Non-monotonic feedback dependence on CO_2 due to a North Atlantic pattern effect

Ivan Mitevski¹, Yue Dong², Lorenzo M. Polvani^{1,2}, Maria Rugenstein³, Clara Orbe^{1,4}

¹Department of Applied Physics and Applied Mathematics, Columbia University, New York, NY ²Lamont-Doherty Earth Observatory, Columbia University, Palisades, NY ³Department of Atmospheric Science, Colorado State University, Fort Collins, CO ⁴NASA Goddard Institute for Space Studies, New York, NY

9 Key Points:

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10	•	Effective climate sensitivity (EffCS) and radiative feedbacks change non-monotonically
11		with increasing CO_2 concentrations
12	•	The non-monotonicity is associated with the formation of a sea-surface temperature
13		cooling pattern over the North Atlantic
14	•	Results imply an overlooked radiative damping effect on global EffCS from the North
15		Atlantic warming hole

 $Corresponding \ author: \ Ivan \ Mitevski, \ \texttt{im2527@columbia.edu}$

16 Abstract

Effective climate sensitivity (EffCS), commonly estimated from model simulations with 17 abrupt CO_2 quadrupling, has been shown to depend on the level of CO_2 forcing. To un-18 derstand this dependency systematically, we performed a series of simulations with a range 19 of abrupt CO₂ forcing in two climate models (CESM1-LE and GISS-E2.1-G). Our results 20 indicate that EffCS is a non-monotonic function of the CO₂ forcing, decreasing between 21 $3 \times$ and $4 \times CO_2$ in CESM1 (2× and $3 \times CO_2$ in GISS) and increasing at higher CO₂ levels. 22 The minimum EffCS value, caused by anomalously negative radiative feedbacks, arises from 23 sea-surface temperature (SST) cooling in the North Atlantic. This localized North Atlantic 24 cooling pattern is associated with the formation of the North Atlantic Warming Hole, ac-25 companied by the collapse of the Atlantic Meridional Overturning Circulation under CO_2 26 forcing. Our findings emphasize the importance of understanding changes in North Atlantic 27 SST patterns for constraining near-future and equilibrium global warming. 28

²⁹ Plain Language Summary

Estimates of effective climate sensitivity (EffCS) are complicated by 1) the nonlinear 30 dependence of feedback on temperature and 2) the sea-surface temperature (SST) pattern 31 effect. We find that EffCS and radiative feedbacks change non-monotonically with CO_2 32 concentrations due to a cooling SST pattern over the North Atlantic associated with the for-33 mation of a "North Atlantic Warming Hole" (NAWH). While most previous studies focused 34 on the impact of tropical Pacific SST patterns on EffCS, we here highlight an overlooked 35 damping effect on EffCS from North Atlantic SST cooling across CO₂ levels. Our results 36 imply that understanding and constraining the NAWH under CO₂ forcings is crucial for 37 transient warming projections and EffCS constraints. 38

39 1 Introduction

Equilibrium climate sensitivity (ECS), the equilibrium global-mean surface air temperature response to a doubling of atmospheric CO₂ relative to pre-industrial (PI) levels, is one of the most important metrics in climate science. The Charney 1979 report estimated a "likely" ECS range of 1.5-4.5K; most recently, a tighter range of ECS values between 2.6-3.9K was established using a Bayesian framework that combines multiple lines of evidence (Sherwood et al., 2020).

When evaluated from climate models, ECS is often approximated with an effective 46 47 climate sensitivity (EffCS), estimated from 150-year abrupt CO_2 quadrupling simulations within coupled global climate models (GCMs), with an underlying assumption that EffCS 48 remains constant with different CO₂ doublings. However, previous modeling (Bloch-Johnson 49 et al., 2021; Meraner et al., 2013; Mauritsen et al., 2019; Sherwood et al., 2020; Mitevski 50 et al., 2021; Zhu & Poulsen, 2020) and paleoclimate studies (Anagnostou et al., 2016, 2020; 51 Farnsworth et al., 2019; Friedrich et al., 2016; Shaffer et al., 2016; Zhu et al., 2019) have 52 shown that EffCS may not be linear with each successive CO₂ doubling. It tends to increase 53 at higher CO_2 values primarily due to a nonlinear temperature dependence of the radiative 54 feedbacks (λ), referred to as the state-dependence of feedbacks (Andrews et al., 2015; Bloch-55 Johnson et al., 2021; Sherwood et al., 2015), with minor contributions from nonlinear CO_2 56 dependence of radiative forcing (Mitevski et al., 2022). 57

