

1 **Antarctic ice-sheet meltwater reduces transient**
2 **warming and climate sensitivity through the**
3 **sea-surface temperature pattern effect**

4 **Yue Dong^{1,2}, Andrew G. Pauling³, Shaina Sadai⁴, Kyle C. Armour^{3,5}**

5 ¹Lamont-Doherty Earth Observatory, Columbia University, Palisades, NY, USA

6 ²Cooperative Programs for the Advancement of Earth System Science, University Corporation for

7 Atmospheric Research, Boulder, CO, USA

8 ³Department of Atmospheric Sciences, University of Washington, Seattle, WA, USA

9 ⁴Union of Concerned Scientists, Cambridge, MA, USA

10 ⁵School of Oceanography, University of Washington, Seattle, WA, USA

11 **Key Points:**

- 12 • Accounting for Antarctic meltwater input in a GCM reduces the global warming
13 rate and produces a warming pattern closer to the observed
14 • Antarctic meltwater impacts not only the Southern Ocean, but also the tropics
15 via teleconnections
16 • The reduced global warming rate is driven by changes in both ocean heat uptake
17 efficiency and radiative feedbacks

Corresponding author: Yue Dong, yd2644@columbia.edu

Abstract

Coupled global climate models (GCMs) generally fail to reproduce the observed sea-surface temperature (SST) trend pattern since the 1980s. The model-observation discrepancies may arise in part from the lack of realistic Antarctic ice-sheet meltwater imbalance in GCMs. Here we employ two sets of CESM1-CAM5 simulations forced by anomalous Antarctic meltwater fluxes over 1980–2013 and into the 21st century. Both show a reduced global warming rate and an SST trend pattern that better resembles observations. The meltwater drives surface cooling in the Southern Ocean and the tropical southeast Pacific, in turn increasing low-cloud cover and driving radiative feedbacks to become more stabilizing (corresponding to a lower effective climate sensitivity). These feedback changes contribute more than ocean heat uptake efficiency changes in reducing the global warming rate. Accurately projecting historical and future warming thus requires improved representation of Antarctic meltwater and its impacts in models.

Plain Language Summary

Observations have shown surface cooling in the Southern Ocean and the tropical southeast Pacific over the last four decades. However, global climate models generally struggle to reproduce this pattern. The model-observation mismatch has been proposed to partly arise from the fact that models lack in representation of realistic Antarctic ice-sheet meltwater. Here we revisit two sets of simulations with meltwater fluxes and examine the impact of meltwater input on global warming and the global energy budget. We find that accounting for meltwater input delays the global warming and produces a surface warming pattern closer to recent observations. The reduced global warming rate is caused by both more efficient ocean heat uptake and stronger radiative feedbacks (more efficient radiative damping) that are associated with changes in the surface warming pattern. These results indicate a critical impact of Antarctic meltwater on the global climate that has been missed in current climate models.

1 Introduction

The observed sea-surface temperature (SST) trend pattern since about 1979 is characterized by a strengthened west-east gradient in the tropical Pacific and a north-south hemispheric asymmetry: the tropical western Pacific and the Arctic have warmed while the tropical southeast Pacific and the Southern Ocean have cooled (England et al., 2014; Armour et al., 2016; Watanabe et al., 2021; Wills et al., 2022). The observed surface cooling in the Southern Ocean has been accompanied by an expansion of Antarctic sea ice (Fan et al., 2014; Parkinson, 2019), broad surface freshening (De Lavergne et al., 2014; Durack, 2015) and sub-surface warming (Gille, 2008; Armour et al., 2016). Yet, all these observed regional features are generally missed in global-climate model (GCM) simulations driven by historical forcings (Luo et al., 2018; Kostov et al., 2018; Chung et al., 2022; Seager et al., 2022; Wills et al., 2022; Roach et al., 2020), with many models also overestimating the global-mean warming rate over this period (Jiménez-de-la Cuesta & Mauritsen, 2019; Nijssen et al., 2020; Tokarska et al., 2020).

Various hypotheses have been put forward to explain the observed changes and model-observation discrepancies (e.g., Andrews et al., 2022). One of the leading hypotheses for the observed changes in the Southern Ocean is freshwater input from the melt of the Antarctic ice sheet and ice shelves (referred to here as Antarctic “meltwater”). The fact that the current generation of GCMs are unable to accurately represent Antarctic meltwater imbalance may explain some of the model-observation discrepancies in the Southern Ocean. Indeed, numerous studies have shown that adding an Antarctic meltwater imbalance in GCMs can produce anomalous surface cooling, subsurface warming, and sea-ice expansion around Antarctica, owing to an increase in upper ocean stratification, reducing the vertical heat flux from the relatively warm subsurface waters below (Kirkman

68 & Bitz, 2011; Ma & Wu, 2011; Bintanja et al., 2013; Swart & Fyfe, 2013; Pauling et al.,
 69 2016; Armour et al., 2016; Bronselaer et al., 2018; Purich et al., 2018; Schloesser et al.,
 70 2019; Park & Latif, 2019; Sadai et al., 2020; Rye et al., 2020). Although the amount of
 71 Antarctic meltwater input needed to cause significant changes in the Southern Ocean
 72 is highly model dependent (e.g., Bintanja et al., 2013; Swart & Fyfe, 2013; Pauling et
 73 al., 2016), meltwater forcing brings projected SST trend patterns closer to those observed
 74 in all models it has been tested in.

75 Antarctic meltwater input may also have remote impacts, with the potential to ex-
 76 plain some of the observed SST trends and model-observation discrepancies in the tropi-
 77 cal Pacific. Ma and Wu (2011) demonstrated that adding anomalous Antarctic melt-
 78 water in a coupled GCM resulted in surface cooling extending from the Southern Ocean
 79 to the tropics. Hwang et al. (2017) found that enhanced Southern Ocean heat uptake
 80 in a slab-ocean model could drive a tropical La Niña-like SST response via changing the
 81 zonal-mean atmospheric heat transport. This Southern Ocean-to-tropics teleconnection
 82 has further been supported by a variety of models with anomalous zonal-mean heat fluxes
 83 in the Southern Ocean (Kang et al., 2020). More recently, Dong et al. (2022) propose
 84 a two-way atmospheric pathway associated with regional atmospheric circulation instead
 85 of zonal-mean heat transport, linking the observed cooling in the tropical eastern Pa-
 86 cific and the southeast Pacific sector of the Southern Ocean. Kim et al. (2022) also finds
 87 that the inter-model spread in the teleconnection between these two regions is largely
 88 determined by differences in the subtropical cloud feedback across models. All these stud-
 89 ies suggest that the observed tropical eastern Pacific cooling may be remotely linked to
 90 the observed Southern Ocean cooling, which itself could be a direct response to Antarc-
 91 tic meltwater input.

