

4 **On the increase of climate sensitivity and cloud feedback with warming in the**
5 **Community Atmosphere Models**

6 Jiang Zhu^{1,2*} and Christopher J. Poulsen¹

7 ¹Department of Earth and Environmental Sciences, University of Michigan, Ann Arbor,
8 MI 48109

9 ²Climate and Global Dynamics Laboratory, National Center for Atmospheric Research,
10 Boulder, CO 80305

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16 **Introduction**

17 Supporting information includes text, table and figures to support the discussion in main
18 article.

*Corresponding to: Jiang Zhu (jiangzhu@ucar.edu)

19 **Text S1. Estimation of the contribution from λ_{cld} to ECS increases**

20 We employ a bulk estimation method that has been used in previous studies (Gettelman
 21 et al., 2019; Gettelman, Kay, & Shell, 2012; Zelinka et al., 2020). To estimate the
 22 contribution from the cloud feedback to ECS changes, this method calculates a hypothetical
 23 ECS that would exist if the cloud feedback was changed but the radiative forcing and non-
 24 cloud feedbacks were kept unchanged. Specifically, ECSs for the climate states with two
 25 CO₂ levels in a model are expressed as

$$26 \quad \text{ECS}_1 = -\frac{F_1}{\lambda_{\text{noncld},1} + \lambda_{\text{cld},1}} \quad (1)$$

$$27 \quad \text{ECS}_2 = -\frac{F_2}{\lambda_{\text{noncld},2} + \lambda_{\text{cld},2}} \quad (2).$$

28 F is the radiative forcing of doubling CO₂, and λ_{noncld} and λ_{cld} are the “non-cloud” and the
 29 cloud feedback, respectively. The two background states are denoted using subscripts 1
 30 and 2, respectively.

31 With values of ECS, F , and λ_{cld} for two climate states in each model, we first compute
 32 λ_{noncld} using equations (1–2). We next calculate hypothetical ECS'_1 and ECS'_2 due to the
 33 cloud feedback changes, as

$$34 \quad \text{ECS}'_1 = -\frac{F_1}{\lambda_{\text{noncld},1} + \lambda_{\text{cld},2}} \quad (3)$$

$$35 \quad \text{ECS}'_2 = -\frac{F_2}{\lambda_{\text{noncld},2} + \lambda_{\text{cld},1}} \quad (4).$$

36 Finally, the contribution of the cloud feedback to the ECS change is estimated as

$$37 \quad \Delta\text{ECS} = \frac{(\text{ECS}'_1 - \text{ECS}_1) + (\text{ECS}_2 - \text{ECS}'_2)}{2} \quad (5).$$

38 The above estimation uses two-way calculations (forward in equation (3) and backward in
 39 equation (4)) to remove part of the dependence of ECS on the initial feedback parameter
 40 and forcing. CO₂ radiative forcing (F) are derived using the simplified equation in Byrne
 41 and Goldblatt (2014). Specifically, F are 3.9, 4.2, 4.6, and 5.4 W m⁻² at 1, 2, 4, and 16×
 42 PIC, respectively. Results are expected to depend on the assumption of CO₂ radiative
 43 forcing but our major conclusion that climate feedbacks dominate the state dependence of

44 ECS still holds based on findings in previous studies (Colman and McAvaney, 2009;
45 Meraner et al., 2013; Caballero and Huber, 2013; Zhu et al., 2019).