⁵⁸ However, previous attempts to study the state dependence have been limited to CO₂-⁵⁹ doubling scenarios (2×, 4×, 8×CO₂) (Good et al., 2016; M. Rugenstein et al., 2019), whereas ⁶⁰ the shared socioeconomic pathway for the highest emission scenarios (SSP5-8.5) projects a ⁶¹ transient increase of greenhouse gas forcing up to 8×CO₂ at the year 2300 (Meinshausen ⁶² et al., 2020) passing through all the intermediate states of $n \times CO_2$ (n = 2, 3, 4, 5, 6, 7, 8). ⁶³ Moreover, previous studies have been focused on how EffCS and feedbacks vary in response ⁶⁴ to changes in global-mean temperatures under different CO₂ forcing (Meraner et al., 2013; ⁶⁵ Caballero & Huber, 2013; Bloch-Johnson et al., 2021), with little attention on how the ⁶⁶ spatial patterns of feedback and local surface warming respond to various CO₂ forcings. ⁶⁷ Hence in this study, we examine the dependence of EffCS on CO₂ levels systematically and ⁶⁸ its connection to the spatial patterns of the climate feedbacks by performing and analyzing ⁶⁹ a hierarchy of GCM experiments with a range of abrupt CO₂ forcings including $2\times$, $3\times$, ⁷⁰ $4\times$, $5\times$, $6\times$, $7\times$, and $8\times$ CO₂ relative to PI level (hereafter denoted as abrupt $n\times$ CO₂ ⁷¹ experiments).

⁷² 2 Materials and Methods

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2.1 Models and Experiments

We use the original large ensemble version of the Community Earth System Model (CESM1-LE). CESM1-LE comprises the Community Atmosphere Model version 5 (CAM5, 30 vertical levels) and parallel ocean program version 2 (POP2, 60 vertical levels) with approximately 1° horizontal resolution in all model components (Kay et al., 2015). Some of the results are shown with the GISS-E2.1-G model (Kelley et al., 2020) in Supplementary Information. All experiments in this work are with abrupt CO₂ forcing.

We perform abrupt $n \times CO_2$ experiments with the coupled version of the CESM1-LE and GISS-E2.1-G models (coupled runs) for 150 years with $2 \times$, $3 \times$, $4 \times$, $5 \times$, $6 \times$, $7 \times$, and $8 \times CO_2$ forcing, with all other trace gases, aerosols, ozone concentrations, and solar forcing fixed at PI values. The response is defined as the difference between the $n \times CO_2$ runs and the PI control run. The same experiments were analyzed in (Mitevski et al., 2021, 2022).

To estimate the effective radiative forcing (ERF) as per Forster et al. (2016), we perform prescribed pre-industrial SST and sea-ice runs for 30 years for each $2\times$, $3\times$, $4\times$, $5\times$, $6\times$, $7\times$, and $8\times$ CO₂. The ERF is then calculated as the global mean net top of the atmosphere (TOA) net radiation between PI and $n\times$ CO₂, and it includes the stratospheric and tropospheric adjustments (Sherwood et al., 2015).

We also utilize atmosphere-only runs (AGCM) with prescribed monthly SST values taken from the 150-year abrupt $n \times CO_2$ runs. The prescribed SST values are monthly data for 150 years. The CO₂ concentration, ozone concentrations, aerosols, solar forcing, and all other trace gases are fixed at pre-industrial values.