92 Antarctic meltwater has been found to reduce projected global warming rates as
 93 well (Bronselaer et al., 2018; Schloesser et al., 2019; Sadai et al., 2020). From the stan-
 94 dard model of global energy balance (Raper et al., 2002; Gregory & Forster, 2008; Gre-
 95 gory et al., 2004, 2015):

$$N = \lambda T + F = \kappa T, \quad (1)$$

96 the global mean near-surface air temperature trend (dT/dt) can be approximated as:

$$dT/dt = \frac{dF/dt}{\kappa - \lambda}, \quad (2)$$

97 where N is the global-mean energy imbalance (unit: Wm^{-2}), F is the effective radiative
 98 forcing (ERF; unit: Wm^{-2}), κ is the ocean heat uptake (OHU) efficiency parameter (unit:
 99 $\text{Wm}^{-2}\text{K}^{-1}$), and λ is the radiative feedback parameter (unit: $\text{Wm}^{-2}\text{K}^{-1}$, negative for
 100 a stable climate). In this zero-layer energy balance model, κ and λ altogether determine
 101 the Earth’s surface temperature response to an ERF, with κ representing the efficiency
 102 with which heat is absorbed by the ocean and λ representing the efficiency with which
 103 heat is radiatively emitted to space at the top of atmosphere (TOA) per degree of global
 104 warming. The reduced dT/dt found in previous meltwater simulations has been commonly
 105 proposed to arise from an increased κ , as meltwater cools the ocean surface but warms
 106 at depth, making the Southern Ocean heat uptake more efficient (Gregory, 2000; Kirk-
 107 man & Bitz, 2011). However, it is also possible that meltwater reduces dT/dt by chang-
 108 ing λ through the SST pattern effect. Recent studies have found that an SST pattern
 109 with enhanced warming in the tropical western Pacific warm pool regions and cooling
 110 in other regions, as recently observed, tends to increase the lower-tropospheric stabil-
 111 ity and low-cloud cover globally, yielding a more-negative λ and therefore a lower effec-
 112 tive climate sensitivity (EffCS; Zhou et al., 2016; Andrews et al., 2018; Dong et al., 2019;
 113 Fueglistaler, 2019; Andrews et al., 2022). Given that Antarctic meltwater could produce
 114 surface cooling in both the Southern Ocean (due to increased ocean stratification) and
 115 the tropical eastern Pacific (due to teleconnections) – an SST pattern closer to the ob-
 116 served - it is possible that some portion of the reduced global warming rate is due to changes

117 in λ via SST pattern effects. A key question is: does Antarctic meltwater primarily in-
 118 fluence the warming rate dT/dt through changes in κ or λ ?

119 To better understand the impact of Antarctic meltwater input on transient and near-
 120 equilibrium global warming, this study aims to quantify (i) changes in κ and λ caused
 121 by Antarctic meltwater input and (ii) the respective impacts of κ changes and λ changes
 122 on the global warming rate. To do that, we employ two sets of published simulations with
 123 additional Antarctic meltwater imbalance, the so called “hosing” simulations. One is fo-
 124 cused on the recent historical period (leveraging simulations performed by Pauling et al.
 125 2016); the other is focused on the 21st century (leveraging simulations performed by Sadai
 126 et al. 2020). Both studies have previously examined the local response to the imposed
 127 meltwater forcing, showing an increased Antarctic sea ice and Southern Ocean surface
 128 cooling response consistent with other studies. Here we revisit these simulations, focus-
 129 ing on the response of global SST patterns and the global energy budget.

130 2 Methods

131 2.1 CESM1 meltwater simulations

132 We analyze two sets of meltwater hosing simulations performed using the fully-coupled
 133 CESM1-CAM5 (Neale et al., 2010). The first set (from Pauling et al. 2016) spans the
 134 historical period from 1980 to 2013. We hereafter refer to these as the “Historical Hos-
 135 ing” runs, though the simulations apply transient historical radiative forcing until 2005
 136 and Representative Concentration Pathway (RCP) 8.5 forcing thereafter to be consis-
 137 tent with CESM1 Large Ensemble (LENS) simulations (Kay et al., 2015). This ensem-
 138 ble consists of six members (we have performed four more since the publication of Paul-
 139 ing et al. 2016 following the same setup). Each of the six members has identical radi-
 140 ative forcing and anomalous meltwater forcing, but they are branched from a different LENS
 141 ensemble member. The anomalous freshwater input is added at a constant rate of 2000Gt/yr
 142 throughout the simulations, distributed at the front of ice shelves around Antarctica to
 143 mimic their basal melt (see Fig. 3b of Pauling et al. 2016 for the imposed freshwater dis-
 144 tribution, with the ice shelf location derived from the RTopo-1 dataset). Note that the
 145 amount of imposed freshwater input is chosen as needed to cause significant change in
 146 the annual-mean sea ice area for CESM1 (Pauling et al. 2016) and Southern Ocean sur-
 147 face temperature within CESM1, which is much larger than the observational estimate
 148 of 350 ± 100 Gt/yr (Rye et al., 2014), a caveat we will come back to in the discussion sec-
 149 tion. We present results from the ensemble-mean of the six meltwater runs, and com-
 150 pare changes relative to the ensemble-mean of 40 LENS runs that have no additional Antarc-
 151 tic meltwater (results remain the same if use the mean of the six LENS members from
 152 which the hosing runs were branched).

153 The other set of simulations (from Sadai et al. 2020) spans the 21st century from
 154 2006 to 2100. We hereafter refer to these as “Future Hosing” runs. This ensemble has
 155 one control run and one meltwater hosing run; both are forced by RCP8.5 transient forc-
 156 ing. Although a single ensemble member, the Future Hosing run includes a large fresh-
 157 water forcing, estimated by an offline ice sheet model forced by RCP8.5. The total amount
 158 of Antarctic freshwater input in the control run (from increasing precipitation only) stays
 159 around 0.1 Sv throughout the 21st century, whereas that in the hosing run (accounting
 160 for ice-sheet melting) reaches ~ 1 Sv in 2100. Note that the total freshwater forcing ap-
 161 plied in this simulation includes both liquid meltwater and solid ice (Fig. S1a) to account
 162 for the latent heat of melting (Fig. S1b), while in the Historical ensemble latent heat is
 163 not included. We take the difference between the control run and the meltwater run as
 164 the effect of Antarctic meltwater input in this ensemble.

