Impact of atmospheric small-scale fluctuations on climate sensitivity

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Received 31 January 2008; revised 7 April 2008; accepted 9 April 2008; published 21 May 2008.

[1] Climate change scenarios are based on numerical models with finite spatial and temporal resolutions. The impact of unresolved processes is parameterized without taking the variability induced by subscale processes into account. This drawback could lead to an over-/underestimation of the climate sensitivity. The aim of this study is to investigate the impact of small-scale atmospheric fluctuations on the modeled climate sensitivity to increased CO₂ concentration. Using a complex coupled atmosphere-ocean general circulation model (ECHAM5/MPI-OM) climate response experiments with enhanced small-scale fluctuations are performed. Our results show that the strength of the global warming due to a CO₂ doubling depends on the representation of small-scale fluctuations. Reducing the horizontal diffusion by a factor of 3 leads to an increase of the equilibrium climate sensitivity by 13%. If white noise is added to the small scales, the climate sensitivity tends to weaken. The largest changes in responses occur in the upper troposphere. Citation: Seiffert, R., and J.-S. von Storch (2008), Impact of atmospheric small-scale fluctuations on climate sensitivity, Geophys. Res. Lett., 35, L10704, doi:10.1029/ 2008GL033483.

1. Introduction

[2] Estimating climate response to increasing greenhouse gas concentrations is one of the main challenges in numerical climate modelling. However, different climate models show different climate responses although they were forced in the same way. The projected globally averaged surface warming at the end of the 21st century ranges from 1.7 K to 4.4 K in the A1B Scenario of 4th Assessment Report published by the Intergovernmental Panel on Climate Change (IPCC) [2007]. This transient climate response is closely related to the equilibrium climate sensitivity, which is defined as the equilibrium change in global surface temperature due to a doubling of CO₂ concentration. This sensitivity is highly uncertain and often estimated by a probability density function [e.g., Murphy et al., 2004; Knutti et al., 2006]. Several nonlinear thermodynamical feedback processes are generally called to account for the great uncertainty of the climate sensitivity. These feedbacks involve interactions of water vapor, clouds, temperature lapse rate and surface albedo with the earth's radiation budget [Bony et al., 2006]. Several studies [e.g., Colman, 2003; Soden and Held, 2006; Webb et al., 2006] have estimated the strength of the feedback processes in different

climate models and provided valuable insight into the origins of varying model sensitivities.

[3] Nevertheless, the climate system is not only determined by the interactions of complex thermodynamical processes but also by nonlinear dynamical coupling of various scales of motion. As a consequence of the limited model resolution the governing equations are truncated at a certain scale. Dynamical processes below this scale cannot be adequately represented in the models and have to be parameterized. Such parameterizations generally do not take the variability induced by the subgrid-scale processes into account, even though variations near the truncation scale depend on the model resolution [von Storch, 2004; Seiffert et al., 2006]. Due to the dynamical coupling between small and large scales, small-scale variability can affect the statistics of large-scale variables [von Storch, 2004]. According to the fluctuation dissipation theorem [Leith, 1975], such statistics are crucial for the responses to a changed external forcing (e.g., CO₂ increase). Neglecting small-scale variability could have an influence on the model sensitivity to CO₂ forcing.

[4] One approach to investigate whether neglecting variability on scales smaller than the truncation scale affects model sensitivities is to increase the model resolution. Kiehl et al. [2006] found that the climate sensitivity of the Community Climate System Model 3 (CCSM3) increases with increased horizontal resolution. The equilibrium climate sensitivity of the high-resolution version (T85) was 17% higher than the model sensitivity of the low-resolution version (T31). In contrast, the atmospheric general circulation model ECHAM5 of the Max Planck Institute (MPI) for Meteorology coupled to a fixed-depth (50m) mixed layer ocean and thermodynamical sea ice module experiences a reversed trend (E. Roeckner, personal communication, 2007). The results of May and Roeckner [2001], who were using the predecessor of the MPI model, ECHAM4, also suggest a weaker climate response to increased CO₂ concentration when changing the resolution from T42 to T106. The opposite trends in the two models, CCSM3 and ECHAM, can be ascribed to many factors. The different parameterizations in the models certainly play an important role. Since parameterizations in general depend on the chosen model resolution, it is hard to distinguish between the effect of changed parameterizations due to different resolution and the effect of additional resolved small-scale processes on the climate sensitivity.

