## Is there a tropical response to recent observed Southern Ocean cooling?

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#### Abstract

Despite global warming, SSTs in the Southern Ocean (SO) have cooled in recent decades largely as a result of internal variability. The global impact of this cooling is assessed by nudging evolving SO SST anomalies to observations in an ensemble of coupled climate model simulations under historical radiative forcing, and comparing against a control ensemble. The most significant remote response to observed SO cooling is found in the tropical South Atlantic, where increased clouds and strengthened trade winds cool the sea surface, partially offsetting the radiatively-forced warming trend. The SO ensemble produces a more realistic tropical South Atlantic SST trend, and exhibits a higher pattern correlation with observed SST trends in the greater Atlantic basin, compared to the control ensemble. SO cooling also produces a significant increase in Antarctic sea ice, but not enough to offset radiatively-induced ice loss; thus, the SO ensemble remains biased in its sea ice trends.

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#### Key Points:

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7	•	The global impact of recent observed Southern Ocean surface cooling is studied
8		using coupled model experiments in a Pacemaker framework
9	•	Southern Ocean cooling induces significant cooling of the tropical South Atlantic
10		via increased cloudiness and strengthened trade winds
11	•	It also slows down the rate of radiatively-induced Antarctic sea ice loss, although
12		model disagreement with observed ice expansion remains

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#### 13 Abstract

Despite global warming, SSTs in the Southern Ocean (SO) have cooled in recent decades 14 largely as a result of internal variability. The global impact of this cooling is assessed 15 by nudging evolving SO SST anomalies to observations in an ensemble of coupled cli-16 mate model simulations under historical radiative forcing, and comparing against a con-17 trol ensemble. The most significant remote response to observed SO cooling is found in 18 the tropical South Atlantic, where increased clouds and strengthened trade winds cool 19 the sea surface, partially offsetting the radiatively-forced warming trend. The SO ensem-20 ble produces a more realistic tropical South Atlantic SST trend, and exhibits a higher 21 pattern correlation with observed SST trends in the greater Atlantic basin, compared 22 to the control ensemble. SO cooling also produces a significant increase in Antarctic sea 23 ice, but not enough to offset radiatively-induced ice loss; thus, the SO ensemble remains 24 biased in its sea ice trends. 25

#### <sup>26</sup> Plain Language Summary

Understanding how the observed pattern of global sea surface temperatures (SST) 27 changes come about remains a key objective in climate science. SSTs are expected to 28 rise as greenhouse gas concentrations increase. However, from 1979 to 2013, SSTs in the 29 Southern Ocean cooled because of natural climate variability, accompanied by Antarc-30 tic sea ice expansion. Yet this cooling and sea ice expansion are not generally captured 31 by climate models. In this study, we artificially incorporate the observed Southern Ocean 32 cooling in a climate model to see how it affects SSTs in other regions. We found that 33 in response to Southern Ocean cooling, the tropical South Atlantic SST cools and Antarc-34 tic sea ice expands, similar to observations. However, in simulations without the South-35 ern Ocean cooling, the Atlantic SST response look distinctly different, and Antarctic sea 36 ice retreats significantly. Our study suggests that realistic simulation of internal decadal 37 SO SST variability may be important for credible decadal SST projections in the trop-38 ical South Atlantic. 39

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#### 40 1 Introduction

Observed global sea surface temperature (SST) trends in recent decades show a dis-41 tinctive spatial pattern, with warming in the western Pacific, Indian Ocean, and North 42 Atlantic, and cooling in the eastern Pacific, South Atlantic, and Southern Ocean (SO, 43 Figure 1a). This pattern is reminiscent of both the Pacific Decadal Oscillation/Interdecadal 44 Pacific Oscillation (PDO/IPO) (Mantua et al., 1997; Power et al., 1999) and the Atlantic 45 Multidecadal Oscillation (AMO) (Kerr, 2000; Enfield et al., 2001), the dominant modes 46 of internal low frequency variability over the Pacific and Atlantic sectors, respectively. 47 The cooling over the SO has been partially attributed to internal variability associated 48 with changes in deep water formation (Latif et al., 2013; Cabré et al., 2017; Kostov et 49 al., 2018; L. Zhang et al., 2019), in addition to contributions from ozone depletion (Ferreira 50 et al., 2014) and melting of the Antarctic ice sheet (Bintanja et al., 2013; Bronselaer et 51 al., 2018). 52

While the role of the tropics in extra-tropical climate variability is well established 53 (Alexander et al., 2002; Deser et al., 2004; Kosaka & Xie, 2013; Newman et al., 2016), 54 the extra-tropics may also influence the tropics via coupled air-sea interactions. For ex-55 ample, midlatitude atmospheric variability can effectively provide stochastic forcing for 56 ENSO via the "seasonal footprinting mechanism" (Vimont et al., 2003; Alexander et al., 57 2010) and via the "meridional mode" (H. Zhang et al., 2014; Amaya et al., 2019); the 58 latter may also play a role in tropical Pacific decadal variability (Sun & Okumura, 2019; 59 Liguori & Di Lorenzo, 2019). 60