165

2.2 Global energy budget analysis

166

167

168

169

170

171

172

173

174

175

176

177

178

179

180

181

182

183

184

185

For both Historical and Future hosing ensembles, we calculate the OHU efficiency κ and the net radiative feedback λ following the conventional global-energy budget framework expressed in Eq. (1) (Gregory et al., 2004, 2015), where $\kappa = dN/dT$, and $\lambda = d(N-F)/dT$. Note that N is taken as the global-mean net TOA radiation imbalance, which is in general equal to the global-mean OHU therefore can be used to calculate κ . One exception is the Future Hosing run, which includes the latent heat taken from the ocean to melt the solid ice. These additional heat fluxes can be accounted for by subtracting the latent heat (LH) needed to melt all the imposed solid ice from the net TOA radiation imbalance (Fig. S1), and we then calculate κ as $d(N-LH)/dT$ for this simulation. The latent heat of ice melt amounts to approximately 10% of the total global energy imbalance, and thus has a small effect on κ . Importantly, it does not influence the calculation of λ which depends only on TOA radiation. For the calculations of κ and λ , we choose to use the regression forms (over 1980–2013 for Historical Hosing and 2006–2100 for Future Hosing) since our focus is on transient warming. We also calculate the corresponding EffCS values ($= -F_{2x}/\lambda$, where F_{2x} is the radiative forcing of CO₂ doubling in CESM1), using the estimate of $F_{2x} = 3.88 \text{ Wm}^{-2}$ from Mitevski et al. (2021). EffCS here indicates an estimate of equilibrium climate sensitivity (ECS) using λ from a transient state, under the assumption that λ stays constant to equilibrium; it should be distinguished from the long-term Earth system sensitivity that involves changes in ice sheets operating on millennium timescales (Knutti et al., 2017).

186

187

188

189

190

191

192

193

194

195

Given that the effective radiative forcing (F) corresponding to these two ensembles is not explicitly available from CESM1 model output, we perform an additional fixed-SST simulation to estimate ERF following CMIP6 Radiative Forcing Model Intercomparison Project (RFMIP) protocol (Pincus et al., 2016). That is, we carry out a simulation using the atmospheric component of CESM1 (i.e., CAM5), with the same transient historical forcing for 1980–2013 and RCP8.5 forcing for 2006–2100 as used in the coupled CESM1 simulations, while fixing SST and sea-ice concentration at their preindustrial levels. ERF is then estimated as the TOA radiation imbalance from this fixed-SST simulation (Pincus et al., 2016; Dong et al., 2021). All variables (F , N , T) used in this study are annual means.

196

3 Results

197

198

3.1 Local and remote temperature responses to Antarctic meltwater input

199

200

201

202

203

204

205

206

207

208

209

210

211

We begin by analyzing the local response of Southern Ocean zonal-mean temperature and salinity trends (Fig. 1). In both ensembles, Antarctic meltwater input causes anomalous surface cooling, subsurface warming, and surface freshening in the Southern Ocean (Fig. 1, right column). The Future Hosing run produces stronger responses because (1) it imposes a larger amount of meltwater input and (2) it includes the effect of the latent heat of melting (which is not in the Historical ensemble). These local responses are qualitatively consistent with other studies (Bintanja et al., 2013; Bronselaer et al., 2018; Rye et al., 2020), reflecting an increase in upper ocean stratification and a decrease in upward heat transport. Moreover, accounting for meltwater produces historical temperature and salinity trends closer to observations (Fig. S2) than those simulated without meltwater input. This suggests that model-observation discrepancies in the Southern Ocean may partly arise from the models lack of ability to simulate additional meltwater due to Antarctic mass imbalance.

212

213

214

Next we consider the remote SST response to Antarctic meltwater input (Fig. 2). In both ensembles, the meltwater-induced Southern Ocean SST cooling extends to lower latitudes in the Southern Hemisphere (Figs. 2b, c), with the largest cooling occurring

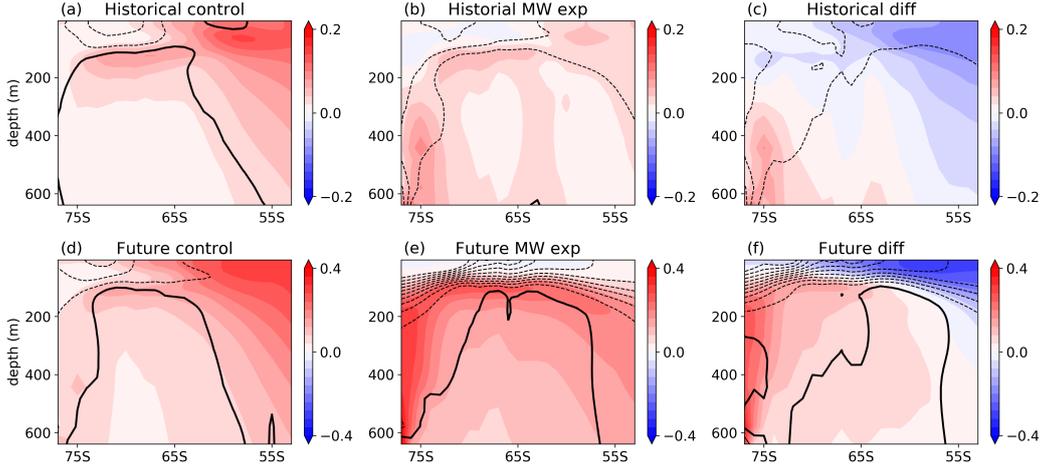


Figure 1. Zonal-mean ocean temperature trends (K/decade; shading) and salinity trends (g/kg/decade; contours) in the Southern Ocean. (a–c) Historical ensemble’s control, meltwater experiment, and the difference (experiment minus control). (d–f) Future ensemble’s control, meltwater experiment, and the difference. Contour interval in (a–d) is 0.01 g/kg/decade and in (e–g) is 0.02 g/kg/decade. Dashed contours denote negative anomalies; zero contours are thickened in all panels. Trends are calculated over 1980–2013 for (a–d) and 2006–2100 for (f–h).

215 in the eastern Pacific resulting in a La Niña-like tropical SST trend pattern (Figs. 2e,
 216 h). Notably, the Historical meltwater runs with all radiative forcings produce net cool-
 217 ing trends in the southeast Pacific sector of the Southern Ocean and the tropical south-
 218 east Pacific (Fig. 2d) – the two regions where observations have shown pronounced cool-
 219 ing trends (Fig. 2a). These two regions have also been found to have the strongest tele-
 220 connection via an atmospheric pathway, involving Rossby-wave dynamics and subtropi-
 221 cal advection (Dong et al., 2022).

222 Finally, we consider how Antarctic meltwater influences the global warming rate
 223 (Fig. 3). In the Historical Hosing runs, the global-mean surface air temperature trend
 224 dT/dt (over 1980–2013) is reduced by 20%, and in the Future Hosing run dT/dt (over
 225 2006–2100) is reduced by 28% (Table 1). Moreover, the Historical Hosing runs produce
 226 a dT/dt of 0.16 K/decade, which is more in line with the observed trend of 0.17 K/decade
 227 from HadCRUT5 (Morice et al., 2021) than that simulated without meltwater input (dT/dt
 228 = 0.2 K/decade). This suggests that the lack of Antarctic meltwater in models may ex-
 229 plain some of the model biases in historical global-mean warming.

230 In summary, we find that Antarctic meltwater input in CESM1 causes local and
 231 remote climate changes that are consistent with previous studies. Accounting for anom-
 232 alous meltwater input qualitatively reduces model biases in the historical record and also
 233 changes the projected warming in the near future. In the following sections we seek to
 234 further understand the relative roles of OHU efficiency κ and radiative feedback λ changes
 235 in reducing the global warming rate under meltwater forcing.

