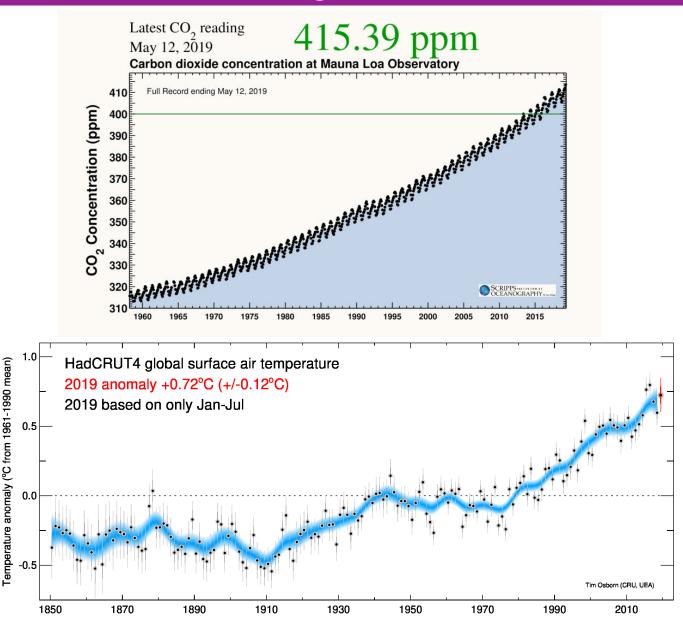
Applications of linear response theory in climate change research



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Climate Change: observations



Main questions?

How warm is it going to be in 2100?

Climate Sensitivity

Are the changes going to be 'smooth' or 'bumpy'?

Tipping Points

When is it too late to act to prevent dangerous climate change through emission reductions?

Safe Carbon Budget, Point of No Return

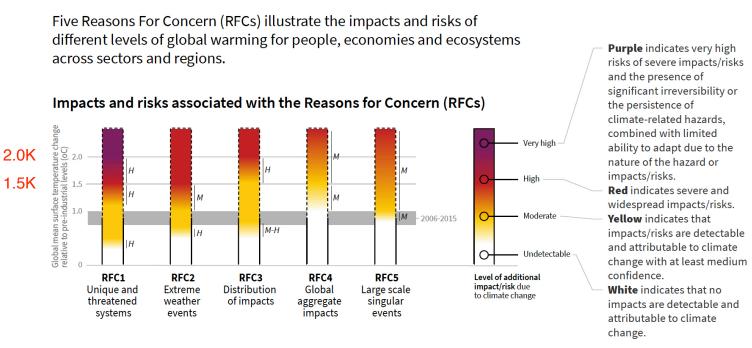
Are there any alternatives to avoid dangerous climate change?

Negative emissions, Geoengineering

Dangerous Climate Change?

IPCC SR1.5

How the level of global warming affects impacts and/or risks associated with the Reasons for Concern (RFCs) and selected natural, managed and human systems



IPCC, Special Report, October 2018

Tools

- Coordinated multi-model experiments with GCMs
- "CMIP5 provides a framework for coordinated climate change experimentation" (Taylor, Stouffer, and Meehl, 2012)
- More than 20 modeling groups
- More than 50 models
- Different design ideas, formulations, parameterizations
- Ensemble of "climate realizations"