At high latitudes, projected sea ice loss in both hemispheres has been shown to im-61 pact the tropical Pacific via dynamical and thermodynamic air-sea interaction processes, 62 although the detailed mechanisms are not yet fully understood (Deser et al., 2015; K. Wang 63 et al., 2018; England et al., 2020). SO SST variability resulting from open-ocean con-64 vection in Weddell Sea has a significant impact on global energy balance redistribution, 65 affecting tropical SSTs and precipitation (Cabré et al., 2017). Idealized studies have ex-66 plored the effects of SO SST cooling, which often extends into the tropical southeast-67 ern Pacific and Atlantic (Hwang et al., 2017; Kang et al., 2019), reducing the warm SST 68 biases in those regions (Mechoso et al., 2016). Hwang et al. (2017) attributed the zonal 69 asymmetry in the tropical South Pacific SST response to wind-evaporation-SST (WES) 70 and shortwave cloud feedbacks, while ocean dynamics play an additional role in coupled 71

models (Kang et al., 2019). Here we show that the same processes contribute to the SST

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response in the tropical South Atlantic to the observed SO cooling.

Unlike the idealized SO studies cited above, which examined the equilibrium re-74 sponse to large amplitude perturbations, we investigate the transient response to a much 75 smaller and realistic perturbation, namely the observed cooling of SO SSTs in recent decades. 76 In particular, we address the following questions. Does observed SO cooling produce a 77 robust, remote climate response that extends into the tropics? If so, what is the nature 78 of the response pattern and its underlying mechanisms? To what extent is the observed 79 expansion of Antarctic sea ice controlled by SO SST cooling? To probe these questions, 80 we apply a "pacemaker" experimental protocol (Kosaka & Xie, 2013) by nudging SO SST 81 anomalies in a fully coupled global climate model to follow the observed evolution over 82 the period 1979–2013. This protocol has previously been applied to the tropical Pacific 83 to study the origins of the global surface warming hiatus (e.g., Kosaka & Xie, 2013; Deser, 84 Guo, & Lehner, 2017), as well as teleconnections from ENSO and the tropical lobe of 85 the PDO (i.e., Schneider & Deser, 2018; Deser, Simpson, et al., 2017), and to the North 86 Atlantic for assessing the global impact of observed Atlantic multidecadal variability (i.e., 87 Ruprich-Robert et al., 2016). To the best of our knowledge, we are the first to apply the 88 "pacemaker" protocol to investigate the influence of observed SO SST evolution on the 89 the global coupled climate system. The effects of observed SO cooling are compared with 90 those from observed tropical Pacific SST changes based on the Pacific Pacemaker sim-91 ulations (TPACE, Deser, Guo, & Lehner, 2017), as well as the radiatively-forced response 92 derived from the Community Earth System Model version 1 (CESM1) Large Ensemble 93 Project (LENS, Kay et al., 2015); note that all three sets of simulations employ the iden-94 tical model version for direct comparison. We describe our experimental design and data 95 in section 2, followed by results in section 3. We end with summary and discussion in 96 section 4. 97

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#### 2 Model and Experimental Design

We conduct a 20-member ensemble of SO "Pacemaker" Experiments (SOPACE) with CESM1.1.2 at 1° horizontal resolution, the same version used for LENS (Kay et al., 2015). The methodology follows Kosaka and Xie (2013) (also detailed in Deser, Guo, and Lehner (2017)). Briefly, SST anomalies (as opposed to absolute SSTs to avoid the model's mean state warm biases in the SO, C. Wang et al., 2014) at each grid box over

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the SO are nudged to the observed evolution of SST anomalies from 1975 to 2016 taken 104 from the NOAA Extended Reconstruction Sea Surface Temperature version 3b (ERSSTv3b) 105 data set (Smith et al., 2008). The nudging region covers ocean grids south of 40°S, with 106 a linearly tapering buffer zone from  $35^{\circ}$ S to  $40^{\circ}$ S. In regions with LENS climatological 107 sea ice cover, SST is nudged to the melting temperature of -1.8°C. The nudging timescale 108 is 2 days for the model's 10-m deep ocean surface layer, which is equivalent to the 10-109 day timescale for a 50-m deep mixed layer used in Kosaka and Xie (2013). Outside of 110 the nudging region, the model's coupled climate system evolves freely. All simulations 111 are subject to historical radiative forcing before 2005 and the RCP8.5 scenario thereafter, 112 following the CMIP5 protocols (Taylor et al., 2011). All SOPACE members are initial-113 ized from the first member of LENS on 1 Jan 1975, with a random initial atmospheric 114 temperature perturbation of  $\mathcal{O}(10^{-14})$  K to create ensemble spread. The surface heat 115 flux forcing used to nudge the model's SST anomalies shows significant spatial and tem-116 poral variations over the SO (not shown). The total energy perturbation in SOPACE 117 is approximately -0.1 PW, which is much less than that used in idealized experiments 118 cited earlier, for example, -0.8 PW in Kang et al. (2019). 119