236 3.2 The response of κ and λ to Antarctic meltwater input

237 Having shown the global surface temperature response to Antarctic meltwater in-
 238 put, next we quantify the changes in κ and λ , two quantities that determine the change
 239 in dT/dt in a zero-layer energy balance model (Eq. 2).

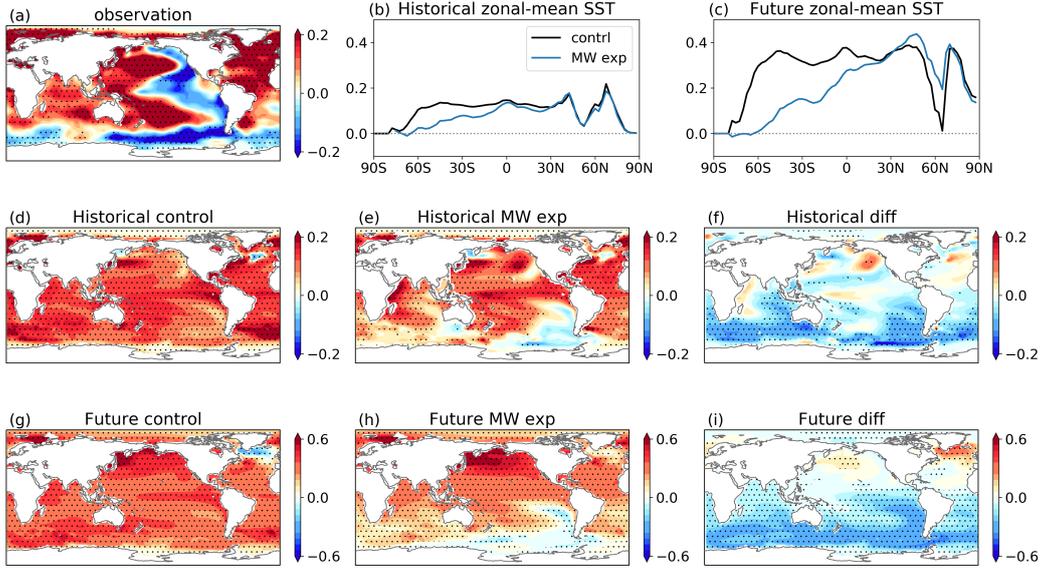


Figure 2. Global patterns of SST trends (K/decade), from (a) ERSSTv5 observation (Huang et al., 2017), (c–e) Historical control, meltwater experiment, and the difference, and (f–g) Future control, meltwater experiment, and the difference, respectively. (b–c) zonal-mean SST trends for the Historical and Future ensemble, respectively. SST trends are calculated over 1980–2013 for (a–e) and 2006–2100 for (f–h). Stippling indicates statistically significant linear trends at 95% level.

Table 1. Estimates of global-mean surface air temperature trend (dT/dt), κ , λ and EffCS for the simulations used in this study.

	Historical		Future	
	control	Meltwater	control	Meltwater
dT/dt [K/decade]	0.2	0.16	0.46	0.33
dT/dt estimated by Eq.2 [K/decade]	0.19	0.17	0.46	0.37
dT/dt_{κ} [K/decade]		0.18		0.43
dT/dt_{λ} [K/decade]		0.18		0.40
κ [$\text{Wm}^{-2}\text{K}^{-1}$]	1.15	1.28	0.5	0.62
λ [$\text{Wm}^{-2}\text{K}^{-1}$]	-1.1	-1.27	-1.01	-1.27
EffCS [K]	3.53	3.06	3.84	3.06

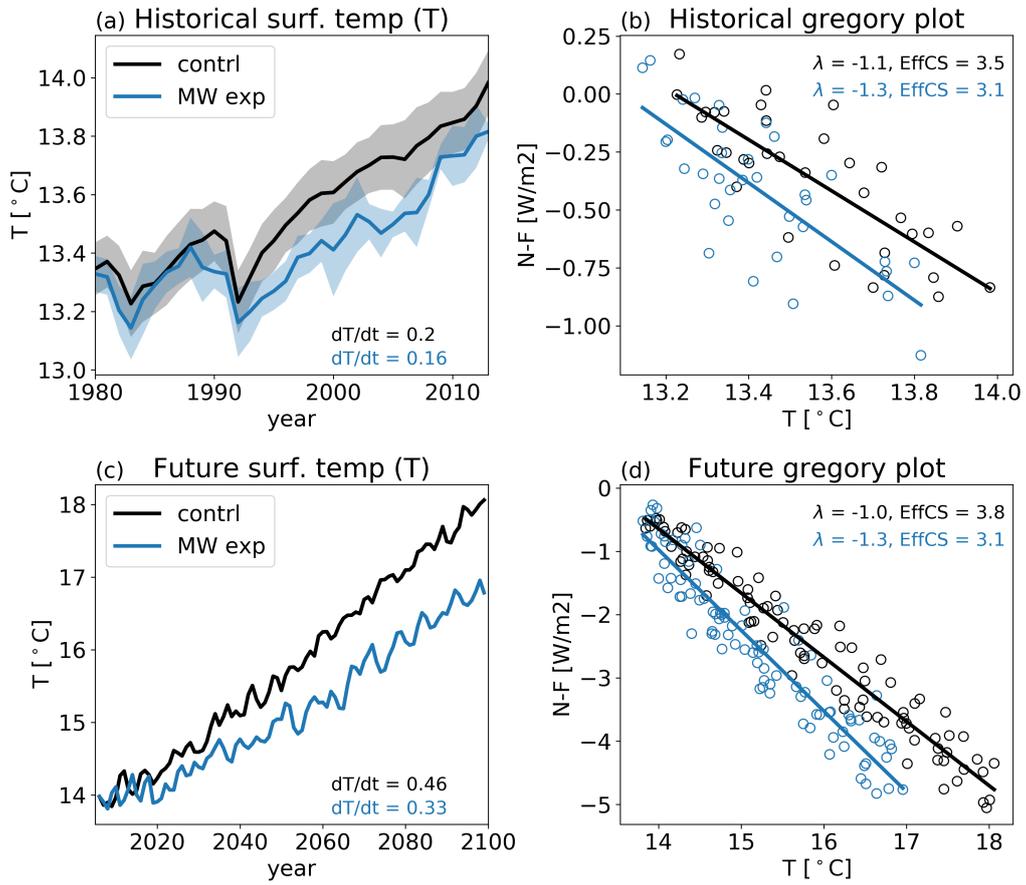


Figure 3. Ensemble-mean responses to meltwater input in (top) the Historical Hosing ensemble and (bottom) Future Hosing ensemble. (left) Global-mean surface air temperatures (T). (right) the Gregory plot (the net TOA radiative response $N-F$ against T). Black and blue denote results of control and meltwater runs respectively. Shading in (a) represents \pm one standard deviation across ensemble members.