In addition to only prescribing SST values from the $n \times CO_2$ runs, we also change the SST patterns. We use the pattern from $3 \times CO_2$ in CESM1-LE and then scale the pattern by the global-mean warming amplitude from $4 \times CO_2$ and $5 \times CO_2$. We do this by

$$\Delta SST(x, y, t) = SST_{3 \times CO_2}(x, y, t) - SST_{PI}(x, y),$$

t is monthly data from 150 years, x is longitude, and y is latitude. Next, we find the pattern S_p as

$$S_p(x, y, t) = \frac{\Delta \text{SST}(x, y, t)}{\overline{\Delta}\text{SST}(t)}$$

where $\overline{\Delta SST}$ is the global mean monthly data for 150 years. Then we have

$$\Delta \mathrm{SST}'_{n \times \mathrm{CO}_2}(x, y, t) = S_p \cdot \overline{\mathrm{SST}_{n \times \mathrm{CO}_2}}.$$

And finally

$$SST_{n \times CO_2}(x, y, t) = SST_{PI}(x, y) + \Delta SST'_{n \times CO_2}(x, y, t).$$

 $_{94}$ One caveat here is that we are only changing the SSTs, and holding sea-ice fixed at $3 \times CO_2$.

Although sea-ice changes also cause albedo feedback changes, their contribution is much

⁹⁶ smaller than the SST-mediated feedback changes.

97 2.2 Analysis

We calculate the effective climate sensitivity EffCS as the x-intercept of regressing the change in net TOA radiation against surface air temperature over the 150 years of the simulations (Gregory et al., 2004; Zelinka et al., 2020). We normalize the EffCS by $\log_2 n$ for the $n \times CO_2$ runs, assuming logarithmic CO₂ forcing, consistent with Bloch-Johnson et al. (2021).

We calculate individual feedbacks with radiative kernels from Pendergrass et al. (2018). For each year, we multiply the spatially resolved kernels by the climate field anomalies of atmospheric temperature T, water vapor q, and surface albedo α . We regress these quantities on the surface temperature response, and the slope of this regression is the feedback. The cloud feedbacks are computed via the residual method (Soden & Held, 2006).

108 **3 Results**



¹⁰⁹ **3.1** Non-monotonic effective climate sensitivity and radiative feedbacks

Figure 1. a) Global mean surface air temperature response (ΔT_s) , b) effective climate sensitivity (EffCS), c) effective radiative forcing (ERF) from 30-year fixed sea-surface temperature runs, and d) net feedback parameter (λ) from the 150-year Gregory regression of abrupt $n \times CO_2$ runs.

Results from CESM1-LE show that although the global-mean surface air temperature increases monotonically as CO₂ increases (Fig. 1a), EffCS changes non-monotonically with CO₂ levels (Fig. 1b). That is, EffCS decreases between $3 \times$ and $4 \times CO_2$ and then increases between $4 \times$ and $5 \times CO_2$, and at higher CO₂ forcing, with a minimum value at $4 \times CO_2$. We find the same non-monotonicity in the GISS-E2.1-G experiments except with a minimum EffCS at $3 \times CO_2$ (Fig. S1). In the rest of the paper, we will focus on the CESM1-LE simulations and note that the results hold for the GISS-E2.1-G simulations unless otherwisenoted.

Changes in EffCS, in principle, are governed by changes in effective radiative forcing 118 (ERF) and radiative feedbacks (λ). While the ERF, calculated from an additional 30-year 119 fixed sea-surface temperature (SST) runs as per Forster et al. (2016), increases slightly more 120 than the logarithm of the CO_2 concentration at higher CO_2 levels than $4 \times CO_2$ (see Mitevski 121 et al. (2022) for more detail), it is strongly monotonic with CO_2 and does not exhibit a min-122 imum value (Fig. 1c). On the other hand, the net radiative feedback parameter λ (Fig. 1d), 123 calculated from 150-year regressions of top-of-atmosphere (TOA) radiative response against 124 surface air temperature change (Zelinka et al., 2020), exhibits a clearly non-monotonic be-125 havior with respect to CO_2 levels, as for EffCS: λ becomes more negative (more stabilizing) 126 between $3 \times$ and $4 \times CO_2$ and less negative between $4 \times$ and $5 \times CO_2$, corresponding to the 127 lowest EffCS at $4 \times CO_2$. Similar results are also found in the GISS-E2.1-G model experi-128 ments (Fig. S1). These results suggest that EffCS depends not only *nonlinearly* on CO₂, as 129 found in previous studies (Meraner et al., 2013; Caballero & Huber, 2013; Bloch-Johnson et 130 al., 2021), but also *non-monotonically*, and that the non-monotonicity is caused by the ra-131 diative feedbacks in our simulations. Hence the question is: what causes the non-monotonic 132 changes in feedbacks? 133