We also analyze the LENS and a 20-member ensemble of the TPACE, also conducted 120 with CESM1.1.2 (Deser, Guo, & Lehner, 2017; Schneider & Deser, 2018). TPACE shares 121 the same experimental design as SOPACE, but with nudging in the tropical eastern Pa-122 cific  $(15^{\circ}N-15^{\circ}S; 80^{\circ}-180^{\circ}W)$  over the period 1920–2013. LENS consists of 40 members 123 that extend from 1920 to 2100, and shares the same historical and RCP8.5 forcing with 124 SOPACE and TPACE, but has no nudging toward observations. The LENS ensemble 125 mean (EM) is used to define the model's response to external forcing, and the spread 126 about the EM defines the model's internal variability. Observational data sets are de-127 scribed in supporting information. 128

#### 129 **3 Results**

#### <sup>130</sup> 3.1 SST Trends

We examine trends over the period 1979–2013 when the SO surface cooled, consistent with Schneider and Deser (2018). Figure 1a–d show the observed and EM SST trends from LENS, TPACE, and SOPACE. Over the Pacific, observations show a largescale pattern reminiscent of the negative phase of the PDO and IPO, with warming in

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the west and cooling in the east. In the Atlantic, positive SST trends in the sub-polar gyre and the tropical North Atlantic resemble the positive phase of the AMO. The Indian Ocean shows near-uniform surface warming. The SO cools overall, except for the Indian sector, with the strongest cooling in the Pacific sector north of the Amundsen and Bellingshausen Sea (Figure 1a).

The LENS-EM SST trend pattern, which represents the model's radiatively-forced 140 response, differs considerably from observations (Figure 1b). LENS-EM shows warming 141 around 0.1–0.2 K/decade over most of the global ocean, with the notable exception of 142 the subpolar North Atlantic, which cools as a result of a slowdown of the Atlantic Merid-143 ional Overturning Circulation (AMOC, Drijfhout et al., 2012). Enhanced equatorial warm-144 ing occurs in the Pacific and Atlantic, while muted warming is found in southeast and 145 northeast subtropical Pacific. Unlike observations, positive SST trends are evident through-146 out the SO (except along the Antarctic coastline), with enhanced warming in the At-147 lantic sector. This leads to the interpretation based on LENS that the recent observed 148 SO cooling is internally generated; however, potential biases in the model's forced response 149 and/or the lack of an interactive Antarctic ice sheet in CESM1 may affect this interpre-150 tation (e.g., Bronselaer et al., 2018). Unlike LENS-EM, TPACE-EM shows a negative 151 PDO/IPO pattern that bears a close resemblance to observations (Figure 1c). However, 152 the observed cooling over the SO is generally not simulated in TPACE-EM, except in 153 the eastern Pacific sector where weak cooling occurs. This indicates that while the trop-154 ical Pacific has some influence on the SO, it is not large enough to overwhelm the radiatively-155 forced response (see also Schneider & Deser, 2018). In the tropical Atlantic, SST trends 156 in TPACE-EM are generally of opposite sign compared to observations. Unlike TPACE-157 EM, SOPACE-EM shows a realistic pattern of SST trends in the tropical Atlantic, with 158 greater warming in the north compared to the south, although the amplitude of this dipole 159 is weaker than observed (Figure 1d). More importantly, the pattern of SST trends in the 160 tropical Atlantic in SOPACE-EM differs from the radiatively-forced response (LENS-161 EM), indicating a significant influence of SO SSTs in this region. SOPACE-EM also shows 162 greater cooling in the southeast subtropical Pacific compared to LENS-EM. 163

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To isolate the non-radiatively-forced responses in TPACE-EM and SOPACE-EM, we subtract the LENS-EM trends. We term this residual trend the internally-forced response (denoted "TPACE-internal" and "SOPACE-internal", Figure 1e and 1f), as it refers to the model's response to an imposed "forcing" that is internally-generated, as opposed

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to the externally-generated radiative forcing. In both TPACE-internal and SOPACEinternal, the SST trends are negative over much of the globe. In TPACE-internal, the
negative PDO/IPO pattern is generally preserved, and we see significant cooling in the
tropical North Atlantic, leading to a north-south tropical Atlantic SST gradient that is
opposite to observations. On the other hand, in SOPACE-internal, significant cooling
is found in the tropical South Atlantic with maximum amplitude along the west coast
of Africa.

In the North Atlantic, SOPACE-internal shows cooling in the subpolar gyre, co-175 located with the radiatively-forced "warming hole" in LENS-EM. Because of the con-176 nection between AMOC and the subpolar gyre SST (Rahmstorf et al., 2015), this cool-177 ing suggests that the observed SO cooling may also contribute to AMOC slowdown. Al-178 though the median of the SOPACE ensemble AMOC trend is lower than that of the LENS 179 ensemble (Figure S2), but not significantly so (p value = 0.3). Furthermore, caution 180 is needed in interpreting the deep ocean circulation response in SOPACE due to the ex-181 perimental protocol: SOPACE-internal exhibits significant subsurface cooling in the SO 182 down to  $\sim 300$  m depth (not shown), contrary to the observed subsurface warming and 183 positive trend in SO heat uptake (Armour et al., 2016; Tung & Chen, 2018). 184

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#### 3.2 Tropical South Atlantic Response