240 We find that the OHU efficiency strengthens in response to Antarctic meltwater:
 241 κ increases by 11% in the Historical Hosing ensemble and by 24% in the Future Hosing
 242 run (Table 1). The strengthening of κ is not surprising given the changes in the verti-
 243 cal temperature distribution in the Southern Ocean shown in Fig. 1, characterized by
 244 anomalous surface cooling and subsurface warming, indicating more efficient OHU. The
 245 anomalous heat accumulation at depth has been proposed to principally arise from a re-
 246 duction in upward heat transport, which is a result of either decreased isopycnic tem-
 247 perature gradient (Gregory, 2000; Kirkman & Bitz, 2011) and/or decreased deep ocean
 248 convection (Russell & Rind, 1999; Bintanja et al., 2013), as the upper ocean becomes
 249 more stratified.

250 Meanwhile, we find that the net radiative feedback also becomes more stabilizing
 251 in the meltwater runs: the magnitude of λ increases (more negative λ) by 18% in the His-
 252 torical Hosing ensemble and by 30% in the Future Hosing run (Fig. 3, Table 1). Further-
 253 more, the more-stabilizing radiative feedbacks imply lower effective climate sensitivity:
 254 EffCS is reduced from 3.5 K to 3.1 K in the Historical Hosing ensemble and from 3.8 K
 255 to 3.1 K in the Future Hosing run, values that are closer to the EffCS estimate of 2K from
 256 atmospheric model simulations forced by the observed historical SSTs (Andrews et al.,
 257 2022). The changes in λ are primarily from the Southern Hemisphere subtropics (Fig.
 258 S3), associated with changes in the low-cloud feedback through the pattern effect (Rose
 259 et al., 2014; Rugenstein et al., 2016; Zhou et al., 2016; Dong et al., 2019). In the trop-
 260 ical and subtropical Pacific, the strengthened west-east SST gradient increases lower tro-
 261 pospheric stability, promoting more subtropical low clouds in the eastern Pacific stra-
 262 tocumulus deck (Wood & Bretherton, 2006; Zhou et al., 2016; Andrews et al., 2018; Dong
 263 et al., 2019). In the Southern Ocean, the meltwater-induced surface cooling locally yields
 264 a more-stable boundary layer, favoring high-coverage stratiform clouds (Dong et al., 2019;
 265 Atlas et al., 2020). In both regions, broad increases in low-cloud cover (Fig. S3) lead to
 266 stronger reflection of incoming shortwave radiation, and therefore a more-negative cloud
 267 feedback.

268 In summary, both κ and λ strengthen in response to Antarctic meltwater input in
 269 our simulations and thus both contribute to slowing the global warming rate. The stronger
 270 κ arises mostly from local changes in the depth of Southern Ocean heat storage. The stronger
 271 λ arises from both local (Southern Ocean) and remote (tropical) changes in cloud feed-
 272 backs owing to changes in the SST pattern.

273 3.3 Relative roles of κ and λ in changing the global warming rate

274 Finally, we come back to Eq. (2) to quantify the relative roles of κ and λ changes
 275 in reducing dT/dt . Although simplified, Eq. (2) provides an excellent approximation for
 276 the global warming rate (Gregory & Forster, 2008; Gregory et al., 2004; Andrews et al.,
 277 2022). Here, we also find that substituting values of dF/dt , κ and λ into Eq. (2) can ac-
 278 curately reproduce values of dT/dt from the corresponding simulations (Table 1). We
 279 can thus use the reconstructed values of dT/dt from Eq. (2), denoted as $dT/dt_{control}$ and
 280 dT/dt_{exp} for the control runs and the Hosing runs respectively, to quantify the relative
 281 contributions of changes in κ and λ .

282 To do so, we first estimate the value of dT/dt that would have occurred if only κ
 283 or λ changed while the other remained at the control level, denoted as dT/dt_{κ} or dT/dt_{λ} .
 284 We then calculate the change in these estimated dT/dt relative to $dT/dt_{control}$. Finally,
 285 we compare the changes in dT/dt due to changes in λ or κ alone to the total change in
 286 dT/dt (calculated as the difference between dT/dt_{exp} and $dT/dt_{control}$). We find that
 287 in the Historical Hosing ensemble, changes in κ and λ each account for approximately
 288 50% of the total change in dT/dt ; while in the Future Hosing run, κ change accounts for
 289 only 33% while λ change accounts for 67% of the total change in dT/dt (Table 1). The
 290 meltwater-induced reduction in global warming rate has long been thought to arise from

291 more efficient OHU in the Southern Ocean (a stronger κ), our results show that Antarc-
 292 tic meltwater input reduces the global warming rate via changes in both OHU efficiency
 293 and radiative feedbacks. Moreover, feedback changes can produce reductions in global
 294 warming rate that are comparable to or even greater than those produced by OHU ef-
 295 ficiency changes.

296 4 Conclusion and Discussion

297 Here we examined the impact of anomalous Antarctic meltwater on global warm-
 298 ing and the global energy budget in two sets of CESM1-CAM5 meltwater simulations.
 299 We find that the transient global warming rate is reduced by Antarctic meltwater input,
 300 owing to both a stronger OHU efficiency and a stronger radiative feedback (correspond-
 301 ing to a lower effective climate sensitivity). The strengthening in κ arises mostly from
 302 local changes in the depth of Southern Ocean heat storage, while the strengthening in
 303 λ arises from both local (Southern Ocean) and remote (tropical) SST changes that en-
 304 hanced negative cloud feedback through the SST pattern effect. Notably, accounting for
 305 anomalous Antarctic meltwater input produces a historical SST trend pattern better re-
 306 sembling observations than that simulated without meltwater input. The pattern effect-
 307 induced feedback changes contribute about equally to (in the Historical Hosing ensem-
 308 ble) or twice as much as (in the Future Hosing run) OHU efficiency changes in reduc-
 309 ing the global warming rate. These findings highlight a key role of Antarctic meltwater
 310 input in modulating regional and global climates that may have been missed in current
 311 GCMs.

312 Our results, which are based on the use of a single GCM, come with caveats. First,
 313 the amount of additional meltwater input needed to cause significant changes in a model
 314 has been found to be highly model dependent. Here we applied an amount of Antarc-
 315 tic meltwater (in the Historical Hosing ensemble) to CESM1 approximately 5 to 8 times
 316 higher than observational constraints. That said, although overestimating the observed
 317 meltwater amount, our meltwater runs still underestimate the observed surface cooling
 318 and freshening over the historical period (c.f. Fig. 1b and Fig. S2). This suggests that
 319 the local and remote response to the observed Southern Ocean surface freshening in na-
 320 ture may be even stronger than in our simulations. Second, the extent to which κ changes
 321 in response to Antarctic meltwater input may also be model dependent. For instance,
 322 it may depend on the Southern Ocean mean state, associated with model representation
 323 of deep ocean convection (Cabr e et al., 2017). Finally, while our results suggest a key
 324 role of the tropical SST pattern effect through teleconnection, the strength of the extratropical-
 325 to-tropical teleconnection appears to be also model dependent, with the inter-model spread
 326 largely coming from differences in the modeled subtropical cloud feedback (Kim et al.,
 327 2022). Thus, different results may arise from model differences in the Southern Ocean
 328 (e.g., representation of ocean mean states) and/or in the tropics (e.g., representation of
 329 atmospheric radiative feedbacks and teleconnection pathways). The robustness of our
 330 findings need to be tested in a range of models to better constrain the impact of Antarc-
 331 tic meltwater on global climate.