Figure 2. a) Global net feedback parameter λ from coupled runs, AGCM prescribed-SST runs with SSTs from coupled runs, and prescribed-SST runs with $3 \times CO_2$ pattern, where the $3 \times CO_2$ SST patterns are scaled with the actual global-mean SST values of $4 \times CO_2$ and $5 \times CO_2$, respectively. Spatial patterns of the local contribution to the global λ at $4 \times CO_2$ from b) prescribed-SST, c) prescribed-SST with $3 \times CO_2$ pattern and d) the difference.

3.2 Non-monotonic λ traced to changes in surface warming patterns

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We hypothesize two reasons for the non-monotonic changes in λ with CO₂:

- 1. The non-monotonic dependence in λ may arise from a nonlinear state-dependence of the feedbacks. As noted above, previous studies have found that radiative feedbacks change nonlinearly with global-mean surface temperature changes (i.e., feedback temperature dependence), mostly owing to the cloud and water vapor feedbacks (Meraner et al., 2013; Caballero & Huber, 2013; Bloch-Johnson et al., 2021). Can the changes in global-mean surface temperature across the CO₂ levels in our simulations (Fig. 1a) explain the non-monotonic behavior of λ and, therefore, EffCS?
- 2. The non-monotonic dependence of λ may arise from a strong dependence of λ on the spatial pattern of SSTs. Recent studies have found a close coupling between SST patterns and radiative feedbacks in observations and model simulations, the so-called "pattern effect" (Zhou et al., 2016; Dong et al., 2019; Sherwood et al., 2020). If the SST pattern effect caused the non-monotonic response in λ , then what SST regions govern the global and local changes in our feedbacks?

To test the hypotheses, we run the atmospheric component of the coupled model 149 CESM1-LE (CAM5) with specified SST boundary conditions, in order to examine the im-150 pacts of different surface warming on λ . First, we perform a set of 150-year long CAM5 151 simulations where we fix all radiative forcing agents at pre-industrial levels, and prescribe 152 the time-varying SSTs produced by the corresponding coupled model $n \times CO_2$ simulations. 153 In these runs (denoted as "prescribed-SST"), TOA radiative fluxes and surface air tempera-154 ture freely adjust to the underlying SSTs. Although not directly forced by CO_2 , we find that 155 the prescribed-SST simulations accurately reproduce the values of λ from the corresponding 156 coupled simulations (c.f. blue and black dots in Fig. 2a). This finding, consistent with other 157 studies (Haugstad et al., 2017; Zhou et al., 2023), suggests that the dependence of λ on CO₂ 158 forcing is primarily shaped by the SSTs induced by the CO_2 forcing and therefore confirms 159 the validity of using prescribed-SST simulations to study radiative feedbacks to understand 160 the coupled $n \times CO_2$ results. 161

Next, we perform another set of prescribed-SST simulations with adjusted SST bound-162 ary conditions. To test hypothesis # 1, i.e., whether λ responds non-monotonically to 163 changes in global-mean surface temperatures, we conduct simulations where we scale the 164 SST pattern from $3 \times CO_2$ by the actual global-mean SST changes in coupled $4 \times CO_2$ and 165 $5 \times CO_2$, respectively. Such that these two runs have the same normalized global SST pat-166 tern (at every monthly time step) as the $3 \times CO_2$ run but different global-mean SST values 167 (denoted "prescribed-SST with $3 \times CO_2$ pattern"). In these experiments, we find that the λ 168 values do not reproduce those in the coupled & prescribed-SST simulations even though the 169 same global-mean SST warming is prescribed (c.f. red and blue dots in Fig. 2a), suggesting 170 that the non-monotonic response in λ arises from changes in the spatial pattern of SSTs 171 (hypothesis # 2) and not the changes in the global-mean values of SSTs (hypothesis # 1). 172