The most significant remote response in SOPACE-internal is the SST cooling in 186 the tropical South Atlantic (Figure 1f and 2a). This cooling extends all the way to the 187 equator, with maximum amplitude in the tropical southeastern Atlantic. As mentioned 188 earlier, this SO-induced cooling of the tropical South Atlantic brings the pattern of SST 189 trends over the entire tropical Atlantic in SOPACE-EM into closer alignment with ob-190 servations. In particular, there is a positive north-south gradient in both observations 191 and SOPACE-EM, opposite to that in TPACE-EM and distinct from LENS-EM (Fig-192 ure 1). To explore the robustness of this feature, we show the ensemble distribution of 193 SST trends averaged over the tropical South Atlantic (TSA: 10–30°S, 20°W–10°E; gray 194 box in Figure 1f) in LENS, TPACE, and SOPACE (Figure 1g). The observed TSA trend 195 is weakly positive at 0.053 K/decade (horizontal gray line in Figure 1g), and lies within 196 the middle 50% of the SOPACE distribution, but only within the bottom 10% of LENS 197 and bottom 25% of TPACE. In addition, the TSA trend spread is half as large in SOPACE 198 compared to LENS and TPACE. In summary, it is more likely for SOPACE members 199

to have a TSA trend that is consistent with observations than it is for LENS or TPACE
members, and the distribution of TSA trend in SOPACE is more constrained than those
in either LENS or TPACE.

We have also compared the basin-wide patterns of Atlantic SST trends between 203 observations and the model ensembles over the domain  $20^{\circ}N-40^{\circ}S$ ,  $70^{\circ}W-15^{\circ}E$  (gray 204 box in Figure 1a). The pattern correlation (r) with observations is considerably higher 205 for SOPACE  $(r(obs, EM)_{SOPACE} = 0.64)$  than for either LENS  $(r(obs, EM)_{LENS} = 0.20)$ 206 or TPACE  $(r(obs, EM)_{TPACE} = -0.28)$ , consistent with visual impression from Fig-207 ure 1 (green stars in Figure 1h). This suggests that the internally-forced response to ob-208 served SO cooling makes a substantial contribution to the spatial pattern of observed 209 SST trends over the broad Atlantic region (and that the radiatively-forced response can-210 not sufficiently explain the observed SST trend pattern over the Atlantic). 211

In order to better quantify the spread due to internal variability, we also computed 212 the pattern correlation between each ensemble member i and the EM of each experiment 213 (denoted r(i, EM)). The range of  $r(i, \text{EM})_{\text{SOPACE}}$  spans from 0.41-0.81, with  $r(\text{obs}, \text{EM})_{\text{SOPACE}}$ 214 lying in the center of the distribution. This indicates that the inclusion of observed in-215 ternal SO cooling results in a more realistic pattern of simulated SST trends over the 216 broad  $(20^{\circ}N-40^{\circ}S)$  Atlantic domain. Moreover, the observed SST trend pattern in this 217 region resembles the simulated response to SO cooling. The distribution of  $r(i, \text{EM})_{\text{TPACE}}$ , 218 on the other hand, does not encompass the negative  $r(obs, EM)_{TPACE}$ , further empha-219 sizing the inability of the observed tropical eastern Pacific cooling to produce the observed 220 pattern of Atlantic SST trends (Figure 1h). While the distribution of  $r(i, \text{EM})_{\text{LENS}}$  also 221 encompasses the observed pattern correlation,  $r(obs, EM)_{LENS}$  is much lower than  $r(obs, EM)_{SOPACE}$ . 222 Furthermore, in the LENS ensemble, the influence of the model's internal SO SST trends 223 on the Atlantic pattern correlations cannot be isolated. 224

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#### 3.3 Ocean Mixed Layer Heat Budget Analysis

What processes contribute to the remote Atlantic SST response in SOPACE? To answer this question, we analyze the heat budget of the upper ocean mixed layer following Xie et al. (2010),

$$\rho c_p H \frac{\partial T_s}{\partial t} = F_{\rm SW} + F_{\rm LW} + SH + LH + O, \tag{1}$$

where  $\rho$  is the density of ocean,  $c_p$  is the specific heat of ocean, H is the ocean mixed-229 layer depth,  $T_s$  is the mixed-layer temperature. Hence the left-hand side (LHS) repre-230 sents the mixed-layer heat storage term. The right-hand side (RHS) consists of net sur-231 face shortwave  $(F_{\rm SW})$  and longwave  $(F_{\rm LW})$  fluxes, sensible (SH) and latent (LH) heat 232 fluxes, and heat flux due to ocean dynamics (O). Our convention is that positive val-233 ues on the RHS warm  $T_s$  and vice versa. We compute the ensemble-mean trend (1979-234 2013) of each term in Equation 1 for SOPACE-internal in order to isolate the forced re-235 sponse to the internal component of observed SO SST trends. Maps of the trends in each 236 quantity (denoted by the superscript t, e.g.,  $F_{SW}^t$ ) are shown in Figure S4 for the trop-237 ical eastern Pacific and Atlantic domain. Heat storage trends are negligible (Figure S4b), 238 indicating that the trends of the RHS terms of Equation 1 are in quasi-equilibrium. A 239 similar result was shown by Cook et al. (2018) based on ocean reanalysis products. This 240 allows us to compute  $O^t$  as a residual term (Equation S2). 241