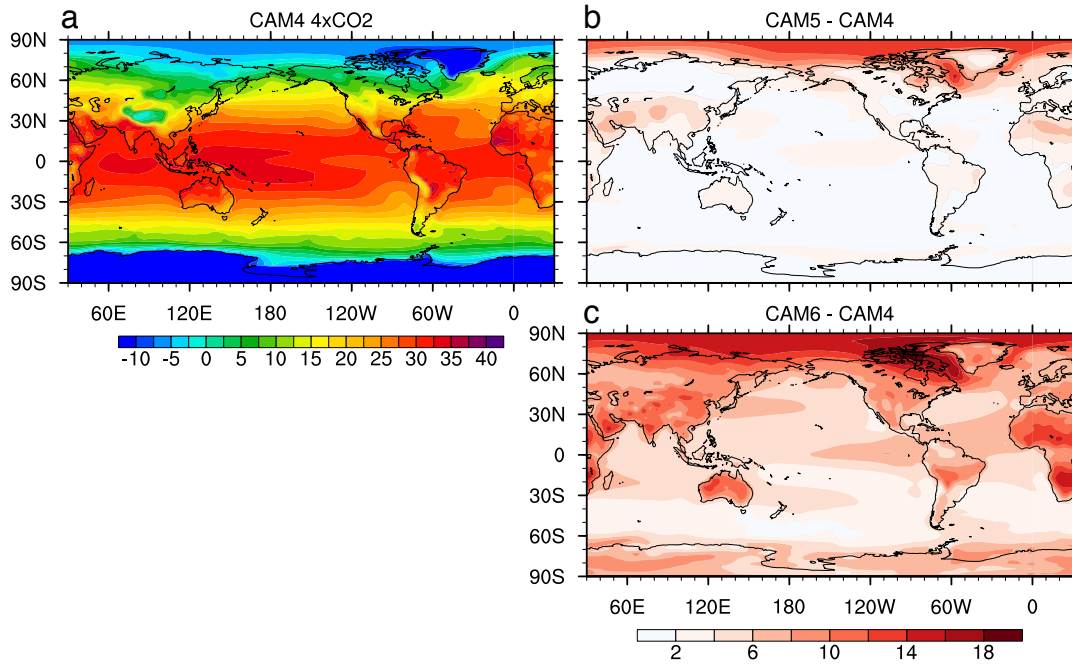
46 Using equations (1–5), we find that the cloud feedback approximately contributes 1.2–
47 1.4 °C (> 70%) to the ECS increases in CAM versions (Table S2). Our results suggest that
48 the non-cloud feedback does not change much in CAM6 (with a relatively narrower GMST
49 and CO₂ range) and slightly decreases in CAM5 and CAM4 with warming (by <~0.1 W
50 m⁻² K⁻¹). These results are overall consistent with partial radiative perturbation analyses in
51 previous Eocene simulations with a similar GMST range (Caballero & Huber, 2013; Zhu,
52 Poulsen, & Tierney, 2019). Caballero and Huber (2013) show that the increase of water
53 vapor feedback with warming is largely compensated by the associated decrease in the
54 lapse-rate feedback in CCSM3 (their Figure 3B). Zhu et al. (2019) further suggest that the
55 increase of the water vapor feedback with warming most likely results from the increase of
56 the cloud feedback by comparing CESM1-CAM5 and CESM1-CAM4 simulations (their
57 Figures 3B and S5).

58 **Text S2. Additional cloud-controlling factors over the low-latitude subsidence regime**

59 *The subsidence strength (ω)*. Observational and large-eddy simulation-based studies
60 (Bretherton, 2015; Myers & Norris, 2013) suggest that a stronger large-scale subsidence
61 (ω) is associated with thinner and shallower clouds, favoring weaker cloud radiative effects
62 ($\frac{\partial \text{CRE}}{\partial \omega} > 0$). In response to CO₂-induced global warming, the tropospheric subsidence (ω
63 at 700hPa) weakens in all CAM simulations ($\frac{d\omega}{d\text{GMST}} < 0$; Figure S2a), forming a negative
64 $\lambda_{\text{cld.}} \frac{d\omega}{d\text{GMST}}$ is not a constant; it increases with GMST for the first CO₂ doubling in CAM 4,
65 5, and 6 (Figure S2b). This suggests that the sensitivity of large-scale dynamical response
66 to global warming decreases with GMST, contributing to an increase of $\lambda_{\text{cld_subs}}$ with GMST.
67 In contrast, our CAM5 and CAM6 simulations exhibit a decrease of shortwave $\lambda_{\text{cld_subs}}$ with
68 GMST for the first CO₂ doubling (Figure 2m). Moreover, the increase of $\frac{d\omega}{d\text{GMST}}$ with
69 GMST appears to reach saturation when GMST exceeding ~20–23 °C, inconsistent with
70 the large increases of $\lambda_{\text{cld_subs}}$ in CAM4 (Figure 2m). Overall, these results suggest that the
71 large-scale dynamics is not a primary factor contributing to the $\lambda_{\text{cld_subs}}$ change with GMST.