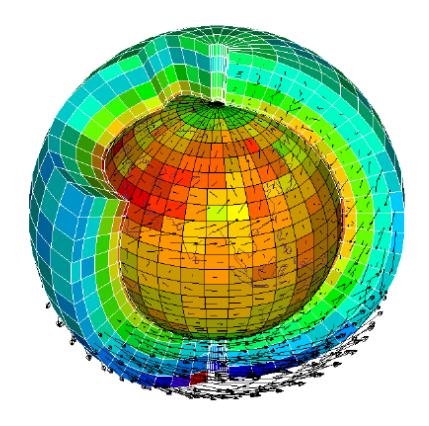
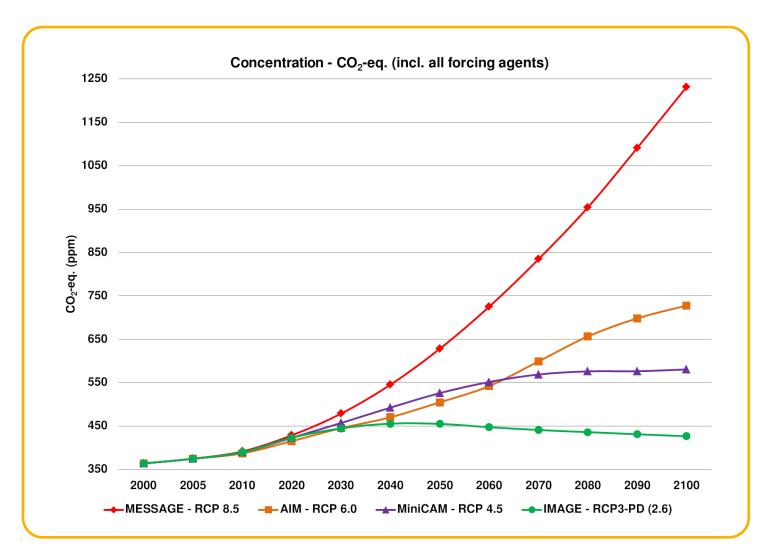


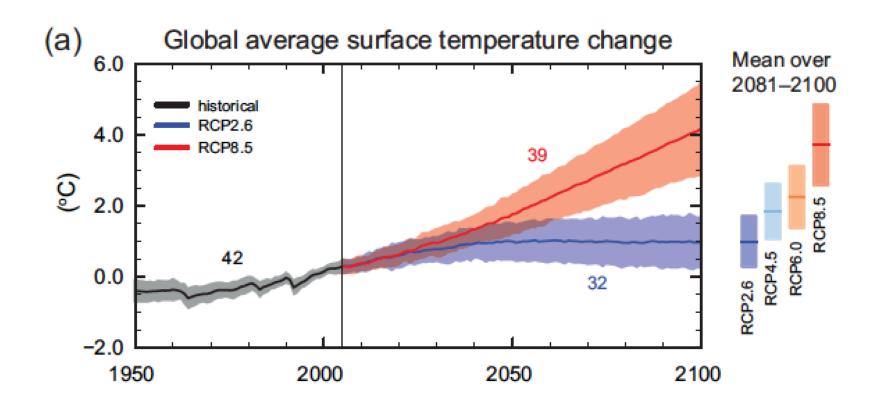
Figure: Global Climate Model (Thual, 2013)

Future Climate Change:

Representative concentration pathways



Projection results



How to avoid dangerous climate change?

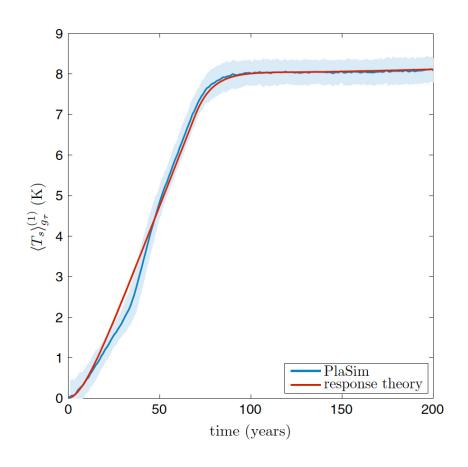
- Determining what happens under different scenario's

- Evaluate effects of action choices

Include uncertainty

Need for an efficient (stochastic) climate (incl. carbon cycle) model!

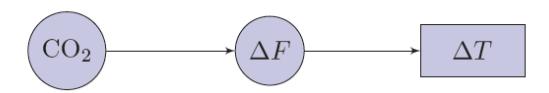
Linear Response Theory (LRT)



Ruelle (1998)
Ragone et al, (2016)
Lucarini et al. (2017)

Using LRT one can determine the response to any forcing!