The dependency of latent heat flux on SST allows us to rewrite Equation 1 as a diagnostic equation of the SST trend  $T_s^t$ , following Jia and Wu (2013) and Hwang et al. (2017) (see derivation in supporting information):

$$T_{s}^{t} = T_{\rm SW}^{t} + T_{\rm LW}^{t} + T_{\rm SH}^{t} + T_{\rm O}^{t} + T_{\rm LH,w}^{t} + T_{\rm LH,RH}^{t} + T_{\rm LH,\Delta T}^{t}.$$
 (2)

The RHS in Equation 2 represents the contributions to SST trend from  $F_{SW}$ ,  $F_{SW}$ , SH, 245 O, and LH. The latent heat term can be further broken down into contributions from 246 trends in near-surface wind  $T_{\rm LH,w}^t$ , near-surface relative humidity  $T_{\rm LH,RH}^t$ , and air-sea tem-247 perature gradient  $T_{\text{LH},\Delta T}^t$ . The sum of the terms on the RHS of Equation 2 (Figure 2a) 248 closely approximates the actual  $T_s^t$  (Figure 1f), validating our methodology. The major 249 terms contributing to cooling trends in both basins include  $T_{SW}^t$  (due to increased cloud 250 liquid water, Figure 2b),  $T_{\rm O}^t$  (largely due to Ekman advection, Figure 2e), and  $T_{\rm LH,w}^t$  (due 251 to strengthened trade winds, Figure 2f). 252

In the Atlantic, the contribution from  $F_{\rm SW}$  is centered at the coastal stratocumulus region between 10–15°S. A positive trend in liquid water path is expected as the SST cools and strengthens the cloud top temperature inversion. The equatorial Atlantic cooling is dominated by the WES feedback, where there is a strengthening of the southeasterly trade winds, which enhances evaporative cooling. At 25°S, the strengthened winds not only cool the SST via WES feedback, but also via Ekman advection (Figure S6). In the Eastern Pacific,  $F_{\rm SW}$  plays a more dominant role in cooling the SST than in the Atlantic, due to a weaker near-surface wind response and associated LH. To summarize, the pattern of the negative  $T_s^t$  is caused by the sum of contributing terms instead of dominated by a single heat flux trend.

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#### 3.4 Antarctic Sea Ice Trends

Observational evidence suggests that multidecadal SST and wind variability over 264 the SO plays an essential role in governing the pattern of Antarctic sea ice trends (Fan 265 et al., 2014; Armour & Bitz, 2015). Can SOPACE reproduce the observed trends in Antarc-266 tic sea ice? The observed Antarctic sea ice extent (SIE) trend during this period is  $0.23 \times$ 267  $10^6 \text{ km}^2$  per decade, which is drastically different from the radiatively-forced SIE trend 268 of  $-0.36 \times 10^6$  km<sup>2</sup> per decade in LENS-EM and  $-0.30 \times 10^6$  km<sup>2</sup> per decade in TPACE-269 EM (Figure 3a). In contrast, SOPACE-EM shows an Antarctic SIE trend of  $-0.040 \times$ 270  $10^{6}$  km<sup>2</sup> per decade, a rate that is an order of magnitude slower than in LENS-EM and 271 TPACE-EM. Furthermore, 40% of SOPACE ensemble members show a positive SIE trend, 272 although none is as large as observed (the largest trend in SOPACE is  $0.084 \times 10^6 \text{ km}^2$ 273 per decade; Figure 3a). The significant differences between Antarctic SIE trends in SOPACE-274 EM, LENS-EM, and TPACE-EM indicate that observed SO SST cooling plays an im-275 port role in influencing Antarctic SIE, though it is not the only factor. 276

In addition to differences in the total SIE trend, there are significant differences 277 in the patterns of sea ice concentration (SIC) trends among observations and the model 278 ensembles. The observed SIC shows positive trends over most of the SO, except for the 279 West Antarctic coastline and north of the Weddell Sea (Figure 3c). LENS-EM shows a 280 nearly zonally-symmetric negative SIC trend pattern (Figure 3d), similar to that in TPACE-281 EM although with weaker magnitude in the Atlantic and Pacific sectors (Figure 3e; see 282 also Figure 3g). On the other hand, SIC trends in SOPACE-EM show a mixture of pos-283 itive and negative values, with ice gain in the Bellingshausen Sea and ice loss in the Wed-284 dell Sea (Figure 3f). Indeed, the imposed SST cooling in the SO leads to a marked Antarctic-285 wide expansion of sea ice relative to the radiatively-forced response (Figure 3h). 286

Although observed SO cooling produces an Antarctic-wide increase in SIC, when combined with the radiatively-forced response, the SIC trend pattern in SOPACE-EM differs from observations. For example, SOPACE-EM produces sea ice gain in the Bellingshausen Sea and Drake Passage but ice loss in the Weddell Sea and Ross Sea, opposite