72 ***The surface wind speed (U10).*** A decrease in lower wind speed (e.g. $U10$; wind speed
73 at a reference height of 10 m) weakens the surface driven shear mixing and the associated
74 latent heat flux, which thins low clouds ($\frac{\partial \text{CRE}}{\partial U10} < 0$) (Bretherton, Blossey, & Jones, 2013).
75 Our CAM simulations consistently show a decrease of wind speed with GMST ($\frac{dU10}{d\text{GMST}} <$
76 0 ; Figure S2c), contributing to a positive $\lambda_{\text{cld_subs}}$. Magnitude of $\frac{dU10}{d\text{GMST}}$ decreases with
77 GMST, which appears to be a robust feature in almost all CAM simulations. Together,
78 changes in wind speed contributes to a decrease of $\lambda_{\text{cld_subs}}$ with GMST. We suggest the
79 mechanism of wind speed change is likely important for CAM6, especially when all the
80 other cloud controlling factors act to increase CAM6 $\lambda_{\text{cld_subs}}$ with GMST (Figure 2,4 and
81 Figure S2).

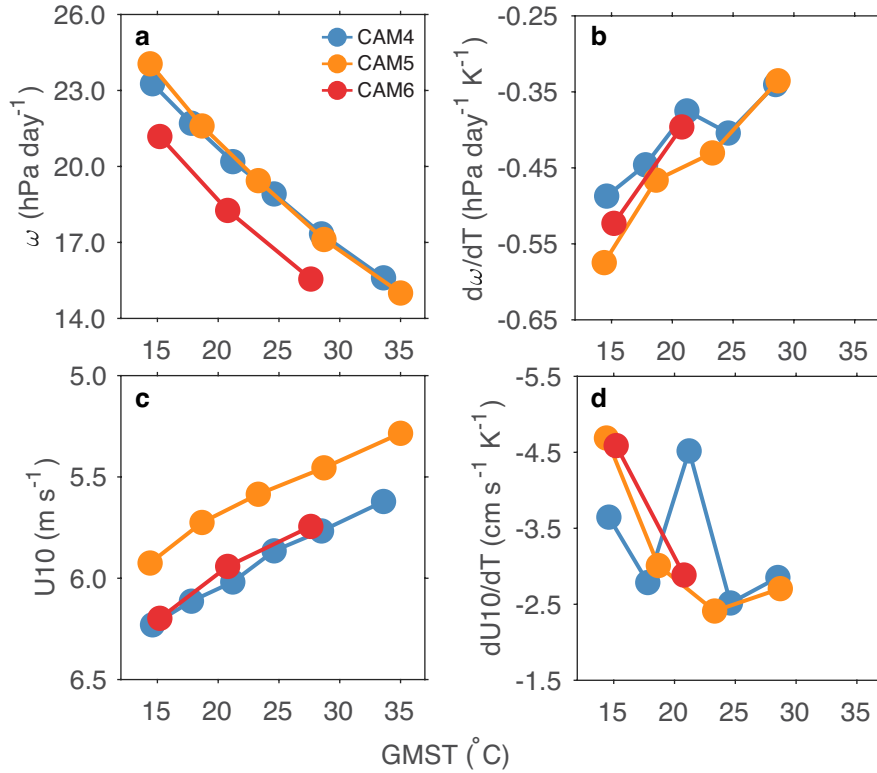
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83

84 **Figure S1.** (a) Annual mean surface temperature in CAM4 SOM simulation with 4× PIC
 85 (units: °C). (b) Difference in surface temperature between CAM5 and CAM4 SOM
 86 simulations with 4× PIC (units: °C). (c) The same as (b), but for the surface temperature
 87 difference in CAM6.

88



89

90 **Figure S2.** (a) ω at 700hPa averaged over the low-latitude subsidence regime as a function
 91 of GMST in the SOM simulations using CAM 4, 5, and 6. (b) The same as (a), but for the
 92 $\frac{d\omega}{dGMST}$, the sensitivity of ω to GMST changes. (c) and (d) The same as (a) and (b), but for
 93 the 10-m wind speed.

