

Mechanisms affecting equilibrium climate sensitivity in the PlaSim Earth System Model with different ocean model configurations



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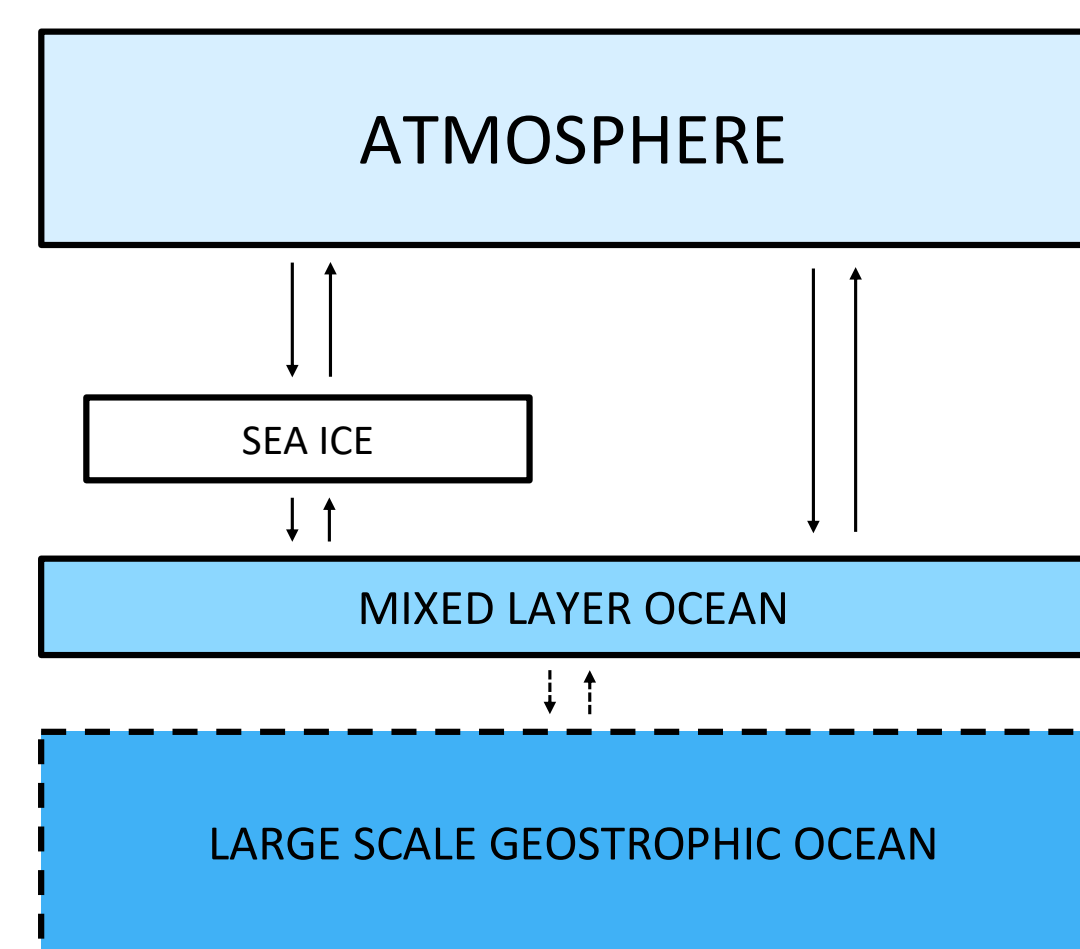
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The equilibrium climate sensitivity (ECS) of a state-of-the-art Earth System Model of intermediate complexity, the **Planet Simulator (PlaSim)**, is determined under three tuned configurations, in which the model is coupled with a simple **Mixed Layer (ML)** or with the full 3D **Large Scale Geostrophic (LSG)** ocean model, at two horizontal resolutions, **T21 (600 km)** and **T42 (300 km)**. Equilibrium climate sensitivity experiments with doubled and quadrupled CO₂ were run, using either dynamic or prescribed sea ice. The resulting ECS using dynamic sea ice is **6.2 K** for PlaSim-ML T21, **5.5 K** for PlaSim-ML T42 and a much smaller **4.2 K** for PlaSim-LSG T21. A systematic comparison between simulations with dynamic and prescribed sea ice helps to identify a strong contribution of sea ice to the value of the feedback parameter and of the climate sensitivity. Additionally, Antarctic sea ice is underestimated in PlaSim-LSG leading to a further reduction of ECS when the LSG ocean is used. The ECS of ML experiments is generally large compared with current estimates of equilibrium climate sensitivity in CMIP5 models and other EMICs: a relevant observation is that the choice of the ML horizontal diffusion coefficient, and therefore of the parameterized meridional heat transport and in turn the resulting equator-poles temperature gradient, plays an important role in controlling the ECS of the PlaSim-ML configurations. This observation should be possibly taken into account when evaluating ECS estimates in models with a mixed layer ocean. The configuration of PlaSim with the LSG ocean shows very different **AMOC regimes**, including 250-year oscillations and a complete shutdown of meridional transport, which depend on the ocean vertical diffusion profile and the CO₂ forcing conditions.

