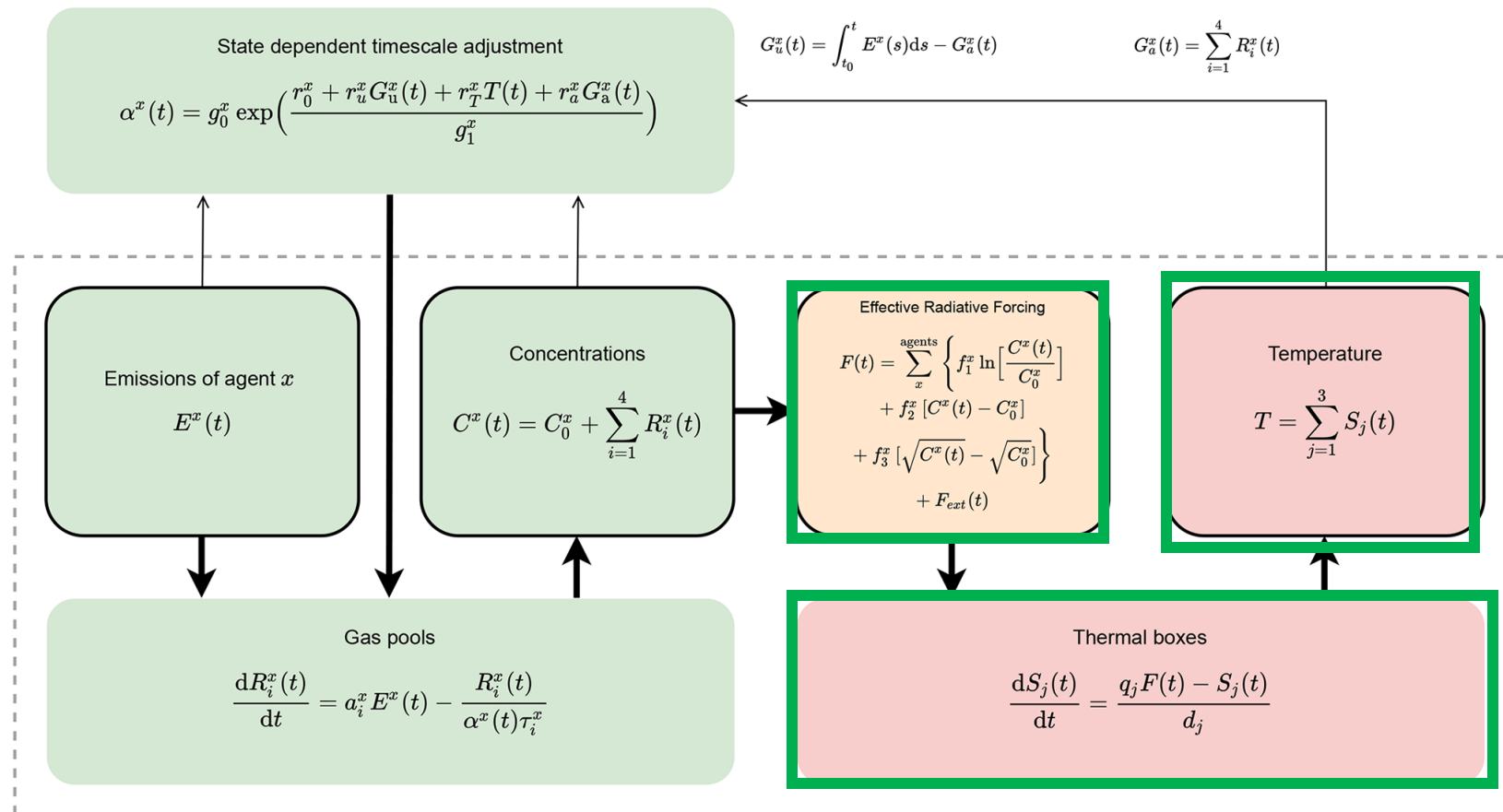
Bayesian FAIR model: Using Gaussian processes

$$\begin{bmatrix} f(t_1) \\ f(t_2) \end{bmatrix} \sim \mathcal{N} \left(\begin{bmatrix} m(t_1) \\ m(t_2) \end{bmatrix}, \begin{bmatrix} k(t_1, t_1) & k(t_1, t_2) \\ k(t_2, t_1) & k(t_2, t_2) \end{bmatrix} \right).$$



Summary

- Epistemic uncertainty in climate models used for climate-economic models need to be quantified
- Bayesian approach is already implemented for FAIR model uncertainty quantification but only partially



Thank you for your attention! Questions?