94

95 **Table S1.** λ_{cld} (units: $\text{W m}^{-2} \text{K}^{-1}$) diagnosed using PRP (first three columns) and APRP
96 (last column, shortwave-only). PRP results are with the year-to-year standard deviation
97 ($n=10$ in CAM4/5 and $n=3$ in CAM6). APRP decomposes shortwave λ_{cld} into
98 contributions from cloud scattering, amount, and absorption. Scattering and amount
99 feedbacks are shown in parentheses. Absorption changes little and is not shown. The
100 standard deviation for APRP analysis ranges from 0.01 to 0.03 $\text{W m}^{-2} \text{K}^{-1}$ and not listed.

		λ_{cld}	$\lambda_{\text{cld_LW}}$	$\lambda_{\text{cld_SW}}$	$\lambda_{\text{cld_SW_APRP}}$
CAM4	1× -> 2×	0.15±0.12	-0.02±0.05	0.16±0.10	0.11 (0.11, 0.11)
	2× -> 4×	--	--	--	0.29 (0.21, 0.19)
	4× -> 8×	0.21±0.07	0.04±0.04	0.18±0.07	0.12 (0.19, 0.04)
	8× -> 16×	--	--	--	0.30 (0.33, 0.09)
	16× -> 32×	0.37±0.05	-0.05±0.05	0.42±0.06	0.33 (0.38, 0.07)
CAM5	1× -> 2×	0.60±0.05	0.15±0.04	0.45±0.03	0.44 (0.06, 0.45)
	2× -> 4×	--	--	--	0.40 (0.09, 0.38)
	4× -> 8×	0.78±0.04	0.21±0.03	0.57±0.04	0.53 (0.26, 0.36)
	8× -> 12×	1.04±0.02	0.26±0.05	0.79±0.04	0.75 (0.44, 0.40)
CAM6	1× -> 2×	0.97±0.03	0.22±0.04	0.76±0.05	0.79 (0.32, 0.55)
	2× -> 4×	1.07±0.02	0.21±0.03	0.86±0.04	0.88 (0.46, 0.51)

101

102 **Table S2.** Calculation of ECS changes (ΔECS) that are attributed to the cloud feedback
 103 increases in CAM 4, 5, and 6. In each model, ECS ($^{\circ}\text{C}$), the cloud and non-cloud feedbacks
 104 (λ_{c} and λ_{nc} ; $\text{W m}^{-2} \text{K}^{-1}$), and CO_2 radiative forcing (F ; W m^{-2}) from two climate states with
 105 the lowest and highest possible CO_2 levels were used. Note the large uncertainty in CAM4
 106 λ_{c} at the $1\times\text{CO}_2$ level (see also Table S1).

	ECS_1	$\lambda_{\text{c},1}$	F_1	$\lambda_{\text{nc},1}$	ECS_2	$\lambda_{\text{c},2}$	F_2	$\lambda_{\text{nc},2}$	ΔECS
CAM6 (1, 2 \times)	5.5	0.97	3.9	-1.67	6.9	1.07	4.2	-1.69	0.9 (70%)
CAM5 (1, 4 \times)	4.2	0.60	3.9	-1.52	5.4	0.79	4.6	-1.65	1.0 (90%)
CAM4 (1, 16 \times)	3.2	0.15	3.9	-1.36	5.1	0.37	5.4	-1.42	0.8 (40%)

107

108 **Table S3.** Fraction area coverage (unitless) of the low-latitude subsidence and ascending
 109 regimes in CAM 4, 5, and 6 simulations. Fraction of the high-latitude cloud regime is
 110 invariant (0.47). CAM4 results include simulations with 1, 2, 4, 8, 16, and 32× PIC. CAM5
 111 results include 1, 2, 4, 8, and 12× PIC. CAM6 results include 1, 2, and 4× PIC.

		1×	2×	4×	8×	16× or 12×	32×
CAM4	low - subs.	0.33	0.33	0.33	0.32	0.33	0.34
	low - asce.	0.20	0.21	0.21	0.21	0.20	0.19
CAM5	low - subs.	0.33	0.32	0.31	0.31	0.32	--
	low - asce.	0.20	0.21	0.22	0.22	0.21	--
CAM6	low - subs.	0.34	0.33	0.33	--	--	--
	low - asce.	0.20	0.20	0.20	--	--	--

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