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to observations (Figure 3f and 3c). At first glance, this seems to contradict the obser-291 vational evidence that positive SIC trends are associated with SST cooling, and vice versa. 292 However, observations also suggest that near-surface winds can drive thermodynamic sea 293 ice changes via temperature advection. In general, anomalous northerly winds contribute 294 to SST increase and sea ice loss via warm air advection, while anomalous southerlies cor-295 respond to opposite conditions (Fan et al., 2014). Indeed, north of the Weddell Sea and 296 Antarctic Peninsula, ERA-5 shows northerly near-surface winds that are generally as-297 sociated with warm-air advection and sea ice loss. In the Ross Sea, southerly winds are 298 co-located with positive SIC trends (Figure 3c). In the Amundsen Sea, southerly wind 299 trends are associated with negative sea ice trends along the coast, consistent with the 300 finding that ice advection driven by local winds dominates the sea ice loss there (Holland 301 & Kwok, 2012). The near-surface wind trends in SOPACE-EM display significant pat-302 tern differences from observations, especially in the Amundsen Sea and the Ross Sea (Fig-303 ure 3f), which partially explain the differences in regional SIC trends between SOPACE-304 EM and observations. We note, however, that the highest pattern correlation between 305 SIC trends in any single member of SOPACE and observations is only 0.16. Furthermore, 306 the range of  $r(i, \text{EM})_{\text{SOPACE}}$  (0.38–0.91) does not encompase  $r(obs, \text{EM})_{\text{SOPACE}} = -0.11$ 307 (Figure 3b). Thus, our results suggest that specifying the observed SO SST trends in 308 CESM1 does not guarantee a match between the observed and simulated response of Antarc-309 tic SIC trends, and exposes model biases in both the radiatively-forced component as 310 well as patterns of internally-generated SIC trends in SOPACE. Processes such as ocean-311 ice shelf interaction are absent in CESM1, potentially contributing to biases, e.g., in the 312 Weddell Sea (Park & Latif, 2019). 313

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#### 4 Summary and Discussion

We have examined the remote impact of observed SO SST cooling during 1979– 2013 in CESM1 using a 20-member initial-condition ensemble in which SST anomalies over the SO are nudged to the observed monthly evolution (SOPACE). The results are compared to the 20-member TPACE, where the same protocol was applied to SSTs in the tropical eastern Pacific, as well as the freely evolving 40-member LENS. The forced response in each experiment is estimated by the EM.

The most significant influence of the observed SO cooling is found in the tropical South Atlantic. In this region, the observed SO cooling leads to a significant reduction

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in the radiatively-forced warming trend, enhancing the north-south gradient in tropical 323 Atlantic SST trends. The muted tropical Atlantic warming south of the equator com-324 pared to the north in SOPACE-EM resembles the observed SST trend pattern, which 325 neither TPACE-EM nor LENS-EM captures. Indeed, TPACE-EM shows an opposite trend 326 in the tropical Atlantic SST gradient. Furthermore, it is more likely for SOPACE mem-327 bers to have realistic tropical South Atlantic SST trends than either LENS or TPACE 328 members. The SOPACE ensemble also has a realistic depiction of the SO cooling's in-329 fluence on the broader tropical Atlantic SST trend pattern, unlike TPACE. The trop-330 ical South Atlantic response suggests an important role the SO SST plays on decadal 331 time scales, consistent with previous findings (Cabré et al., 2017). 332

A mixed layer heat budget analysis of the tropical South Atlantic region shows that 333 increased low-level cloud amount and enhanced trade winds contribute to reducing SSTs 334 via shortwave radiative flux and the WES-feedback mechanism, respectively. Ekman ad-335 vection also contributes to SST cooling along the African coast where strong climato-336 logical SST gradients exist. The observed trends of low-level cloud amount are signif-337 icantly positive over the southeastern Atlantic and Pacific during 1984–2009, consistent 338 with our findings based on SOPACE-internal (See thala et al., 2015). Multilinear regres-330 sion analysis suggests that strengthened inversion stability is the dominant cloud-controlling 340 factor for the observed positive low-cloud amount trends. The processes that contribute 341 to the SO induced cooling are consistent with the idealized studies (Hwang et al., 2017; 342 Kang et al., 2019), although the WES feedback is much weaker in the tropical eastern 343 Pacific in SOPACE-internal. 344

Antarctic SIE has increased during 1979–2013, opposite to the radiatively-forced 345 decrease in LENS-EM. While the SIE trend in TPACE-EM is similar to that in LENS-346 EM, SIE trend in SOPACE-EM is near zero, and 40% of SOPACE ensemble members 347 show positive SIE trends although none as large as observed. Furthermore, the observed 348 SO cooling leads to Antarctic-wide positive SIC trends in SOPACE-internal, but because 349 of differences in near-surface wind trends, the spatial pattern of the simulated SIC trends 350 differs from observations. The low pattern correlations between observed SIC trends and 351 those in the individual members of SOPACE, and the negative pattern correlation be-352 tween observed SIC trends and those in SOPACE-EM (which lies outside the distribu-353 tion of pattern correlations between individual members of SOPACE and SOPACE-EM), 354

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indicates a likely bias in the amplitude and pattern of the model's radiatively-forced SIC
 trends.

Our results may have implications for understanding the future trajectory of Antarctic sea ice and tropical South Atlantic SSTs. Indeed, SSTs have recently warmed in the SO while Antarctic SIE reached record low values in 2017 (Parkinson, 2019; Meehl et al., 2019). If this SO SST warming trend continues, the tropical South Atlantic may experience accelerated warming due to the combined influence from SO SSTs and increasing greenhouse gas concentrations. Future work is needed to investigate the predictability of these impacts relative to other sources of internal variability in the coming decades.