Procedure



• Perturbation theory

$$\Delta T_{\Delta F}(t) = \Delta T_0 + \sum_{n=1}^{\infty} \Delta T_{\Delta F}^{(n)}(t)$$
 (1)

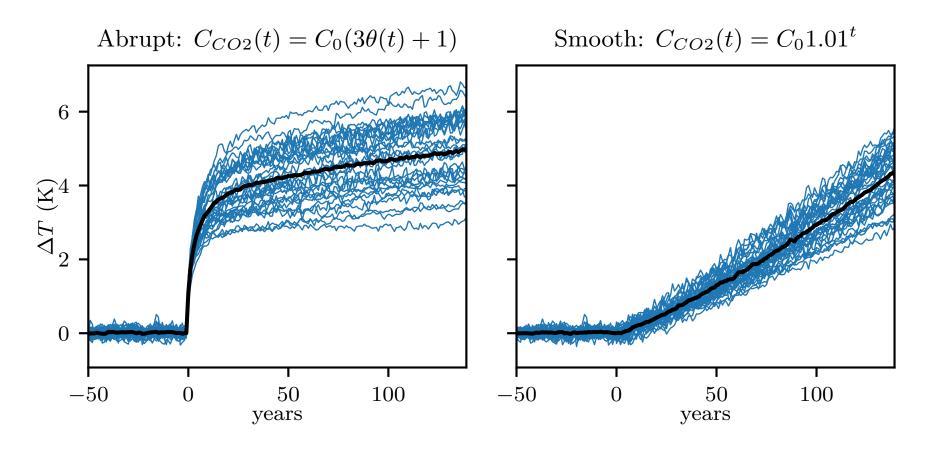
• Linear Response Theory: stop series at n=1

$$\Delta T_{\Delta F}^{(1)}(t) = \int_0^t G_T(t') \Delta F(t - t') dt'$$
 (2)

• Take a forcing-response pair $\Delta F_{abrupt}(t) = A\theta(t), \Delta T_{abrupt}(t)$

$$G_T(t) = \frac{1}{A} \frac{d}{dt} \Delta T_{abrupt} \tag{3}$$

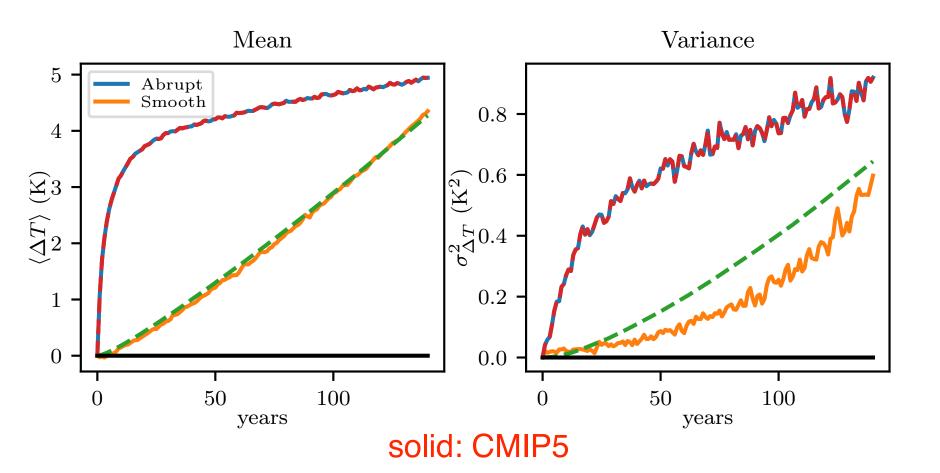
CMIP5 simulations



Abrupt forcing

Smooth forcing

Results Linear Response Theory

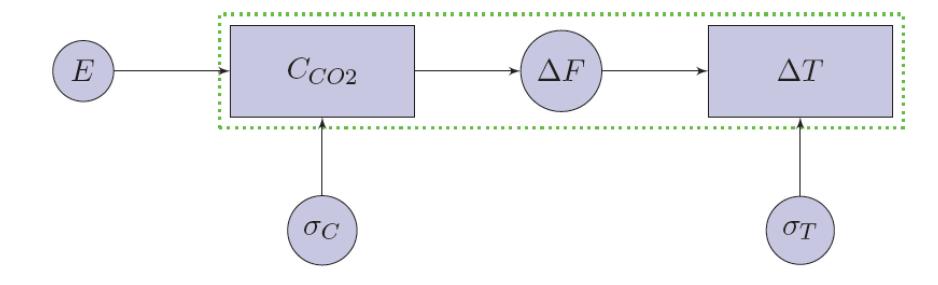


dashed: LRT

Aengenheyster et al. (2018)