[5] The purpose of this paper is to systematically address the question of whether and how small-scale atmospheric fluctuations affect the modeled climate sensitivity to increased CO_2 concentration. Using the coupled atmosphereocean general circulation model ECHAM5/MPI-OM we carried out idealized response experiments with enhanced small-scale fluctuations. Two different approaches are used to modify the small-scale fluctuations. These approaches

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Table 1. Overview of the Experiments Carried out WithECHAM5/MPI-OM

	$1 \times CO_2$	$2 \times CO_2$
Control experiments	ctrl1x	ctrl2x
Reduced diffusion $\tau_0 = 24$ h	diffus1x_24	diffus2x_24
Reduced diffusion $\tau_0 = 36$ h	diffus1x_36	diffus2x_36
Moderate noise $\sigma_{noise} = 3 \cdot 10^{-2} K$	noise1x_3	noise2x_3
$(3 \cdot 10^{-7} s^{-1})$		_
High noise $\sigma_{noise} = 6 \cdot 10^{-2} K$	noise1x_6	noise2x_6
$(6 \cdot 10^{-7} s^{-1})$	_	_

and the experimental design are described in section 2. Results are presented in section 3. Discussions and conclusions are given in the final section.

2. Method

2.1. Model

[6] ECHAM5/MPI-OM is a coupled atmosphere-oceansea ice general circulation model developed at the MPI for Meteorology in Hamburg. The model (with different resolutions) has been successfully applied in several research projects [e.g., *Jungclaus et al.*, 2006; *Roeckner et al.*, 2006; *IPCC*, 2007]. Details of the different model components are given by *Roeckner et al.* [2003] and *Marsland et al.* [2003]. For this study we used the low-resolution version (atmosphere: T31 (\approx 3.8° by 3.8°) and 19 vertical levels, ocean: \approx 3° by 3° and 40 vertical levels).

2.2. Two Approaches to Modify the Small-Scale Fluctuations

[7] The first method to increase the small-scale fluctuations involves a modification of the horizontal diffusion parameterization scheme of ECHAM5. The horizontal diffusion damps to a large extent the variability of small-scale components and can therefore be used to enhance smallscale variability. This parameterization is applied in spectral space on the prognostic variables: temperature, vorticity and divergence. The horizontal diffusion is mainly used to ensure a realistic energy spectrum of the resolved scales. The time rate of change of the spectral coefficient $X_{l,m}$ caused by the horizontal diffusion is defined as

$$\frac{\partial X_{l,m}}{\partial t}\Big|_{\text{horizontal diffusion}} = -K_l X_{l,m} \tag{1}$$

with

$$K_l = \frac{1}{\tau_0} \left(\frac{l(l+1)}{l_0(l_0+1)} \right)^q.$$
 (2)

[8] The damping factor K_l depends strongly on the total wavenumber *l*. l_0 marks the truncation scale of the model (here: $l_0 = 31$). The exponent *q* depends on the vertical level ranging from 2 in the uppermost three levels to 10 in the middle and lower troposphere. The damping time scale τ_0 controls the strength of the diffusion. In the standard low-resolution version it is set to $\tau_0 = 12$ h. An increase in τ_0 leads to a weaker damping and hence an enhancement of primarily the small-scale variability.

[9] In the second approach the horizontal diffusion parameterization scheme remains unchanged. Instead we add noise to the smallest resolved scales. At each time step white noise is added to the spectral coefficients of temperature, divergence and vorticity with a total wavenumber ≥ 26 . Note that the explicit representation of the small-scale processes related to wavenumbers close to the truncation scale ($l_0 = 31$) is not reliable anyway. The noise mimics a possible impact of unresolved processes. However, it does not qualify for a realistic parameterization of the subgrid-scale variability. Rather, it aims to isolate the impact of enhanced small-scale fluctuations on the climate sensitivity in the framework of idealized experiments. We do not try to improve the representation of subgrid-scale processes by stochastic parameterization.

2.3. Experiments

[10] The experiments are carried out in pairs. Each set of experiments comprises one integration done with preindustrial CO_2 concentration (280 ppm) and one integration with doubled CO_2 concentration. The difference of these two integrations can then be referred to as the response of the system to the increased CO_2 concentration. The 2×CO₂ experiment was originally started from a state of the preindustrial integration. The doubled CO₂ concentration was achieved by a 1% per year increase until the final value of 560 ppm was reached. Holding the concentration constant, the model was further integrated for 880 years. The preindustrial integration has a total length of 1500 years. From these long integrations only the last 50 years were used for the analysis. Table 1 gives an overview of the experiments. Experiments ctrl1x and ctrl2x correspond to the control integrations of the standard model with no changes of the representation of the small-scale fluctuations. '1x' and '2x' denote $1 \times CO_2$ and $2 \times CO_2$. The other experiments are 150 years continuations of ctrl1x and ctrl2x in which the model was altered in the way described below. The first 100 years of the continuations are excluded from the analysis. All model runs are sufficiently equilibrated.