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Figure 1. Monthly SST trends over 1979–2013 for (a) ERSSTv3b, (b) LENS-EM, (c) TPACE-EM, (d) SOPACE-EM, (e) TPACE-internal, and (f) SOPACE-internal. Stippling indicates local trend that is significant at or above 95% level. Black lines outline the SST nudging domain. (g) Box-and-whisker plot of SST trends averaged over the tropical South Atlantic (gray box in (f)) for each model ensemble. Green stars show the EM values. Orange lines show the median values. Gray horizontal line shows the observed value. (h) Same as (g) but for pattern correlations of SST trends with EM r(i, EM) over the broader Atlantic region (gray box in (a)). Green triangles show r(obs, EM).



Figure 2. SST trend decomposition based on Equation 2 using monthly trends during 1979–2013 from SOPACE-internal over the tropical eastern Pacific and Atlantic: (a) net  $T_s^t$ , (b)  $T_{SW}^t$ , (c)  $T_{LW}^t$ , (d)  $T_{SH}^t$ , (e)  $T_O^t$ , (f)  $T_{LH,W}^t$  with near-surface wind trends overlaid (vectors), (g)  $T_{LH,RH}^t$ , and (h)  $T_{LH,\Delta T}^t$ .



Figure 3. (a) Box-and-whisker plot of Antarctic SIE trends for each model ensemble. Green stars show the EM values. Orange lines show the median values. Gray horizontal line shows the observed value. (b) Same as (a) but for pattern correlations of SIC trends with EM r(i, EM) over 50–80°S. Green triangles show r(obs, EM). Monthly SST (colors), SIC (contours), and near-surface wind (vectors, m/s/decade) trends over 1979–2013 for (a) ERSSTv3b, (b) LENS-EM, (c) TPACE-EM, (d) SOPACE-EM, (e) TPACE-internal, and (f) SOPACE-internal. Blue (positive) and red (negative) contours outline regions with SIC trend magnitudes greater than 0.5%/decade.

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Figure 1.



Figure 2.















(d) SH









80°W

40°W



Figure 3.



- 0.3









# - 0.1 (**yqqc**) - 0.0 (**yqqc**) - -0.1 - -0.2

# (h) SOPACE-internal \_\_\_\_\_



## **@AGU**PUBLICATIONS

#### Geophysical Research Letters

Supporting Information for

#### Is there a tropical response to recent observed Southern Ocean cooling?

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#### Contents of this file

Text S1 Figures S1 to S6

#### Introduction

This supporting information provides description of our observational data sets, analysis methods, mixed layer budget equations and figures that are complementary to the main article.

#### Text S1.

#### **Observational Data Sets**

We use a suite of data sets to capture the observed trends during 1979-2013. SSTs are taken from the NOAA ERSSTv3b data set with 2° global resolution (Smith et al., 2008), for compatibility with those used in TPACE and SOPACE. We also show the SO SST anomaly timeseries from ERSSTv5 data set for comparison (Huang et al., 2017). The two data sets don't show significant differences in their SO SST trends (Figure S1). Surface wind data are taken from ERA-5 Reanalysis at 0.25° global resolution (Hersbach et al., 2019). Sea ice concentration data are from the passive-microwave-derived NASA Goddard Bootstrap version 2 dataset on a 25 km x 25 km grid (Peng et al., 2013).

#### Analysis Methods

We calculate monthly anomalies of all fields by subtracting the climatological monthly means of the 1981-2010 base period. Linear least-square regression is used to compute monthly trends during 1979-2013. We use stippling to indicate regions of statistical significance at 95% confidence level, based on a two-sided *t*-test method adjusted for autocorrelation (Santer et al., 2000; Schneider & Deser, 2018).

#### **Ocean Mixed Layer Budget Equations**

We focus on the ensemble-mean ocean mixed layer budget in SOPACE-internal, in order to isolate the forced response to the internal component of observed SO SST trends. The surface mixed-layer budget equation is:

$$\rho c_p H \frac{\partial T_s}{\partial t} = F_{\rm SW} + F_{\rm LW} + SH + LH + 0.$$
(S1)

We can take the linear trend of Equation S1, the left-hand side which represents the trend of heat storage is negligible (Cook et al. 2018, also see Figure S3b), we will have

$$0 = F^{t}{}_{SW} + F^{t}{}_{LW} + SH^{t} + LH^{t} + O^{t},$$
(S2)

where the superscript *t* denotes *linear trends* during 1979-2013.