From LRT to Stochastic State Space Model

- Up to now: determine ΔT from CO₂ concentrations
- Climate is forced by *fluxes*
- Relate CO₂ <u>emissions</u> to concentrations
- Risk, uncertainty: introduce stochasticity



Coupling a Carbon Model

• Carbon Model (Joos et al., 2013):

$$G_{CO2}(t) = a_0 + \sum_{i=1}^{3} a_i e^{-\frac{t}{\tau_i}}$$
(4)

• Full Reponse Function Model

$$C_{CO2}(t) = C_{CO2,0} + \int_0^t G_{CO2}(\tau) E_{CO2}(t - \tau) d\tau$$
 (5)

$$\Delta F_{CO2} = A \ \alpha_{CO2} \ln(C/C_0) \tag{6}$$

$$\Delta T(t) = \int_0^t G_T(\tau) \Delta F_{CO2}(t - \tau) d\tau \tag{7}$$

• We also find:

$$G_T(t) = \sum_{i=0}^{2} b_i e^{-t/\tau_{bi}}$$
 (8)

Stochastic State Space Model

Carbon

$$dC_P = a_0 E dt$$

$$dC_1 = \left(a_1 E - \frac{1}{\tau_1} C_1\right) dt$$

$$dC_2 = \left(a_2 E - \frac{1}{\tau_2} C_2\right) dt$$

$$+ \sigma_{C2} dW_t$$

$$dC_3 = \left(a_3 E - \frac{1}{\tau_3} C_3\right) dt$$

$$C_{CO2} = C_P + \sum_{i=1}^3 C_i$$

Temperature

$$\Delta F = A \alpha \ln(C_{CO2}/C_0)$$

$$d\Delta T_0 = \left(b_0 \Delta F - \frac{1}{\tau_{b0}} \Delta T_0\right) dt$$

$$+ \sigma_{T0} dW_t$$

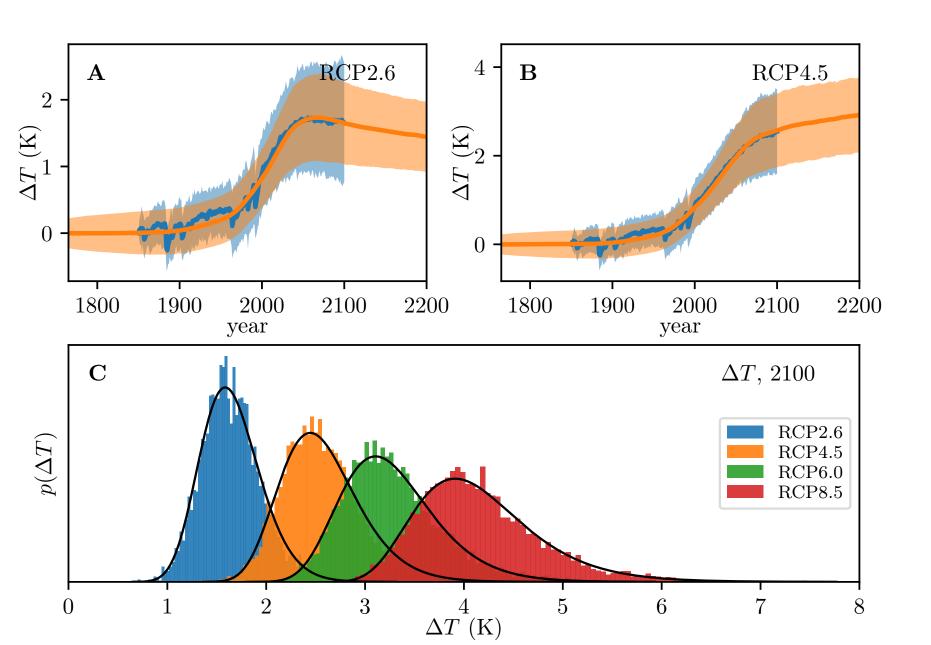
$$d\Delta T_1 = \left(b_1 \Delta F - \frac{1}{\tau_{b1}} \Delta T_1\right) dt$$