The above CAM5 prescribed-SST simulations highlight the role of SST patterns in 173 driving the non-monotonic response in λ . To understand what regions contribute to this 174 non-monotonicity, we show the spatial pattern of λ calculated as the local net TOA radiation 175 regressed to global-mean surface air temperature response, shown in Fig. 2b-d. The spatial 176 pattern of λ in the 4×CO₂ prescribed-SST run (4×CO₂ SST pattern), is shown in Fig. 2b, 177 corresponding to the globally averaged λ at $4 \times CO_2$ shown by the blue dot in Fig. 2a. The 178 spatial pattern of λ in a 4×CO₂ run with 3×CO₂ SST pattern (red dot in Fig. 2a) is shown 179 in Fig. 2c. Taking the difference between Fig. 2b and c (panel d) shows substantially more 180 negative feedback in the North Atlantic with the $4 \times CO_2$ pattern, and not much change 181 when we use the $3 \times CO_2$ pattern, indicating that the anomalously low EffCS at $4 \times CO_2$ in 182 our coupled simulations is primarily associated with an anomalously negative λ in the North 183 Atlantic. 184



Figure 3. Maps of SST patterns (calculated as the regression of local temperature changes to global temperature changes for 150 years) in the coupled runs for a) $3 \times CO_2$, b) $4 \times CO_2$, and c) $5 \times CO_2$. The differences between $4 \times$ and $3 \times CO_2$, and $5 \times$ and $4 \times CO_2$, are shown in d) and e), respectively. Figures f-j) show λ maps for the same CO₂ experiments.

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3.3 A local pattern effect from the North Atlantic

While the stronger negative feedbacks appear to be located mainly in the North At-186 lantic, it is unclear whether they are driven by the local North Atlantic SST changes, or by 187 remote SST impacts from other basins. In Fig. 3, we show the normalized SST patterns 188 (over the full 150 years of the simulations) from $3\times$, $4\times$, and $5\times$ CO₂ simulations (panels 189 a-c). We find that anomalous SST cooling primarily occurs in the North Atlantic: $4 \times CO_2$ 190 produces a strong cooling pattern in the North Atlantic, largely resembling the pattern of 191 the North Atlantic Warming Hole (NAWH) (Chemke et al., 2020). However, this North 192 Atlantic relative-cooling pattern does not emerge at $3 \times CO_2$ (panel a) and is much weaker 193 at $5 \times CO_2$ (panel c). Concurrently, we find that local feedbacks exhibit patterns that closely 194 match the SST patterns (Fig. 3f,g, and h). Most of the strengthening of negative feedback 195 that would result in lower EffCS is found at $4 \times$ relative to $3 \times CO_2$ (Fig. 3i), and it occurs 196 in the North Atlantic, corresponding with the local cooling pattern (Fig. 3d); while most 197 of the weakening of feedbacks at $5 \times$ relative to $4 \times CO_2$ (higher EffCS, Fig. 3j) which also 198 occurs in the North Atlantic, corresponds with the local warming pattern (Fig. 3e). These 199 results suggest that the non-monotonic response of the feedbacks found in our simulations 200 (Fig. 2a) is predominately from feedback changes in the North Atlantic, associated with 201 North Atlantic local SST changes. We note that significant feedback changes also occur in 202 the tropical Pacific (Fig. 3i,j), particularly the tropical Eastern Pacific, but these feedback 203 changes are in the opposite sign to the global-mean feedback changes, and thus cannot account for the total feedback response we showed in Fig. 2a. While some other regions 205 may contribute to the negative feedback change (e.g., the tropical Western Pacific and the 206 Southern Ocean), we find that the North Atlantic local λ (area between 0 to 60N and 80W 207

to 10E) explains up to 2/3 of the total change in the global-mean λ (Fig. S2). This suggests

that most of the non-monotonicity at $4 \times CO_2$ is due to the North Atlantic pattern effect.