332 Despite these caveats, our results have important implications for understanding
 333 historical and future climate change. First, accounting for Antarctic meltwater in the
 334 Historical simulations reduces the global warming rate from 0.2 K/decade to 0.17 K/decade,
 335 which is more in line with the observational estimate of 0.16 K/decade (HadCRUT5).
 336 EffCS is also reduced from 3.5 K to 3.1 K, which is closer to the EffCS estimate of 2 K
 337 from atmospheric model simulations forced by the observed historical SST pattern (i.e.,
 338 AMIP simulations) and from observed energy budget constraints (Andrews et al., 2022).
 339 Second, with realistic meltwater input, the Future Hosing run projects a muted global
 340 warming over the coming century and a lower EffCS value than those simulated with-
 341 out meltwater input. If our results hold in other models, this finding suggests that the
 342 near-future warming projections by current GCMs may be overestimated. Furthermore,

343 many studies attribute the recent observed SST trend pattern (with cooling in the trop-
344 ical eastern Pacific and the Southern Ocean) to internal variability (e.g., the negative
345 phase of Inter-decadal Pacific Oscillation), and therefore hypothesize a reversed SST pat-
346 tern to appear in coming decades (Watanabe et al., 2021; Chung et al., 2022). However,
347 our simulations show that a similar historical SST pattern can arise with sufficient Antarc-
348 tic meltwater forcing and that this SST pattern can persist into the 21st century in the
349 presence of continued Antarctic meltwater input.

350 Additionally, several emergent constraints have been recently proposed linking model
351 simulated historical warming to the model’s equilibrium climate sensitivity (ECS). Some
352 studies find that models with higher ECS tend to overestimate the observed global-mean
353 warming rate over recent decades (Jiménez-de-la Cuesta & Mauritsen, 2019; Nijse et
354 al., 2020; Tokarska et al., 2020). Other studies find that even when models reproduce
355 the global-mean warming, the models with too positive cloud feedback and higher ECS
356 tend to produce less realistic interhemispheric asymmetry in surface temperatures (Wang
357 et al., 2021). Both suggest that lower ECS values from models that better reproduce his-
358 torical warming are more likely to happen. However, our simulations show that adding
359 anomalous Antarctic meltwater can reduce model biases by producing a lower global-
360 mean warming rate and an enhanced northern-southern hemispheric temperature asym-
361 metry, more in line with observations (Table S1). This suggests that model biases in the
362 transient historical warming may be (in part) due to the lack of realistic historical forc-
363 ing, not necessarily due to model biases in equilibrium response to CO₂ as those emer-
364 gent constraints suggested. Thus, high ECS on equilibrium timescales may be more re-
365 alistic than previously thought if Antarctic meltwater input has slowed the recent south-
366 ern hemispheric and global warming rates.

367 This work has shown a nontrivial impact of Antarctic meltwater input on climate
368 across spatial scales (both local and global) and time scales (both transient warming and
369 equilibrium climate sensitivity). Accurately projecting historical and future climate change
370 thus requires improved representation of realistic Antarctic meltwater input and its im-
371 pacts in GCMs.

372 Acknowledgments

373 YD was supported by the NOAA Climate and Global Change Postdoctoral Fellowship
374 Program, administered by UCAR’s Cooperative Programs for the Advancement of Earth
375 System Science (CPAESS) under award NA210AR4310383. KCA was supported by the
376 National Science Foundation (Grant AGS-1752796), the National Oceanic and Atmo-
377 spheric Administration MAPP Program (Award NA20OAR4310391), and an Alfred P.
378 Sloan Research Fellowship (Grant FG-2020-13568). We would like to acknowledge high-
379 performance computing support from Cheyenne (doi:10.5065/D6RX99HX) provided by
380 NCAR’s Computational and Information Systems Laboratory, sponsored by the National
381 Science Foundation.

382 Open Research

383 The Future Hosing simulations (first published in Sadai et al. 2020) are available
384 at <https://doi.org/10.15784/601449>. The Historical Hosing ensemble (first published in
385 Pauling et al. 2016) are available at <https://doi.org/10.5281/zenodo.7072848>. The CESM1
386 LENS simulations are obtained from [https://www.cesm.ucar.edu/projects/community-
387 projects/LENS/data-sets](https://www.cesm.ucar.edu/projects/community-projects/LENS/data-sets).

388 References

389 Andrews, T., Bodas-Salcedo, A., Gregory, J. M., Dong, Y., Armour, K. C., Paynter,
390 D., . . . others (2022). On the effect of historical SST patterns on radiative