(1) The Planet Simulator model

PlaSim is an Earth system Model of Intermediate Complexity developed at the University of Hamburg (*Fraedrich et al., 2005*).

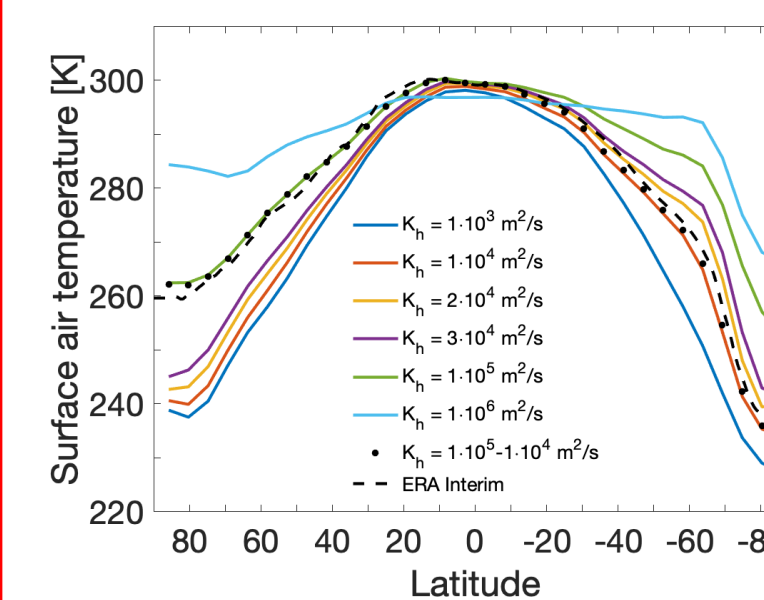
- The dynamical core is a simplified General Circulation Model, the **Portable University Model of Atmosphere (PUMA)**
- The oceanic component can be prescribed by sea surface temperature climatology, introduced by a simple **Mixed Layer (ML) ocean model** or coupled as a dynamical model such as the **Large Scale Geostrophic (LSG) Ocean Circulation Model** (*Maier-Reimer et al., 1993*)
- Sea ice distribution can either be prescribed by climatology or simulated by a **thermodynamic sea ice model**



(2) Tuning

T21 + ML

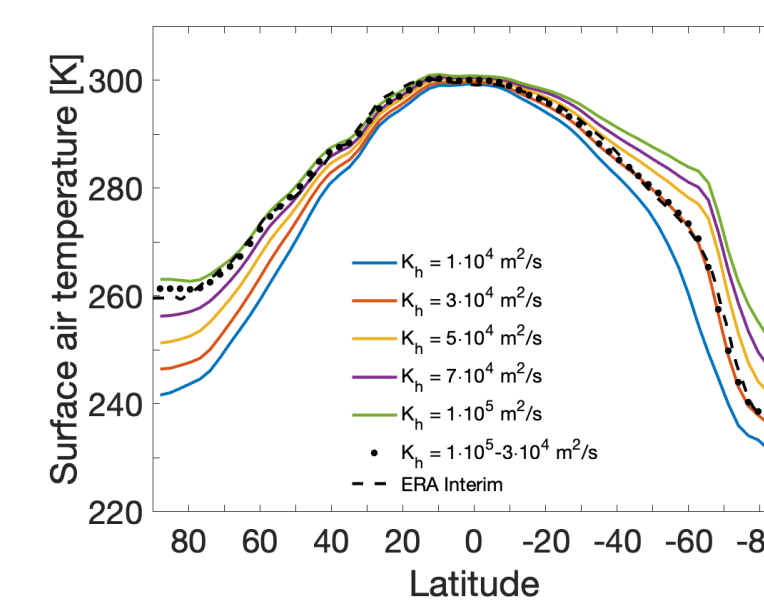
Mixed Layer ocean
T21 (~ 600 km) resolution
Tuned parameter:
horizontal oceanic diffusion coefficient K_h



$K_h = 1 \cdot 10^5 \text{ m}^2/\text{s}$ (NH)
 $K_h = 1 \cdot 10^4 \text{ m}^2/\text{s}$ (SH)

T42 + ML

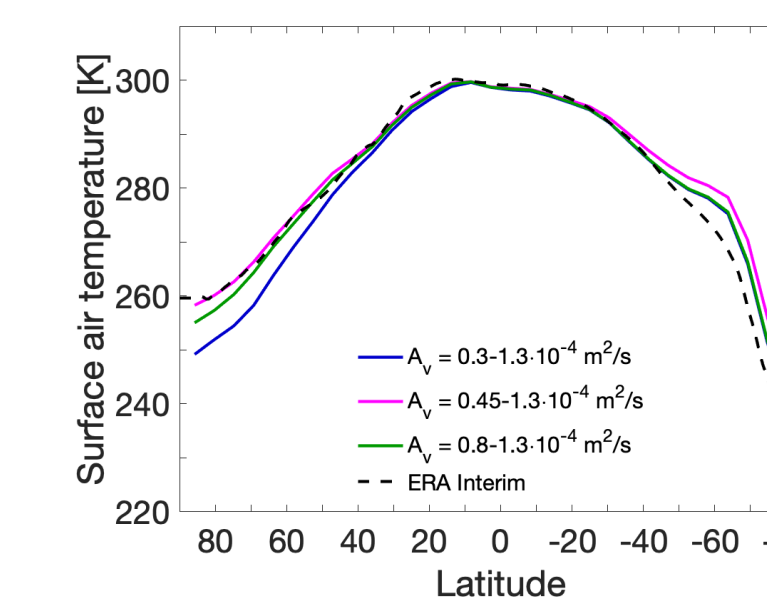
Mixed Layer ocean
T21 (~ 300 km) resolution
Tuned parameter:
horizontal oceanic diffusion coefficient K_h