$$d\Delta T_2 = \left(b_2 \Delta F - \frac{1}{\tau_{b2}} \Delta T_2\right) dt$$

$$+ \sigma_{T2} \Delta T_2 dW_t$$

$$\Delta T = \sum_{i=0}^{2} \Delta T_i$$

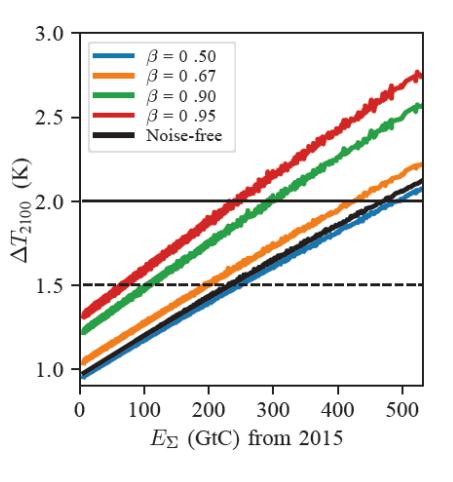
Results: Probability density functions



Save Carbon Budget

Maximum Cumulative Emissions that reach a certain warming target

$$p(\Delta T \le T_{max}) = \beta$$



$$T_{max} = \Delta T_{2100}$$

Table 3. Safe Carbon Budget (in GtC since 2015) as a function of threshold and safety probability β .

β	0.5	0.67 0.	.9 0.95	Noise-free
$T_{\text{max}} = 1.5 \text{ K}$	247	198 10	07 69	233
$T_{\text{max}} = 2 \text{ K}$	492	424 / 29	8 245	469

IPCC-AR5: 377 - 517 GtC to likely stay below 2 K Millar et al. (2017): 200 GtC to likely stay below 1.5 K

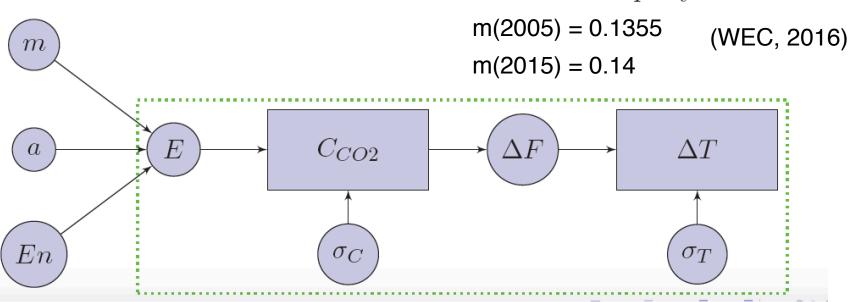
Economy & Transition pathways

$$Y = Y_0 e^{gt}$$

$$En = \gamma_0 e^{-r_{\gamma} t} Y$$

$$E = (1 - a)(1 - m)En$$

- Extreme Mitigation (EM): From time t_a on, we set m = 1, i.e. E = 0.
- 2 Fast Mitigation (FM): From time t_a on, both a, m increase by 0.05 per year.
- 3 Ambitious Mitigation (AM): As FM, but the increase is 0.02 per year.



Point of No Return (PONR)

- Use economical assumptions to determine emissions
 ⇒ baseline 'business-as-usual' scenario
- Control emissions by mitigation m(t) and abatement $a(t) \Rightarrow$ actions on climate change modify 'business-as-usual' scenario

Definition (Point of No Return)

The Point of No Return (PONR) is the time t_P from which on no <u>allowed</u> [a(t), m(t)] such that $0 \le a(t), m(t) \le 1, t_P \le t \le t_f$ can be chosen to fulfill