Figure 4. Maps of individual feedbacks calculated from prescribed-SST runs for: a-c) net, d-f) Planck, g-i) lapse rate + water vapor, j-i) net cloud.

To further understand the processes causing the λ non-monotonicity, we further de-210 compose the net feedback parameter λ into the individual feedbacks using radiative kernels 211 (Pendergrass et al., 2018) (Fig. 4). In the North Atlantic at $4 \times CO_2$, the Planck feedback 212 (Fig. 4e) is strongly positive as the local cooling reduces outgoing radiation, whereas the 213 combined lapse rate and water vapor feedback (Fig. 4h) and the cloud feedback (Fig. 4k) 214 contribute negatively. The strong negative feedback at $4 \times CO_2$ compared to $3 \times CO_2$ in the 215 subtropical North Atlantic is primarily due to the SW cloud feedback (Fig. S2 and Fig. S3h); 216 hence, it is one of the key contributors to the λ non-monotonicity at 4×CO₂. This is be-217 cause local cooling strengthens lower tropospheric stability (often measured as estimated 218 inversion strength (EIS)), which increases low-cloud cover (negative cloud feedback) and 219 more-negative lapse rate feedback. This mechanism is consistent with the leading mecha-220 nism found in the tropical Pacific pattern effect (Zhou et al., 2016; Andrews & Webb, 2018; 221 Dong et al., 2019), except this pattern effect here is associated with the North Atlantic SST 222 changes (Lin et al., 2019), and causes the non-monotonic response in EffCS and λ across 223 CO_2 levels in our experiments. We refer the reader to Lin et al. (2019) for more details on 224 this mechanism. Additionally, it is important to note that 2/3 of the difference in feedbacks 225 between $4 \times CO_2$ and $3 \times CO_2$ comes from the North Atlantic and 1/3 from the rest of the 226 globe. At $4 \times CO_2$, there are strong responses in the individual feedbacks in the tropical Pa-227 cific (see Fig. S3 for albedo and longwave cloud feedbacks). However, the negative Planck 228 feedback response in the tropical Pacific is compensated by the local positive feedback re-229 sponse from lapse rate, water vapor, and clouds (Fig. 4b,e,h,k), which makes the tropical 230 Pacific less pronounced in the λ non-monotonic changes. 231

Having shown that feedback changes primarily come from the North Atlantic associated 232 with local SST cooling, we finally return to the key pattern of North Atlantic SST cooling 233 found in our simulations, the North Atlantic warming hole (NAWH). In the literature, the 234 appearance of the NAWH has been attributed to the slowdown in the AMOC and linked to 235 an atmospheric response (Rahmstorf et al., 2015; Sévellec et al., 2017; Caesar et al., 2018; 236 Latif et al., 2022). Our previous work (Fig. S3 in (Mitevski et al., 2021)) found the North 237 Atlantic cooling (NAWH) in our experiments is primarily due to AMOC collapse. The 238 AMOC collapses at $4 \times CO_2$ in our GCM, and at all other higher CO_2 forcings. At higher 239 CO_2 forcings (5× CO_2 and above), the AMOC collapse no longer produces anomalous North 240 Atlantic cooling compared to the previous level of CO_2 forcing (e.g., $4 \times CO_2$) because the 241 AMOC collapse-induced SST cooling is further overwhelmed by the surrounding warming. 242 Hence, the cooling over the NAWH is less pronounced at higher CO₂ forcings (Fig. 3c) and 243 has a smaller impact on the feedbacks (Fig. 3h). The collapse of the AMOC under CO_2 244 forcing has been widely reported in climate models, including the GISS-E2.1-G model in this 245 study (occurring at $3 \times CO_2$ & higher) and many other CMIP5 and CMIP6 models (Fig. S3 246 in (Mitevski et al., 2021)). 247

²⁴⁸ 4 Discussion and Conclusion

In a series of $n \times CO_2$ (n = 2, 3, 4, 5, 6, 7, 8) experiments, we find a non-monotonic response in the effective climate sensitivity (EffCS) to CO_2 forcing using two state-of-the-art, coupled climate models. EffCS becomes anomalously low at an intermediate level of CO_2 ($4 \times CO_2$ in CESM1-LE and $3 \times CO_2$ in GISS-E2.1-G) but increases at higher CO_2 levels. This EffCS non-monotonicity is primarily linked to changes in radiative feedback λ due to North Atlantic cooling; λ becomes anomalously negative when cooling emerges in the North Atlantic and forms a North Atlantic Warming Hole (NAWH).