$K_h = 1 \cdot 10^5 \text{ m}^2/\text{s}$ (NH)
 $K_h = 3 \cdot 10^4 \text{ m}^2/\text{s}$ (SH)

T21 + LSG

Large Scale Geostrophic ocean
T21 (~ 600 km) resolution
Tuned parameter:
vertical oceanic diffusion coefficient A_v

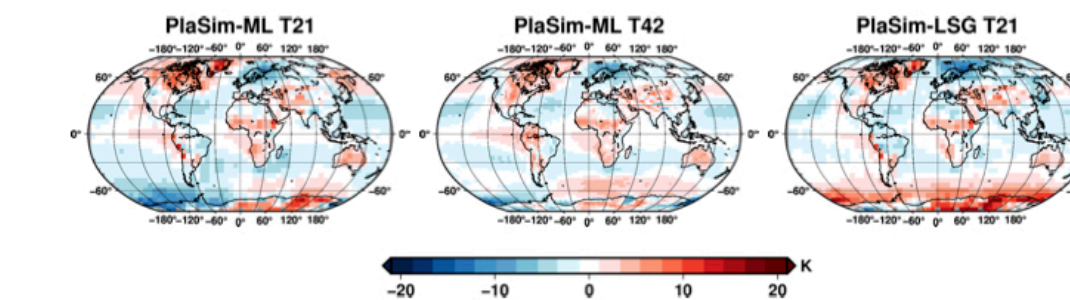


A_v from $0.45 \cdot 10^{-4} \text{ m}^2/\text{s}$ (top) to
 $1.3 \cdot 10^{-4} \text{ m}^2/\text{s}$ (bottom)
to reduce the Sou. Ocean bias

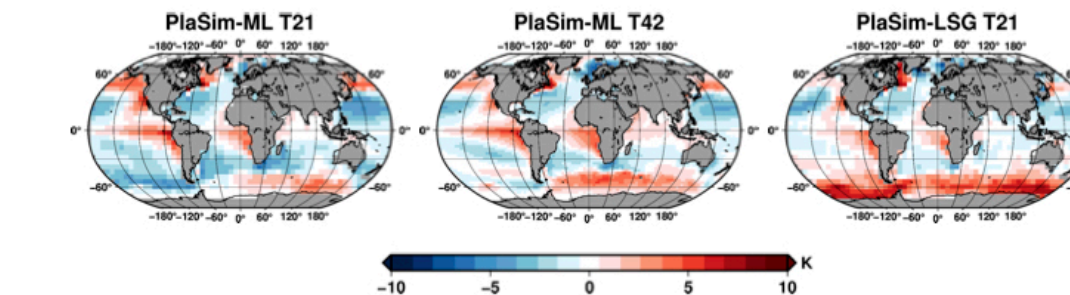
- ✓ perennial runs (CO₂ = 354 ppm)
- ✓ dynamic sea ice

(3) Simulated climate and energy balance

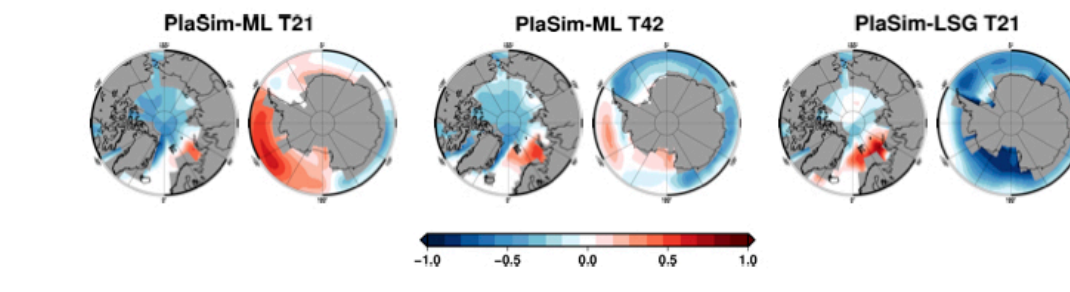
Surface air temperature anomaly [K]
with respect to the ERA Interim
reanalysis dataset (2005-2015 average)
(*Dee et al., 2011*)