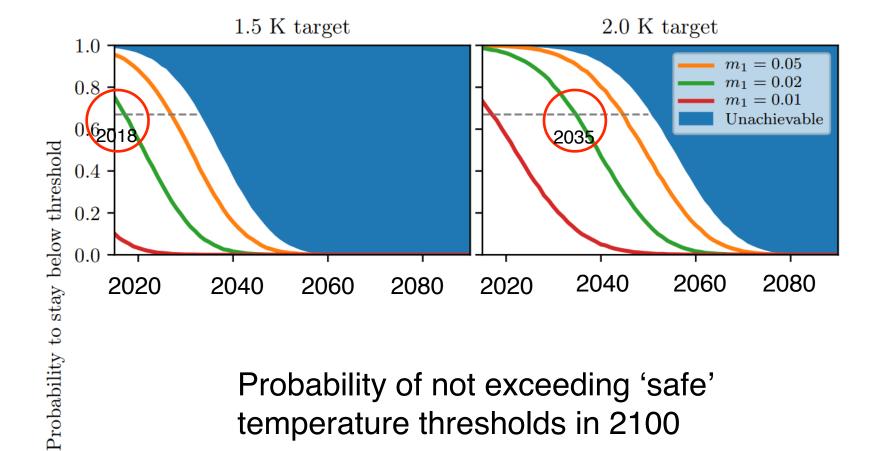
$$p(\Delta T(t_f) \le T_{max}) \ge \beta$$

Save Carbon Budget: We cannot reach target X by emitting more than Y

Point of No Return:

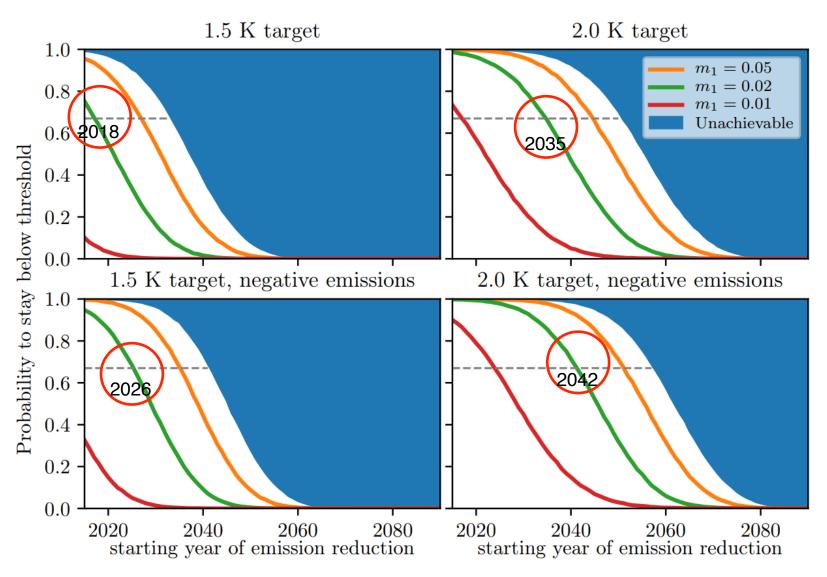
We cannot reach target X by starting reduction after year T