[11] In experiments diffus1x_24, diffus2x_24, diffus1x_36 and diffus2x_36, the horizontal diffusion is reduced. The damping time τ_0 is increased by a factor of 2 or 3 from the standard value of 12h to 24h or 36h. Furthermore, four experiments with additional noise in the smallest resolved scales are carried out: noise1x_3, noise2x_3, noise1x_6 and noise2x_6. '_3' and '_6' distinguishes between two noise intensities. The standard deviations of the noise σ_{noise} are 3×10^{-2} K in experiments noise1x_6 and noise2x_6) for temperature and $3 \times 10^{-7} s^{-1} (6 \times 10^{-7} s^{-1})$ for vorticity and divergence. These values are roughly 3 and 6 times larger than the values obtained in the experiment ctrl1x.

[12] In addition to the experiments described above, we will further use data obtained from a pre-industrial control-experiment with a higher resolution (atmosphere: T63L31, ocean: $1.5^{\circ} \times 1.5^{\circ}$).

3. Results

[13] How the additional noise or the reduced horizontal diffusion affect the variability of the system is illustrated in



Figure 1. (a)–(c) Standard deviations of spectral coefficients of experiment ctrl1x. Ratio of the standard deviations of the spectral coefficients: (d)–(f) diffus1x_3/ctrl1x, (g)–(i) noise1x_36/ctrl1x, and (j)–(l) ctrl1x(T63L31)/ctrl1x(T31L19). All estimates shown here are based on 6-hourly data and 50-year averages at model levels \approx 500 hPa. Please note the different color scales.

Figure 1. Figures 1a–1c show the standard deviations of the spectral coefficients of temperature, vorticity and divergence in ctrl1x. Low wavenumbers correspond to largescale components and high wavenumbers to small-scale components. Figures 1d-1i display the ratio of the standard deviations from experiments with modified small-scale components to the standard deviations obtained from the experiment ctrl1x. As expected the variability of spectral coefficients with high total wavenumbers is significantly enhanced if the horizontal diffusion is reduced or if noise is added to the small-scale components. (The other two experiments diffus1x 24 and noise1x 6, which are not shown, show qualitative the same behavior.) Note that the much larger gain of small-scale variability in noise1x 3 compared to diffus1x 36 is due to the rather high noise intensity used. By decreasing the noise intensity it would be possible to induce changes of similar magnitude.

[14] How well do our idealized experiments compare to the situation if the model resolution is actually increased? The experiment with higher resolution (atmosphere: T63L31, ocean: $1.5^{\circ} \times 1.5^{\circ}$) includes small-scale processes beyond the truncation scale of T31. In this experiment nonlinear scale-interactions across this scale are possible. In Figures 1j–11 the variability of the two different model resolutions are compared. When comparing Figures 1d-1i with Figures 1j-11 some similarities but also clear differences can be observed. The increased model resolution leads to enhanced small-scale variability of vorticity and divergence (similar to our idealized experiments). However, the temperature pattern differs. Spectral coefficients on smallest scales show a slightly higher variability. Except for the total wavenumbers l = 31, higher variability is mostly found for large zonal wavenumbers m, independent of *l*. There are also spectral coefficients, in particular those

Table 2. Global Means and Responses to a CO₂ Doubling for Temperatures at the Surface, $\langle T_{surf} \rangle$, and at 300 hPa, $\langle T300 \rangle$ in [K], Based on 50-Year Averages

	$\langle T_{surf} \rangle$, in [K]		$\langle T300 \rangle$, in [K]	
Experiment	1xCO ₂	$2xCO_2 - 1xCO_2$	1xCO ₂	$2xCO_2 - 1xCO_2$
ctrl	288.0	3.8	232.2	5.6
diffus 24	288.2	4.1	232.6	6.1
diffus 36	288.3	4.3	232.8	6.4
noise 3	287.9	3.9	230.8	5.1
noise_6	285.9	3.3	230.1	3.6

with small zonal wavenumbers and large total wavenumbers, which display less variability.

[15] The climate sensitivities (i.e., the changes of the global mean surface temperatures due to doubled CO_2 concentration) for all pairs of experiments are given in Table 2. In the control integrations the climate sensitivity is 3.8 K. This value increases to 4.3 K if the horizontal diffusion is reduced by a factor of three. The climate sensitivity in the noise experiments with moderate noise intensity does not significantly change, but it decreases to 3.3 K in noise2x_6 - noise1x_6.