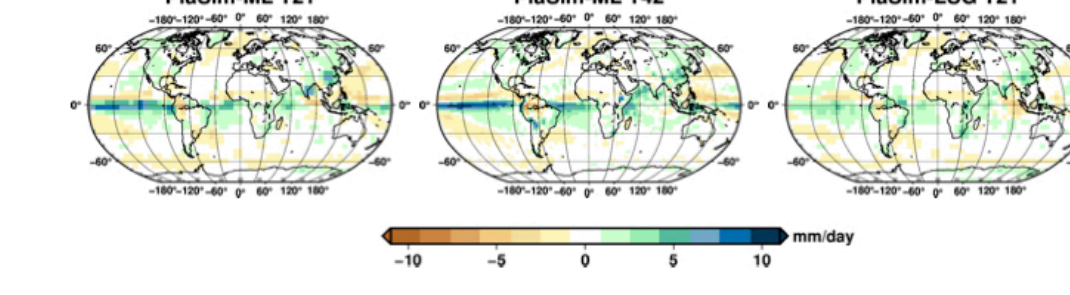
Sea surface temperature anomaly [K]
with respect to the HadISST dataset
(2005-2015 average)
(*Titchner & Rayner, 2014*)



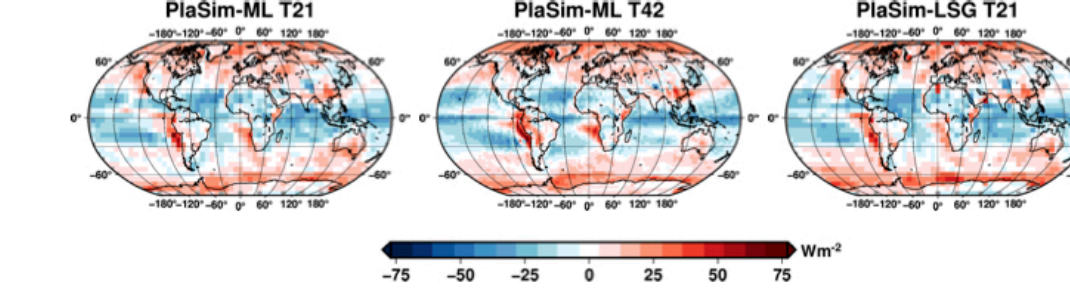
Sea ice cover anomaly
with respect to the HadISST dataset
(2005-2015 average)
(*Titchner & Rayner, 2014*)



Precipitation anomaly [mm/day]
with respect to the GPCP dataset
(2005-2015 average)
(*Adler et al., 2003*)



TOA energy budget anomaly [W/m²]
with respect to the CERES-EBAF dataset
(2005-2015 average)
(*Loeb et al., 2018*)



Simulated and observed
energy fluxes (Wm⁻²) at the
top of the atmosphere
and at the surface