PONR: results



Probability of not exceeding 'safe' temperature thresholds in 2100

Effect of Negative Emissions



Probability of not exceeding 'safe' temperature thresholds in 2100

PONR: results

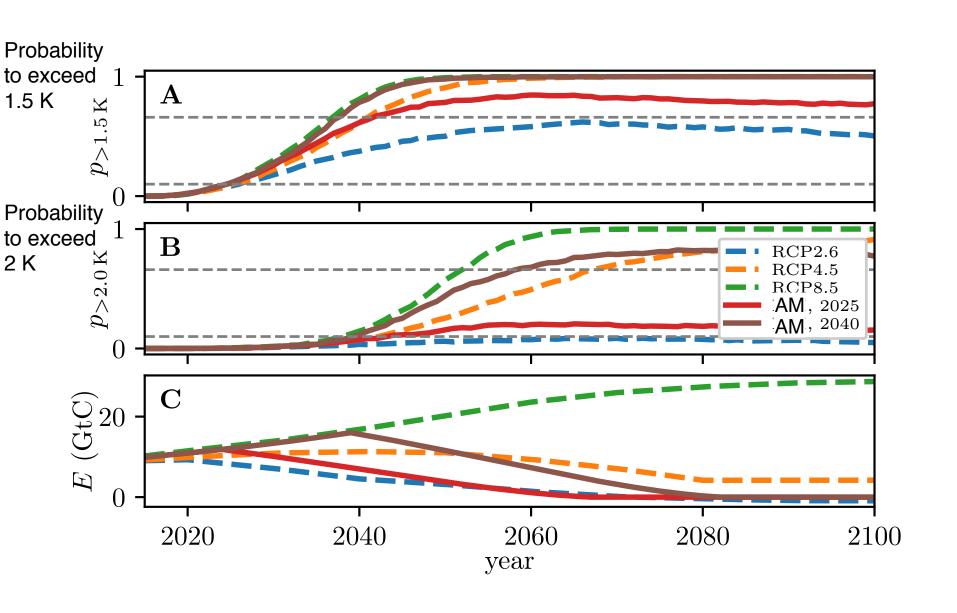
- ① Extreme Mitigation (EM): At time t_a , we set m = 1, so E = 0 from then onward.
- ② Fast Mitigation (FM): From time t_a onwards, both a, m increase by 0.05 per year.
- **3** Ambitious Mitigation (AM): As FM, but the increase is 0.02 per year.

Table 4. Point of no return as a function of threshold and safety probability β without and with strong negative emissions.

β		0.5		0.67		0.9		0.95		noise-free	
	E_{neg}	none	strong								
EM	$T_{\text{max}} = 1.5 \text{ K}$ $T_{\text{max}} = 2 \text{ K}$	2038 2056	2046 2062	2034 2051	2042 2058	2026 2042	2035 2049	2022 2038	2032 2046	2037 2055	2045 2061
FM	$T_{\text{max}} = 1.5 \text{ K}$ $T_{\text{max}} = 2 \text{ K}$	2032 2050	2039 2056	2027 2045	2036 2052	2020 2036	2028 2043	2016 2032	2025 2039	2030 2048	2038 2055
AM	$T_{\text{max}} = 1.5 \text{ K}$ $T_{\text{max}} = 2 \text{ K}$	2022 2040	2029 2046	2018 2035	2026 2042	2026	2019 2033	2022	2030	2021 2038	2029 2045

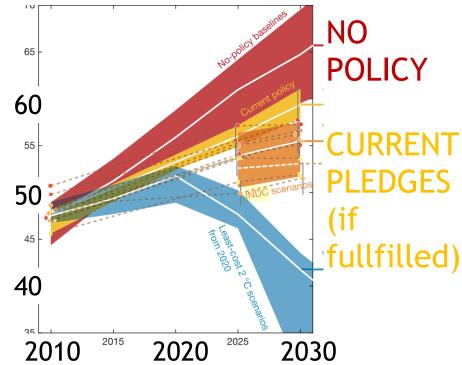
The PNR for 1.5 K has been passed!

Scenarios



Current Climate Policy

annual GHG emissions (Gt(CO2 eq) / year)



Rogelj et al. 2016

NEEDED TO REACH EVEN 2°C keep their intended contributions, we are NOT on the right path to reach the Paris agreement!

Summary

Shown value of Linear Response Theory in the CMIP5 context

- Build an stochastic state space model to adequately determine the climate response to emissions.
- Easy to communicate metrics:
 - SCB: "We cannot reach target X when emitting more than \$
 - PONR: "We cannot reach target X when starting after year PONR"

PONR with realistic action pathways is close (2035 for 67%) for the 2K target and already passed for the 1.5K target.

Further reading

M. Aengenheyster, Q. Feng, F. van der Ploeg and H. A. Dijkstra. The point of no return for climate action: effects of climate uncertainty and risk tolerance Earth System Dynamics, 9, 1085-1095, (2018)

K. Helwegen, C. Wieners, J. Frank and H. A. Dijkstra. Complementing CO₂ emission reduction by solar radiation management might strongly enhance future welfare, Earth System Dynamics, 10, 453-472, (2019).