And Does it Matter?

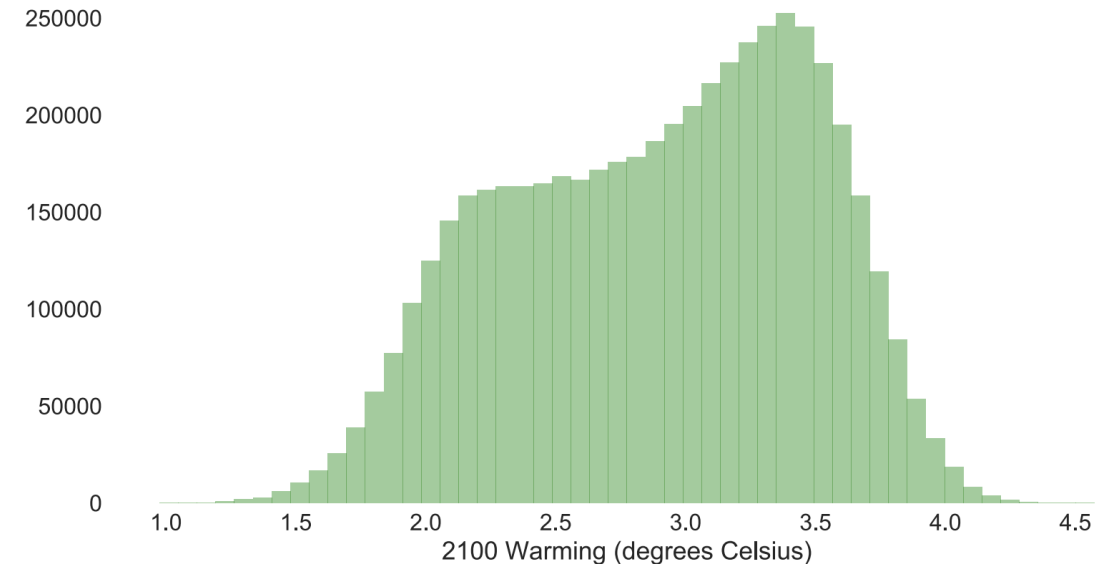
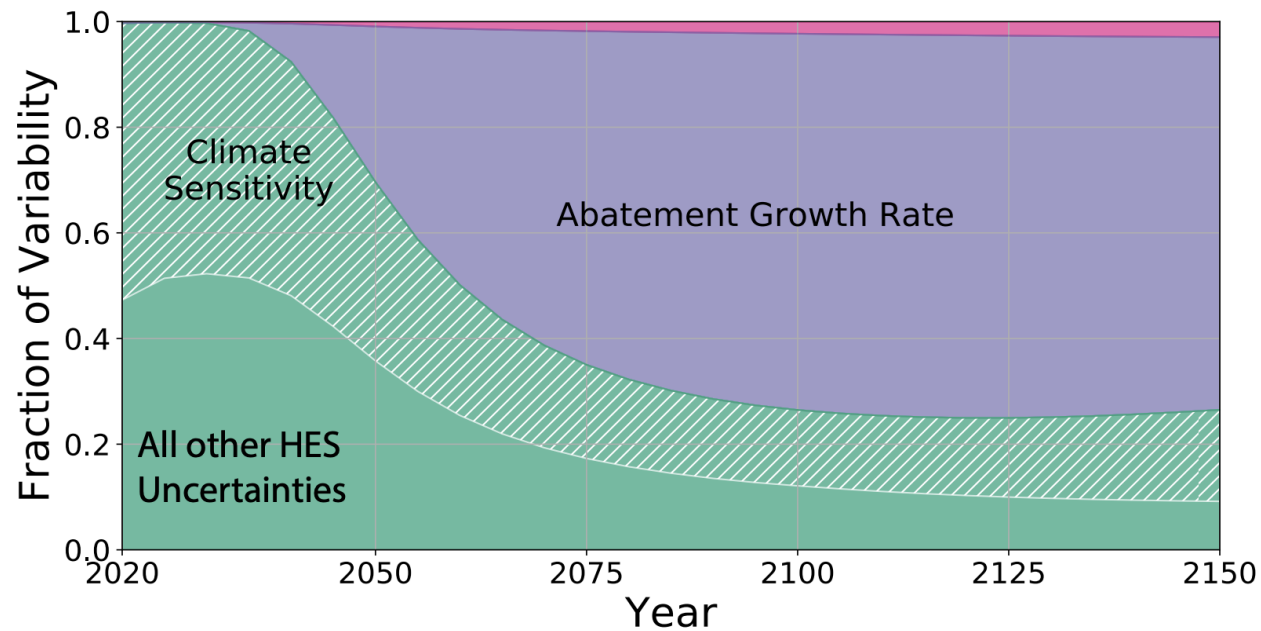
nature
climate change