[16] The influence of the enhanced small-scale fluctuations is not confined to the surface response. Rather, it increases with height and reaches the largest values in the upper troposphere (Figure 2). Reducing the horizontal diffusion by a factor of 3 leads to a maximum amplification of the temperature response at 250 hPa from 5.5 K to 6.4 K (16%). However, if white noise is added to the small-scale spectral coefficients, the temperature response in the troposphere is decreased. The warming in the experiments with high noise intensity (noise1x_6 and noise2x_6) is much smaller than the warming in the control experiments. In the experiments noise1x 3 and noise2x 3 the temperature response decreases only in the high troposphere with a maximum reduction at 200 hPa from 4.8 K to 4.1 K (15%). The changes in mean temperature responses are statistical significant in all tropospheric levels except for the vertical range of 1000 hPa to 450 hPa of noise2x 3 - noise1x 3.

[17] When enhancing the small-scale fluctuations not only the sensitivity of the temperature is changed but also the absolute mean temperature values of the $1 \times CO_2$ experiments. This is most pronounced in the noise experiments. The mean temperature at 300 hPa drops, for example, from 232.2 K in ctrl1x to 230.8 K in noise1x_3 (Table 2). Apart from the temperature also other climate variables are changed noticeable in the $1 \times CO_2$ -noiseexperiments. In experiment noise1x_3 we observe, for example, decreased zonal winds, a weakened atmospheric meridional circulation and less clouds (not shown). The changes of the general circulation could be explained by the 'tail wagging the dog' effect [*Frederiksen et al.*, 2003]. In experiment noise1x_6 these changes are enhanced and lead to a rather unrealistic climate.

4. Discussion and Conclusions

4.1. Differences of the Two Approaches

[18] At first sight it is surprising that the climate sensitivity increases when using a reduced horizontal diffusion but it tends to decrease when adding noise to the small-scale

components. Both methods enhance the small-scale variability and should therefore have a similar effect on the system. However, when having a closer look at the two methods, it becomes clear that they change the small-scale spectral coefficients in very different ways. The reduced horizontal diffusion results in less damping of small-scale eddies. Therefore, small-scale eddies are stronger and can interact more effectively with large-scale processes. These intensified small-scale eddies are correlated in time and across wavenumbers, and they are consistent with the governing equations of the model. In contrast, adding white noise to the small-scale components artificially perturbs the model. The resulting fluctuations are uncorrelated in time and spectral space. This could be appropriate, if the scales of the resolved variability were clearly separated from those of the dominant unresolved scales.

[19] Apart from the factors discussed above, the changes in the mean climate may also play a role. While in diffus1x 36 the mean climate is not very different in comparison to ctrl1x, it has changed in noise1x 3. Owing to the new climate in noise1x 3, the thermodynamical feedback processes (e.g., cloud feedbacks) are altered. In this case, the climate sensitivity can be changed both due to nonlinear dynamical and thermodynamical processes. Small-scale variability can therefore not only have a direct effect on the model sensitivity but also indirect effects via influencing the thermodynamical feedback processes. In general it is hard to distinguish between direct and indirect effects of enhanced small-scale fluctuations on the climate sensitivity. Nevertheless, the noise experiments might be affected more by indirect effects than the runs with reduced horizontal diffusion.

4.2. Validity of Horizontal Diffusion

[20] The results shown in Figure 2 suggest that a reduction of the horizontal diffusion leads to a higher climate sensitivity. The question arising is how well established is the strength and structure of the horizontal diffusion. Several studies [e.g., *Koshyk and Boer*, 1995; *Kaas et al.*, 1999; *Frederiksen and Kepert*, 2006] sought to improve the



Figure 2. Climate response of temperature to CO_2 forcing obtained from experiments with different representations of small scale fluctuations.

parameterization of the nonlinear interactions between subgrid scales and resolved scales. Besides confirming the idea of damping high wavenumbers stronger than lower wavenumbers, they found also a 'negative damping' at intermediate wavenumbers. This suggests that certain scales are enhanced instead of damped through the scale interactions with subgrid scales. The damping strength of the horizontal diffusion in climate models is generally tuned to ensure a kinetic energy spectrum close to observations. But, as mentioned by Stephenson [1995], why should the kinetic energy spectrum of a discrete model (especially near the truncation scale) look alike the spectrum of a continuous system? Enhanced small-scale energy might even be necessary to simulate the large-scale circulation and future climate changes correctly. Given the uncertainty of the form and strength of the horizontal diffusion in climate models, our results suggest that the modeled climate sensitivity to CO₂ forcing does not only depend on parameterizations related to thermodynamical feedback processes but also on the parameterization of nonlinear interactions between dynamical subgrid-scale processes and resolved scales.

[21] Acknowledgments. We thank Johann Jungclaus and Erich Roeckner for useful discussions. We thank Jorgen Frederiksen and one anonymous reviewer for their constructive comments and suggestions. This work was supported by the Max Planck Society and the International Max Planck Research School on Earth System Modelling.

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