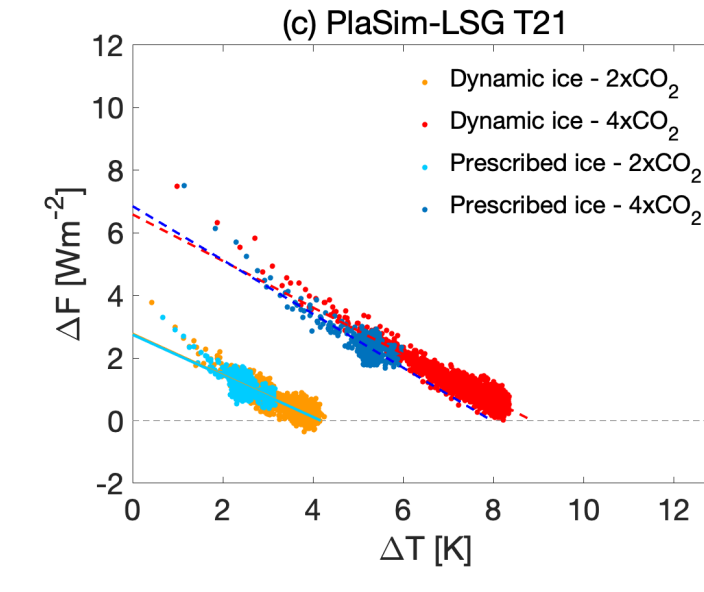
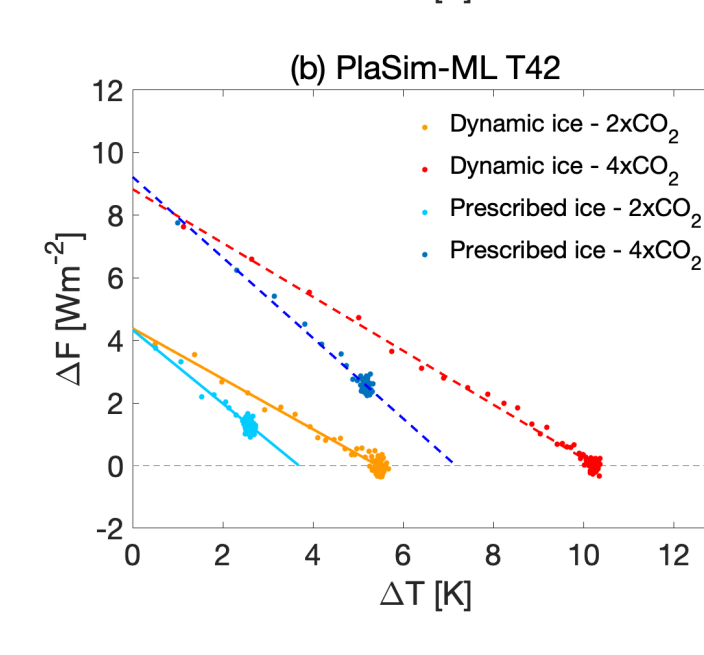
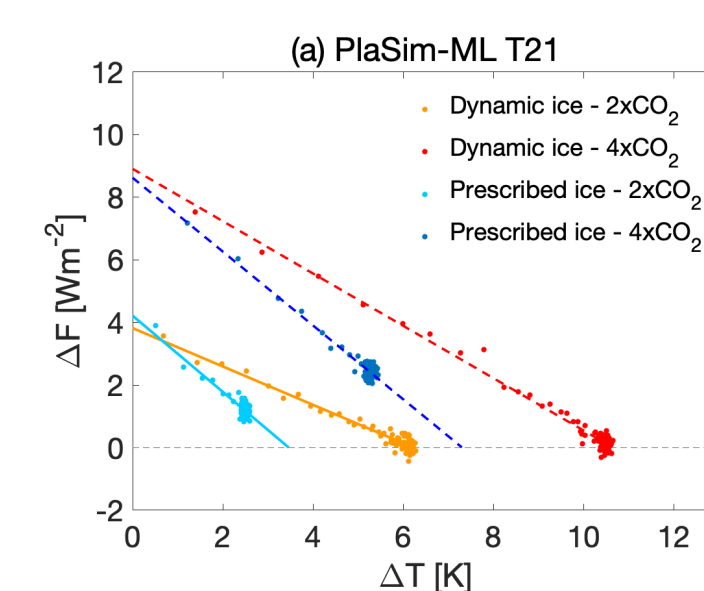
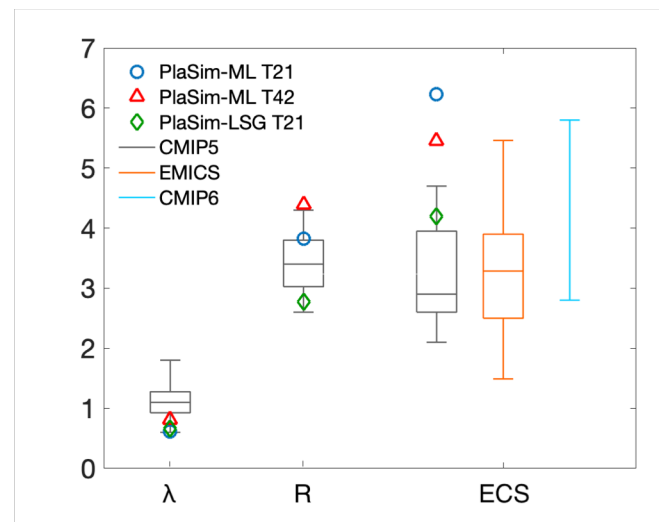
	PlaSim-ML T21	PlaSim-ML T42	PlaSim-LSG T21	Stephens et al. (2012)
TOA net longwave	231.5	235.8	232.8	240.2
TOA net shortwave	-232.3	-236.0	-232.9	-239.7
TOA energy budget	-0.8	-0.1	-0.1	0.6
Surface net longwave	163.2	169.4	164.1	165
Surface net shortwave	-62.8	-62.4	-63.0	-52.4
Surface energy budget	-18.0	-20.8	-18.3	-24
Latent heat flux	42.0	46.5	42.7	48
Surface energy budget	-0.5	-0.2	0.1	0.6
TOA-surface net	-0.2	0.1	-0.2	0

Please notice that the estimates from *Stephens et al. (2012)* are only reported for reference, since we are comparing **equilibrium model results** with the current observed transient.

(4) Equilibrium climate sensitivity and feedback parameter

EQUILIBRIUM CLIMATE SENSITIVITY (ECS)

Equilibrium (steady state) change in the annual global mean surface temperature following a **doubling** of the atmospheric equivalent carbon dioxide (CO₂) concentration (*IPCC*)