The dependence of λ on sea-surface temperature (SST) patterns has been widely stud-256 ied, with a focus on the time-evolution of those patterns (Andrews et al., 2015; Zhou et 257 al., 2016; Dong et al., 2019; Andrews et al., 2022; Sherwood et al., 2020). For example, 258 estimates of EffCS from the observed historical energy budget constraints are lower than 259 those from long-term warming under CO_2 quadrupling, primarily owing to changes in the 260 tropical Pacific SST patterns (Zhou et al., 2016; Dong et al., 2019; Andrews et al., 2018, 261 2022; Gregory et al., 2020). This "pattern effect" has been studied with a Green's func-262 tion approach (Dong et al., 2019; Zhou et al., 2017; Zhang et al., 2023), which shows that 263 the global feedback has a predominant dependence on tropical western Pacific SSTs and 264 is less sensitive to the North Atlantic SSTs. This tropical Pacific SST pattern effect has 265 been found to be a leading mechanism for the time evolution of EffCS estimates. However, 266 our study proposes a new North Atlantic pattern effect that accounts for changes in EffCS and feedbacks across different CO₂ forcing levels. This North Atlantic pattern effect shows 268 that SST cooling in the North Atlantic due to the formation of NAWH causes λ to become 269 more negative and, therefore, lower EffCS. We note that the North Atlantic pattern effect 270 operates on the dimension of increasing CO_2 forcing, instead of on the dimension of time 271 evolution reported in previous studies (Dong et al., 2019; Andrews & Webb, 2018; Andrews 272 et al., 2022; Zhou et al., 2016). 273

The NAWH has been proposed to arise from the reduction of surface meridional ocean 274 heat transport (Chemke et al., 2020) or AMOC slowdown that reduces transient warming 275 due to increased ocean heat uptake (Caesar et al., 2020; Palter, 2015; M. A. Rugenstein et 276 al., 2013; Trossman et al., 2016; Winton et al., 2013). In our study, we find that the NAWH 277 can further reduce EffCS and transient warming by causing more negative feedback (more 278 efficient radiative damping at the top of the atmosphere). The fact that the NAWH has 279 been observed in the historical period and is projected to persist in future scenarios with 280 increasing GHG (Menary & Wood, 2018; Chemke et al., 2020; Keil et al., 2020; Gervais et 281 al., 2018; Liu et al., 2020; Ren & Liu, 2021) suggests a considerable damping effect on future 282 global warming from the North Atlantic, which may have been overlooked in the literature. 283

We further analyzed two subsets of CMIP6 models with and without NAWH in the 284 abrupt- $4 \times CO_2$ runs (Fig. S4). Models with NAWH in the abrupt- $4 \times CO_2$ scenario also 285 show more surface cooling in the North Atlantic in transient 21st-century simulations (under 286 both SSP5-8.5 and SSP2-4.5 scenarios) than models without NAWH. This suggests that 287 uncertainty in the projected long-term North Atlantic SST patterns in response to abrupt 288 CO₂ forcing also persists in transient projections. Thus, understanding North Atlantic SST 289 changes is crucial for constraining global climate change at both transient and equilibrium 290 timescales. 291