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<https://doi.org/10.1038/s41558-019-0426-8>

Robust abatement pathways to tolerable climate futures require immediate global action

J. R. Lamontagne^{1*}, P. M. Reed², G. Marangoni^{3,6}, K. Keller^{3,4} and G. G. Garner⁵



- How to study parameter uncertainties of the climate model?
- Simulations vs Stochastic Optimal control?

Appendix

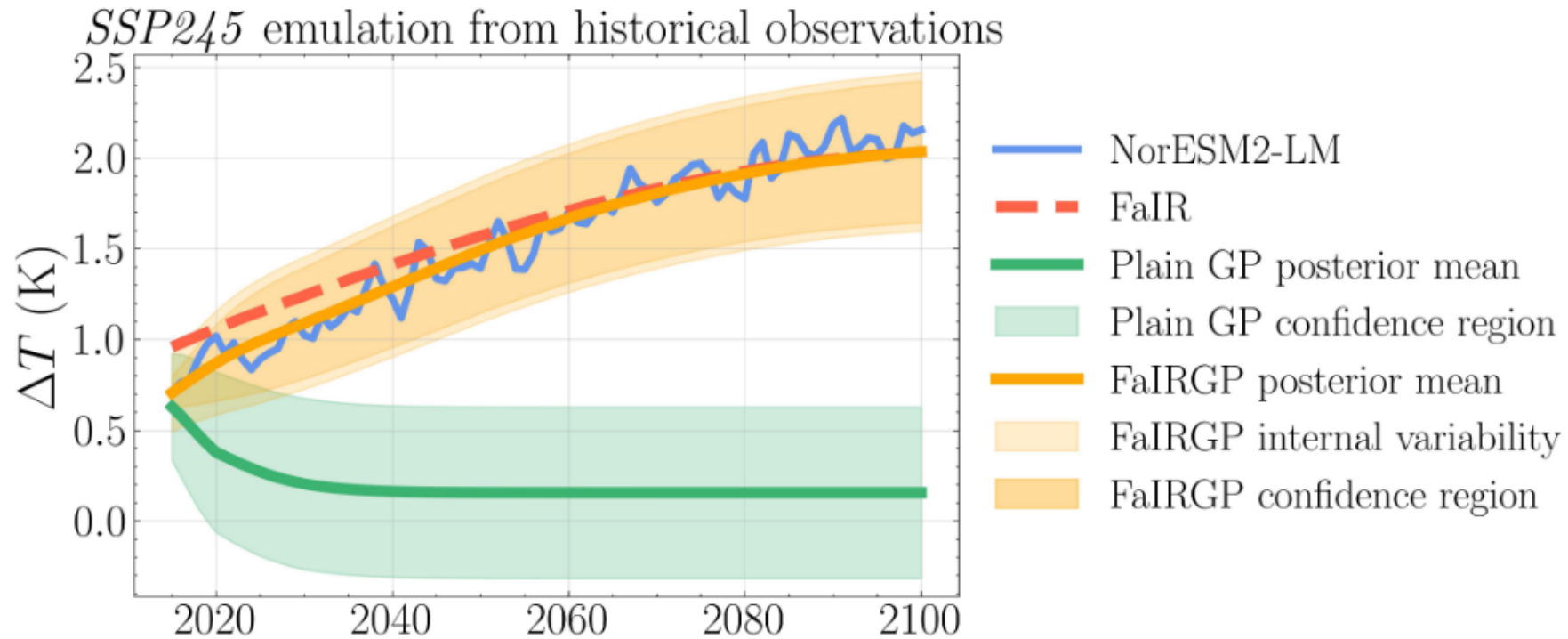


Figure 13: Emulated global mean surface temperature anomaly over 2015-2100 for the *SSP245* scenario with emulators trained over historical temperatures only.