Feedback parameter λ

$$\Delta F = R - \lambda \Delta T$$

ΔF = net TOA radiative flux change

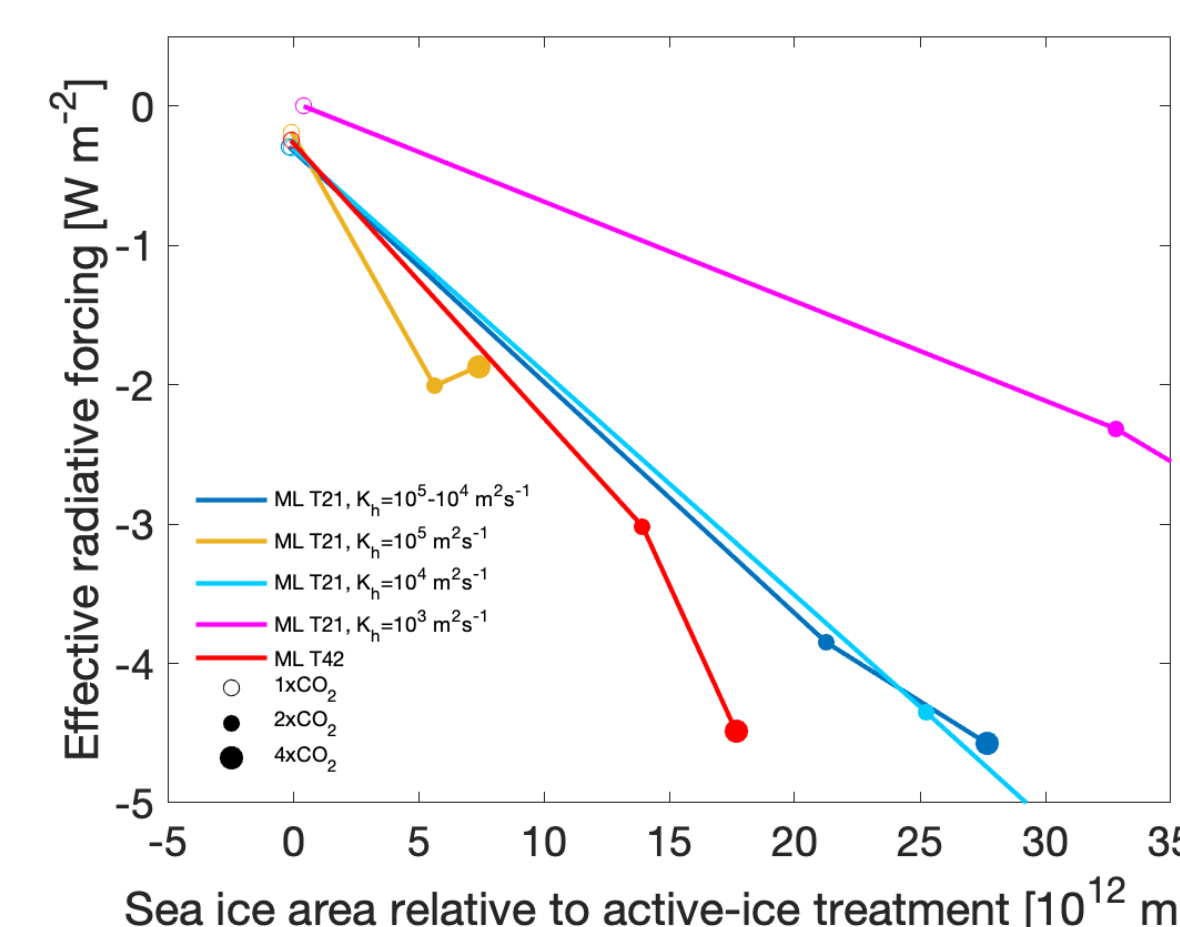
R = radiative forcing

ΔT = surface air temperature change

ECS and λ calculated as in *Gregory et al. (2004)*

- ✓ CO₂ doubling experiments (285 ppm → 570 ppm)
- ✓ CO₂ quadrupling experiments (285 ppm → 1140 ppm)
- ✓ dynamic or prescribed sea ice