292 One caveat to our findings is that the AMOC collapse in our models occurs at $3 \times$ and $4 \times CO_2$, which are relatively low CO_2 values, where the collapse can induce a substantial 203 cooling in the North Atlantic. When the AMOC collapses at a low CO_2 value, such as $2\times$ 294 or $3 \times CO_2$, the North Atlantic cooling is strong, leading to a considerable non-monotonicity 295 in EffCS. However, if the AMOC collapses at a higher CO_2 value, such as $5 \times CO_2$, then the 296 overwhelming CO₂ warming from the surrounding areas results in a weaker North Atlantic 297 SST cooling pattern. In this case, the EffCS non-monotonicity would be smaller than the one 298 reported in this study. Hence our results suggest that future changes in AMOC and NAWH 299 may add additional uncertainty to EffCS and transient 21st century warming projections. 300 Finally, we recognize that our results are obtained using only two GCMs. It would be 301 important to repeat the same exercise with a broad range of models to test the robustness 302 of our results. 303

The fact that EffCS is nonlinear and even non-monotonic with respect to CO_2 lev-304 els complicates equilibrium climate sensitivity constraints using models, observations, the 305 paleoclimate record, and process-based understanding. While the non-constant λ across 306 307 different CO₂ levels has been mainly attributed to feedback temperature dependence within models (Bloch-Johnson et al., 2021; Meraner et al., 2013; Mauritsen et al., 2019; Sherwood 308 et al., 2020; Zhu & Poulsen, 2020) and paleoclimate records (Anagnostou et al., 2016, 2020; 309 Farnsworth et al., 2019; Friedrich et al., 2016; Shaffer et al., 2016; Zhu et al., 2019), we 310 here have shown that the SST pattern also plays a role. Our study adds additional evidence 311 of EffCS state dependence and pattern effects, which need to be further examined in other 312 lines of evidence. 313

³¹⁴ 5 Open Research Section

Part of the computing and data storage resources, including the Cheyenne supercomputer (https://doi.org/10.5065/D6RX99HX), were provided by the Computational and Information Systems Laboratory at National Center for Atmospheric Research (NCAR). The CESM-LE model data can be obtained at https://doi.org/10.5281/zenodo.5725084 and GISS-E2.1-G model data at https://doi.org/10.5281/zenodo.3901624.

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Supporting Information for "Non-monotonic feedback dependence on CO_2 due to a North Atlantic pattern effect"

Ivan Mitevski¹, Yue Dong², Lorenzo Polvani^{1,2}, Maria Rugenstein³, Clara

 $Orbe^{1,4}$

¹Department of Applied Physics and Applied Mathematics, Columbia University, New York, NY

²Lamont-Doherty Earth Observatory, Columbia University, Palisades, NY

 $^{3}\mathrm{Department}$ of Atmospheric Science, Colorado State University, Fort Collins, CO

 $^4\mathrm{NASA}$ Goddard Institute for Space Studies, New York, NY

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Figure S1. Same as Figure 1 in main text but for the GISS-E2.1-G model.



Figure S2. Individual feedback difference between $4 \times CO_2$ and $3 \times CO_2$ for Global, Global -North Atlantic, and the North Atlantic region. The North Atlantic is defined as a box between 0 to 60N and 80W to 10E.



Figure S3. Maps of individual feedbacks calculated from prescribed-SST runs for: a-c) albedo, d-f) longwave cloud, and g-i) shortwave cloud.



Figure S4. Maps of surface temperature patterns from two CMIP6 models composites with (a,b,c) and without (d,e,f) North Atlantic Warming Hole (NAWH), defined as cooling in the North Atlantic, and the difference (g,h,i). Composites are shown for SSP2-4.5 (a,d,g), SSP5-8.5 (b,e,h), and abrupt-4xCO2 scenario (c,f,i). The models without NAWH are ACCESS-CM2, AWI-CM-1-1-MR, CAMS-CSM1-0, CMCC-CM2-SR5, CanESM5, INM-CM4-8, IPSL-CM6A-LR, MIROC6, MPI-ESM1-2-HR, MPI-ESM1-2-LR. Models with NAWH are BCC-CSM2-MR, CESM2-WACCM, FGOALS-g3, GFDL-ESM4, IITM-ESM, KACE-1-0-G, MRI-ESM2-0, NorESM2-MM, TaiESM1. The surface temperature patterns are calculated as local surface temperature changes regressed to global surface temperature response for years 2015 to 2100 for the SSP scenarios and the first 150 years of the abrupt-4xCO2 runs, and then averaged across models.