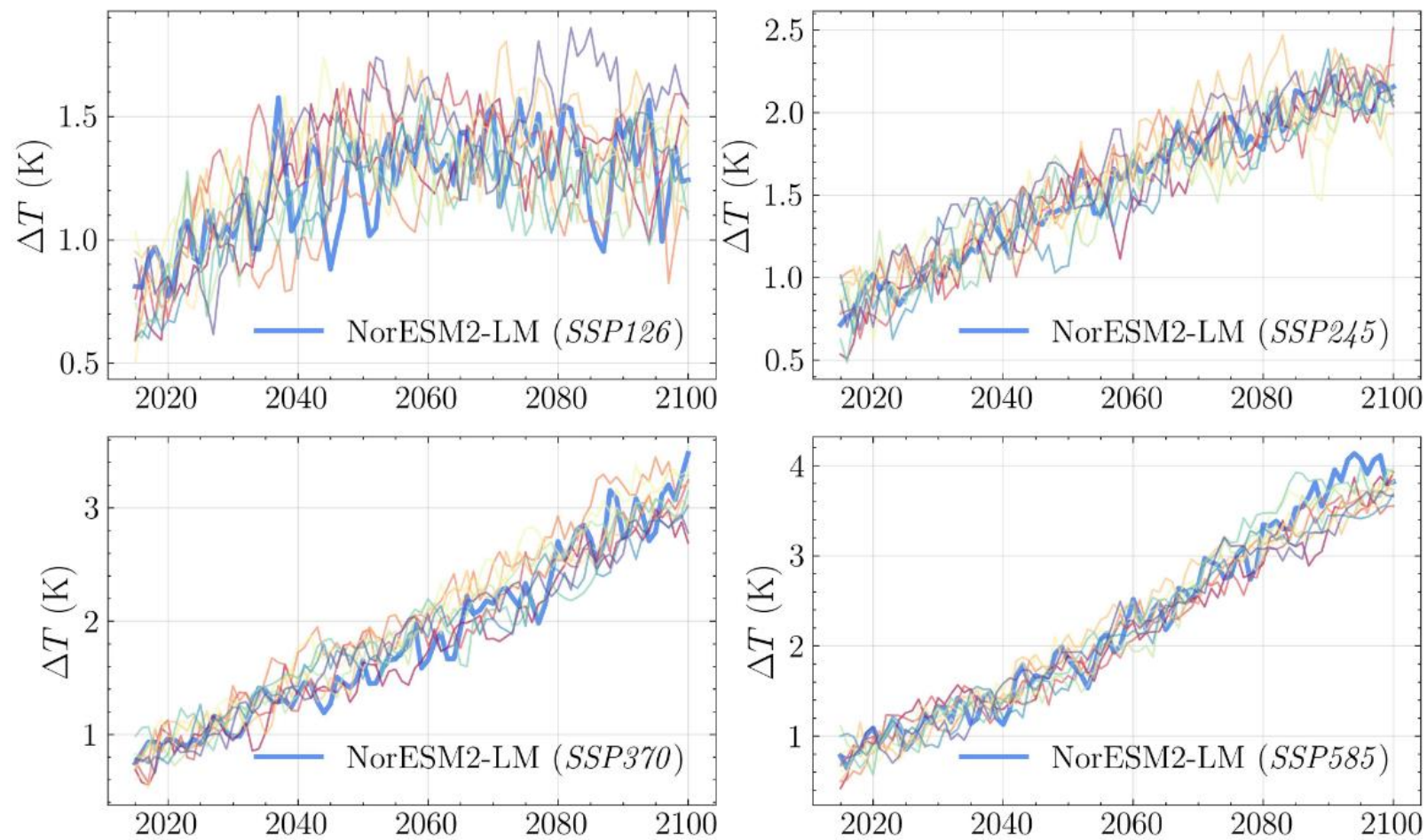


Figure 12: Example of sample paths from the FaIRGP posterior for the emulation of *SSP126* (top-left), *SSP245* (top-right), *SSP370* (bottom-left) and *SSP585* (bottom-right) global mean surface temperature anomaly over the 2015-2100 period.