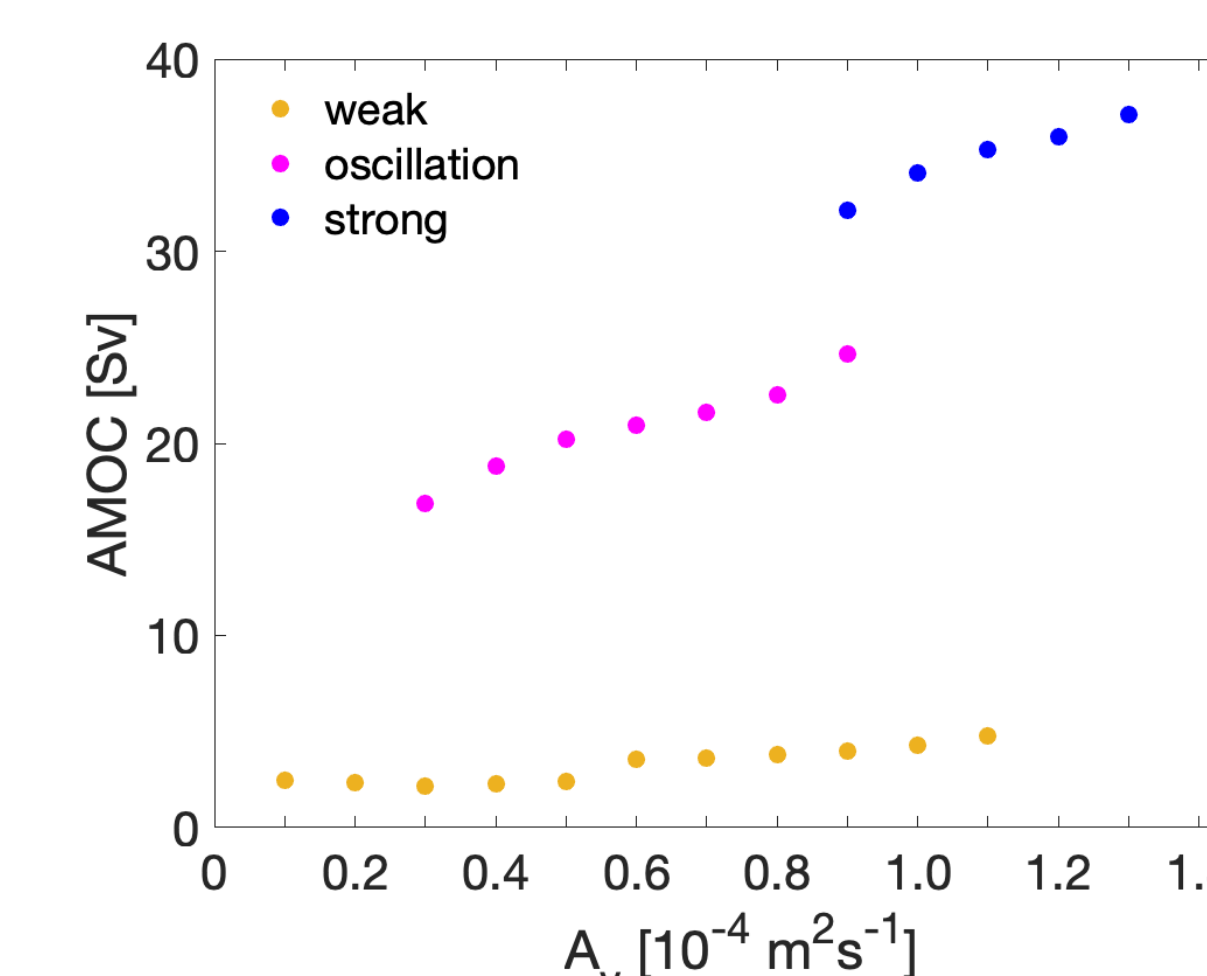
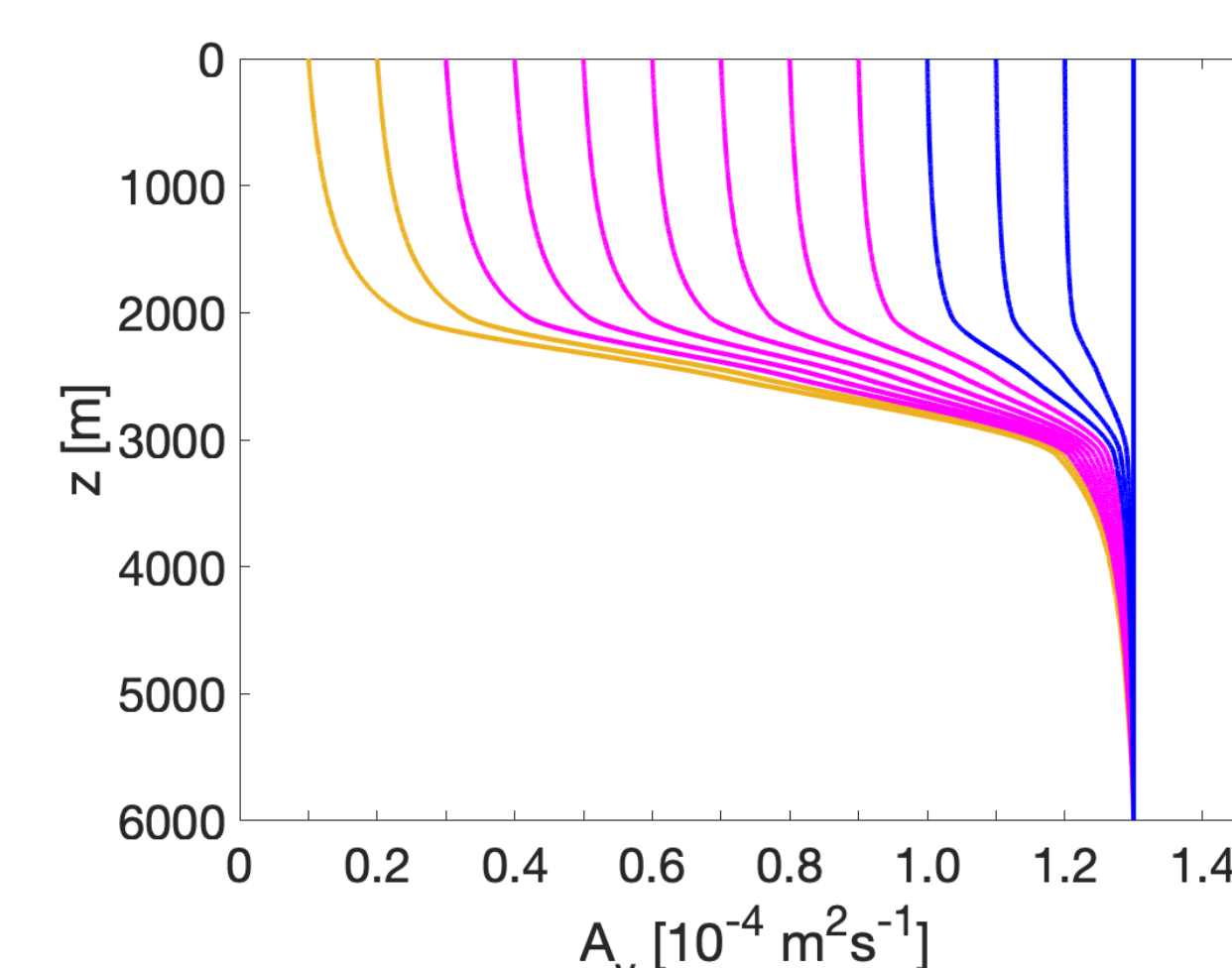
Approach of
Caldeira & Cvijanovic (2014)