- 391 feedback. *Journal of Geophysical Research: Atmospheres*, e2022JD036675.
- 392 Andrews, T., Gregory, J. M., Paynter, D., Silvers, L. G., Zhou, C., Mauritsen, T.,
 393 et al. (2018). Accounting for changing temperature patterns increases his-
 394 torical estimates of climate sensitivity. *Geophysical Research Letters*, *45*(16),
 395 8490–8499.
- 396 Armour, K. C., Marshall, J., Scott, J. R., Donohoe, A., & Newsom, E. R. (2016).
 397 Southern Ocean warming delayed by circumpolar upwelling and equatorward
 398 transport. *Nature Geoscience*, *9*(7), 549–554.
- 399 Atlas, R., Bretherton, C. S., Blossey, P. N., Gettelman, A., Bardeen, C., Lin, P.,
 400 & Ming, Y. (2020). How well do large-eddy simulations and global climate
 401 models represent observed boundary layer structures and low clouds over the
 402 summertime southern ocean? *Journal of Advances in Modeling Earth Systems*,
 403 *12*(11), e2020MS002205.
- 404 Bintanja, R., van Oldenborgh, G. J., Drijfhout, S., Wouters, B., & Katsman, C.
 405 (2013). Important role for ocean warming and increased ice-shelf melt in
 406 Antarctic sea-ice expansion. *Nature Geoscience*, *6*(5), 376–379.
- 407 Bronselaer, B., Winton, M., Griffies, S. M., Hurlin, W. J., Rodgers, K. B., Sergienko,
 408 O. V., . . . Russell, J. L. (2018). Change in future climate due to Antarctic
 409 meltwater. *Nature*, *564*(7734), 53–58.
- 410 Cabré, A., Marinov, I., & Gnanadesikan, A. (2017). Global atmospheric tele-
 411 connections and multidecadal climate oscillations driven by Southern Ocean
 412 convection. *Journal of Climate*, *30*(20), 8107–8126.
- 413 Chung, E.-S., Kim, S.-J., Timmermann, A., Ha, K.-J., Lee, S.-K., Stuecker, M. F., et
 414 al. (2022). Antarctic sea-ice expansion and Southern Ocean cooling linked to
 415 tropical variability. *Nature Climate Change*, *12*(5), 461–468.
- 416 De Lavergne, C., Palter, J. B., Galbraith, E. D., Bernardello, R., & Marinov, I.
 417 (2014). Cessation of deep convection in the open Southern Ocean under an-
 418 thropogenic climate change. *Nature Climate Change*, *4*(4), 278–282.
- 419 Dong, Y., Armour, K. C., Battisti, D. S., & Blanchard-Wrigglesworth, E. (2022).
 420 Two-way teleconnections between the Southern Ocean and the tropical Pacific
 421 via a dynamic feedback. *Journal of Climate*, 1–37.
- 422 Dong, Y., Armour, K. C., Proistosescu, C., Andrews, T., Battisti, D. S., Forster,
 423 P. M., et al. (2021). Biased estimates of equilibrium climate sensitivity and
 424 transient climate response derived from historical CMIP6 simulations. *Geo-*
 425 *physical Research Letters*, *48*(24), e2021GL095778.
- 426 Dong, Y., Proistosescu, C., Armour, K. C., & Battisti, D. S. (2019). Attribut-
 427 ing historical and future evolution of radiative feedbacks to regional warming
 428 patterns using a Green’s function approach: The preeminence of the western
 429 Pacific. *Journal of Climate*, *32*(17), 5471–5491.
- 430 Durack, P. J. (2015). Ocean salinity and the global water cycle. *Oceanography*,
 431 *28*(1), 20–31.
- 432 England, M. H., McGregor, S., Spence, P., Meehl, G. A., Timmermann, A., Cai, W.,
 433 et al. (2014). Recent intensification of wind-driven circulation in the Pacific
 434 and the ongoing warming hiatus. *Nature climate change*, *4*(3), 222–227.
- 435 Fan, T., Deser, C., & Schneider, D. P. (2014). Recent Antarctic sea ice trends in
 436 the context of Southern Ocean surface climate variations since 1950. *Geophys-*
 437 *ical Research Letters*, *41*(7), 2419–2426.
- 438 Fueglistaler, S. (2019). Observational evidence for two modes of coupling between
 439 sea surface temperatures, tropospheric temperature profile, and shortwave
 440 cloud radiative effect in the tropics. *Geophysical Research Letters*, *46*(16),
 441 9890–9898.
- 442 Gille, S. T. (2008). Decadal-scale temperature trends in the Southern Hemisphere
 443 ocean. *Journal of Climate*, *21*(18), 4749–4765.
- 444 Gregory, J. M. (2000). Vertical heat transports in the ocean and their effect on time-
 445 dependent climate change. *Climate Dynamics*, *16*(7), 501–515.

- 446 Gregory, J. M., Andrews, T., & Good, P. (2015). The inconstancy of the transient
 447 climate response parameter under increasing CO₂. *Philosophical Transactions*
 448 *of the Royal Society A: Mathematical, Physical and Engineering Sciences*,
 449 *373*(2054), 20140417.
- 450 Gregory, J. M., & Forster, P. (2008). Transient climate response estimated from ra-
 451 diative forcing and observed temperature change. *Journal of Geophysical Re-*
 452 *search: Atmospheres*, *113*(D23).
- 453 Gregory, J. M., Ingram, W. J., Palmer, M. A., Jones, G. S., Stott, P. A., Thorpe,
 454 R. B., et al. (2004). A new method for diagnosing radiative forcing and cli-
 455 mate sensitivity. *Geophysical research letters*, *31*(3).
- 456 Huang, B., Thorne, P. W., Banzon, V. F., Boyer, T., Chepurin, G., Lawrimore,
 457 J. H., ... Zhang, H.-M. (2017). Extended reconstructed sea surface tem-
 458 perature, version 5 (ERSSTv5): upgrades, validations, and intercomparisons.
 459 *Journal of Climate*, *30*(20), 8179–8205.
- 460 Hwang, Y.-T., Xie, S.-P., Deser, C., & Kang, S. M. (2017). Connecting tropical cli-
 461 mate change with southern ocean heat uptake. *Geophysical Research Letters*,
 462 *44*(18), 9449–9457.
- 463 Jiménez-de-la Cuesta, D., & Mauritsen, T. (2019). Emergent constraints on Earth’s
 464 transient and equilibrium response to doubled CO₂ from post-1970s global
 465 warming. *Nature Geoscience*, *12*(11), 902–905.
- 466 Kang, S. M., Xie, S.-P., Shin, Y., Kim, H., Hwang, Y.-T., Stuecker, M. F., et al.
 467 (2020). Walker circulation response to extratropical radiative forcing. *Science*
 468 *advances*, *6*(47), eabd3021.
- 469 Kay, J. E., Deser, C., Phillips, A., Mai, A., Hannay, C., Strand, G., ... others
 470 (2015). The Community Earth System Model (CESM) large ensemble project:
 471 A community resource for studying climate change in the presence of internal
 472 climate variability. *Bulletin of the American Meteorological Society*, *96*(8),
 473 1333–1349.
- 474 Kim, H., Kang, S. M., Kay, J. E., & Xie, S.-P. (2022). Subtropical clouds key to
 475 Southern Ocean teleconnections to the tropical pacific. *Proceedings of the Na-*
 476 *tional Academy of Sciences*, *119*(34), e2200514119.
- 477 Kirkman, C. H., & Bitz, C. M. (2011). The effect of the sea ice freshwater flux on
 478 Southern Ocean temperatures in CCSM3: Deep-ocean warming and delayed
 479 surface warming. *Journal of Climate*, *24*(9), 2224–2237.
- 480 Knutti, R., Rugenstein, M. A., & Hegerl, G. C. (2017). Beyond equilibrium climate
 481 sensitivity. *Nature Geoscience*, *10*(10), 727–736.
- 482 Kostov, Y., Ferreira, D., Armour, K. C., & Marshall, J. (2018). Contributions of
 483 greenhouse gas forcing and the southern annular mode to historical south-
 484 ern ocean surface temperature trends. *Geophysical Research Letters*, *45*(2),
 485 1086–1097.
- 486 Luo, J.-J., Wang, G., & Dommenges, D. (2018). May common model biases re-
 487 duce CMIP5’s ability to simulate the recent Pacific la niña-like cooling? *Cli-*
 488 *mate Dynamics*, *50*(3), 1335–1351.
- 489 Ma, H., & Wu, L. (2011). Global teleconnections in response to freshening over the
 490 Antarctic ocean. *Journal of climate*, *24*(4), 1071–1088.
- 491 Mitevski, I., Orbe, C., Chemke, R., Nazarenko, L., & Polvani, L. M. (2021). Non-
 492 monotonic response of the climate system to abrupt CO₂ forcing. *Geophysical*
 493 *research letters*, *48*(6), e2020GL090861.
- 494 Morice, C. P., Kennedy, J. J., Rayner, N. A., Winn, J., Hogan, E., Killick, R., ...
 495 Simpson, I. (2021). An updated assessment of near-surface temperature
 496 change from 1850: the HADCRUT5 data set. *Journal of Geophysical Research:*
 497 *Atmospheres*, *126*(3), e2019JD032361.
- 498 Neale, R. B., Chen, C.-C., Gettelman, A., Lauritzen, P. H., Park, S., Williamson,
 499 D. L., ... others (2010). Description of the NCAR community atmosphere
 500 model (CAM 5.0). *NCAR Tech. Note NCAR/TN-486+ STR*, *1*(1), 1–12.