Increasing the horizontal diffusion coefficient of the Mixed Layer (K_h) the oceanic temperatures become globally more homogeneous. As a consequence the average outgoing longwave radiation (proportional to T^4) strongly decreases and the model tends to increase global average temperature in order to keep the energy balance. The high PlaSim-ML ECS depends on the choice of K_h during the model tuning.

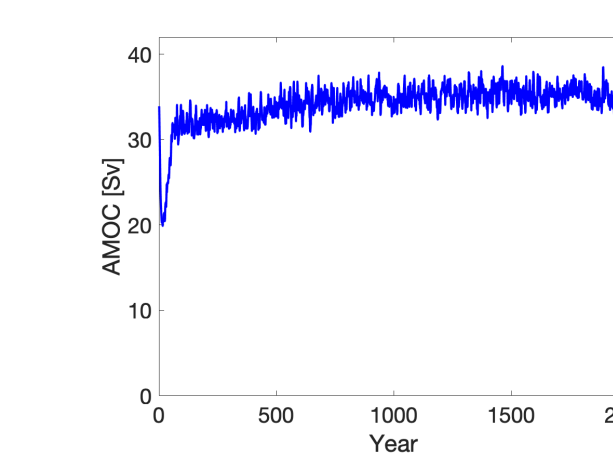
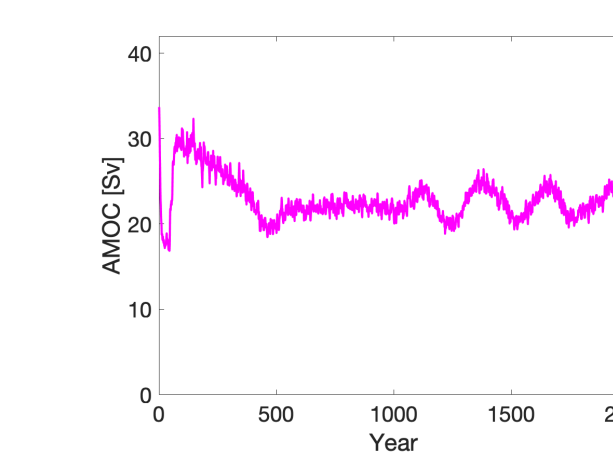
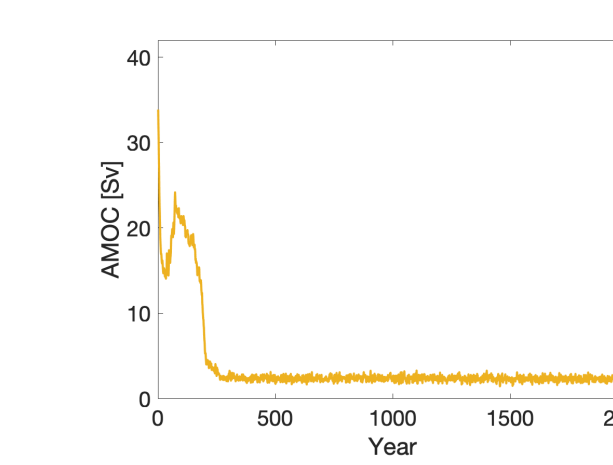
(5) Atlantic Meridional Overturning Circulation regimes

Tuned parameter:
vertical oceanic diffusion coefficient A_v
from $0.1 \cdot 10^{-4} \text{ m}^2/\text{s}$ to $1.3 \cdot 10^{-4} \text{ m}^2/\text{s}$ (top)
 $1.3 \cdot 10^{-4} \text{ m}^2/\text{s}$ (bottom)



- ✓ perennial runs (CO₂ = 285 ppm)
- ✓ dynamic sea ice
- ✓ **T21 + LSG**

Example of weak, oscillating and strong regimes



(6) Conclusions

The tuning of PlaSim with the ML ocean suggests the use of **two different values** for the oceanic horizontal diffusion coefficient. The tuning of the coupled PlaSim-LSG model recommends an oceanic vertical diffusion ranging from $0.45 \cdot 10^{-4} \text{ m}^2/\text{s}$ to $1.3 \cdot 10^{-4} \text{ m}^2/\text{s}$. The climatic variables are well simulated but there is a warm bias in the Southern Ocean using PlaSim-LSG. The ECS is estimated from CO₂ doubling and quadrupling experiments. The large ECS of PlaSim-ML depends on the choice of the horizontal diffusion coefficient, that is on the intensity of the meridional heat transport. The PlaSim-LSG configuration shows different AMOC regimes as a function of the vertical diffusion coefficient: in addition to two steady states, characterized by **weak** and **strong** AMOC, an intermediate regime with **oscillations** is also possible.

(7) Bibliography

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