- 501 Nijssen, F. J., Cox, P. M., & Williamson, M. S. (2020). Emergent constraints on tran-
 502 sient climate response (TCR) and equilibrium climate sensitivity (ECS) from
 503 historical warming in CMIP5 and CMIP6 models. *Earth System Dynamics*,
 504 *11*(3), 737–750.
- 505 Park, W., & Latif, M. (2019). Ensemble global warming simulations with idealized
 506 Antarctic meltwater input. *Climate Dynamics*, *52*(5), 3223–3239.
- 507 Parkinson, C. L. (2019). A 40-y record reveals gradual Antarctic sea ice increases
 508 followed by decreases at rates far exceeding the rates seen in the Arctic. *Pro-
 509 ceedings of the National Academy of Sciences*, *116*(29), 14414–14423.
- 510 Pauling, A. G., Bitz, C. M., Smith, I. J., & Langhorne, P. J. (2016). The response
 511 of the Southern Ocean and Antarctic sea ice to freshwater from ice shelves in
 512 an Earth system model. *Journal of Climate*, *29*(5), 1655–1672.
- 513 Pincus, R., Forster, P. M., & Stevens, B. (2016). The Radiative Forcing Model In-
 514 tercomparison Project (RFMIP): experimental protocol for CMIP6. *Geoscientific
 515 Model Development*, *9*(9), 3447–3460.
- 516 Purich, A., England, M. H., Cai, W., Sullivan, A., & Durack, P. J. (2018). Impacts
 517 of broad-scale surface freshening of the Southern Ocean in a coupled climate
 518 model. *Journal of Climate*, *31*(7), 2613–2632.
- 519 Raper, S. C., Gregory, J. M., & Stouffer, R. J. (2002). The role of climate sensitivity
 520 and ocean heat uptake on AOGCM transient temperature response. *Journal of
 521 Climate*, *15*(1), 124–130.
- 522 Roach, L. A., Dörr, J., Holmes, C. R., Massonnet, F., Blockley, E. W., Notz, D., ...
 523 others (2020). Antarctic sea ice area in CMIP6. *Geophysical Research Letters*,
 524 *47*(9), e2019GL086729.
- 525 Rose, B. E., Armour, K. C., Battisti, D. S., Feldl, N., & Koll, D. D. (2014). The de-
 526 pendence of transient climate sensitivity and radiative feedbacks on the spatial
 527 pattern of ocean heat uptake. *Geophysical Research Letters*, *41*(3), 1071–1078.
- 528 Rugenstein, M. A., Caldeira, K., & Knutti, R. (2016). Dependence of global radi-
 529 ative feedbacks on evolving patterns of surface heat fluxes. *Geophysical Research
 530 Letters*, *43*(18), 9877–9885.
- 531 Russell, G. L., & Rind, D. (1999). Response to CO₂ transient increase in the GISS
 532 coupled model: regional coolings in a warming climate. *Journal of Climate*,
 533 *12*(2), 531–539.
- 534 Rye, C. D., Marshall, J., Kelley, M., Russell, G., Nazarenko, L. S., Kostov, Y., ...
 535 Hansen, J. (2020). Antarctic glacial melt as a driver of recent Southern Ocean
 536 climate trends. *Geophysical Research Letters*, *47*(11), e2019GL086892.
- 537 Rye, C. D., Naveira Garabato, A. C., Holland, P. R., Meredith, M. P.,
 538 George Nurser, A., Hughes, C. W., et al. (2014). Rapid sea-level rise along
 539 the Antarctic margins in response to increased glacial discharge. *Nature Geo-
 540 science*, *7*(10), 732–735.
- 541 Sadai, S., Condrón, A., DeConto, R., & Pollard, D. (2020). Future climate re-
 542 sponse to Antarctic Ice Sheet melt caused by anthropogenic warming. *Science
 543 advances*, *6*(39), eaaz1169.
- 544 Schloesser, F., Friedrich, T., Timmermann, A., DeConto, R. M., & Pollard, D.
 545 (2019). Antarctic iceberg impacts on future Southern Hemisphere climate.
 546 *Nature Climate Change*, *9*(9), 672–677.
- 547 Seager, R., Henderson, N., & Cane, M. (2022). Persistent discrepancies between ob-
 548 served and modeled trends in the tropical Pacific Ocean. *Journal of Climate*,
 549 1–41.
- 550 Swart, N., & Fyfe, J. (2013). The influence of recent Antarctic ice sheet retreat
 551 on simulated sea ice area trends. *Geophysical Research Letters*, *40*(16), 4328–
 552 4332.
- 553 Tokarska, K. B., Stolpe, M. B., Sippel, S., Fischer, E. M., Smith, C. J., Lehner, F.,
 554 & Knutti, R. (2020). Past warming trend constrains future warming in CMIP6
 555 models. *Science advances*, *6*(12), eaaz9549.

- 556 Wang, C., Soden, B. J., Yang, W., & Vecchi, G. A. (2021). Compensation between
557 cloud feedback and aerosol-cloud interaction in CMIP6 models. *Geophysical re-*
558 *search letters*, *48*(4), e2020GL091024.
- 559 Watanabe, M., Dufresne, J.-L., Kosaka, Y., Mauritsen, T., & Tatebe, H. (2021).
560 Enhanced warming constrained by past trends in equatorial Pacific sea surface
561 temperature gradient. *Nature Climate Change*, *11*(1), 33–37.
- 562 Wills, R. C. J., Dong, Y., Proistosescu, C., Armour, K. C., & Battisti, D. S. (2022).
563 Systematic climate model biases in the large-scale patterns of recent sea-
564 surface temperature and sea-level pressure change. *Geophysical Research*
565 *Letters*, *49*(17), e2022GL100011.
- 566 Wood, R., & Bretherton, C. S. (2006). On the relationship between stratiform low
567 cloud cover and lower-tropospheric stability. *Journal of climate*, *19*(24), 6425–
568 6432.
- 569 Zhou, C., Zelinka, M. D., & Klein, S. A. (2016). Impact of decadal cloud variations
570 on the Earth’s energy budget. *Nature Geoscience*, *9*(12), 871–874.