Figure S3 shows all terms in Equation S3. The sensible heat flux trend is the smallest term. The rest of the terms can be considered forcing on SST trend  $T_s^t$  due to clouds  $(F_{SW}^t)$ , ocean dynamics  $(O^t)$ , and atmospheric temperature and humidity  $(F_{LW}^t)$ . The latent heat flux term directly depends on  $T_s$  via saturation vapor pressure:

$$LH = -L_v c_E \rho_a W[q_s(T_s) - q_a].$$
(S3)

And  $q_a$  is the specific humidity of air above the sea surface, thus can be written as

$$q_{\rm a} = RH_0 q_s (T_{\rm s} + \Delta T), \tag{S4}$$

where  $RH_0$  is the relative humidity at the sea surface, and  $\Delta T = T_a - T_s$  is the temperature gradient near the sea surface. Using the Clausius-Clapeyron equation, Equation S4 can be written as

$$q_{\rm a} = RH_0 q_{\rm s}(T_{\rm s}) e^{\alpha \Delta T},\tag{S5}$$

Where  $\alpha = \frac{L_{\rm v}}{R_{\rm v}T^2} \approx 0.06~{\rm K}^{-1}$ . We can plug Equation S5 into S3 and get

$$LH = -L_{\rm v}c_{\rm E}\rho_{\rm a}W(1-RH_0e^{\alpha\Delta T})q_s(T_{\rm s})$$

Following Hwang et al. (2017) and Jia & Wu (2013), the linear trend of latent heat flux can therefore be linearized as

$$LH^{t} = \frac{\partial LH}{\partial T_{s}} T_{s}^{t} + \frac{\partial LH}{\partial W} W^{t} + \frac{\partial LH}{\partial RH_{0}} RH_{0}^{t} + \frac{\partial LH}{\partial \Delta T} \Delta T^{t}$$
(S6)

On the right-hand side (RHS), the last three terms can also be considered as forcing from the atmosphere due to changes in near-surface wind speed, near-surface relative humidity, and air-sea temperature gradient:

$$LH_{W}^{t} = \frac{\partial LH}{\partial W}W^{t} = \overline{LH}\frac{W^{t}}{\overline{W}},$$
(S7)

$$LH_{\rm RH}^t = \frac{\partial LH}{\partial RH_0} RH_0^t = -\frac{\overline{LH}RH_0^t}{e^{\alpha \overline{\Delta T}} - \overline{RH_0}},$$
(S8)

$$LH_{\Delta T}^{t} = \frac{\partial LH}{\partial \Delta T} \Delta T^{t} = \frac{\alpha \overline{LH} \overline{RH_{0}} \Delta T^{t}}{e^{\alpha \overline{\Delta T}} - \overline{RH_{0}}},$$
(S9)

while the first term on RHS of Equation S6 becomes the SST damping term:

$$\frac{\partial LH}{\partial T_{\rm s}}T_{\rm s}^t = \alpha \overline{LH} T_{\rm s}^t, \tag{S10}$$

where  $\overline{LH}$  is the climatological mean latent heat flux, and it is negative when evaporation happens. This term in Equation S10 means that higher SST evaporates more to cool, therefore damps the temperature change. Equation S10 allows us to rewrite Equation S2 as a diagnostic equation of the SST trend:

$$T_{\rm s}^t = -\frac{F^t_{\rm SW} + F^t_{\rm LW} + SH^t + O^t + LH_{\rm W}^t + LH_{\rm RH}^t + LH_{\Delta \rm T}^t}{\alpha LH}.$$
(S11)

We can rewrite Equation S11 to show how different forcing terms contribute to the SST trends:

$$T_{\rm s}^{t} = T_{\rm SW}^{t} + T_{\rm LW}^{t} + T_{\rm SH}^{t} + T_{\rm O}^{t} + T_{\rm LH,w}^{t} + T_{\rm LH,RH}^{t} + T_{\rm LH,\Delta T}^{t},$$
(S12)

where the RHS terms can be defined using Equation S7–S9.

$$T_{\rm SW}^t = -\frac{F^t_{\rm SW}}{\alpha LH},\tag{S13}$$

$$T_{\rm LW}^t = -\frac{F_{\rm LW}^t}{\alpha LH},\tag{S14}$$

$$T_{\rm SH}^t = -\frac{F_{\rm SH}^t}{\alpha LH},\tag{S15}$$

$$T_{\rm O}^t = -\frac{F_{\rm O}^t}{\alpha L H'},\tag{S16}$$

$$T_{\rm LH,w}^t = -\frac{LH_w^t}{\alpha \overline{LH}} = -\frac{W^t}{\alpha \overline{W'}}$$
(S17)

$$T_{\rm LH,RH}^t = -\frac{LH_{\rm RH}^t}{\alpha LH} = \frac{RH_0^t}{\alpha (e^{\alpha \overline{\Delta T}} - \overline{RH_0})},$$
(S18)

$$T_{\rm LH,\Delta T}^t = -\frac{LH_{\Delta T}^t}{\alpha LH} = -\frac{\overline{RH_0}}{e^{\alpha \overline{\Delta T}} - \overline{RH_0}} \Delta T^t.$$
(S19)

Equation S12–S19 are shown in Figure 2, while the heat flux trend terms used to calculate Equation S12–S19 are shown in Figure S3.



**Figure S1.** Southern Ocean (40-65S) averaged monthly SST anomaly from SOPACE (orange shading), LENS (blue shading), ERSSTv3b dataset (black), and ERSSTv5 (green). The 1975-2016 climatology is used to calculate the monthly anomaly.



**Figure S2**. (a) Annual mean AMOC anomaly timeseries (defined as the maximum streamfunction value in the latitude band 20–60N and below 500 m in the Atlantic Ocean following Zhang et al. 2017). The black line shows the CESM forced ocean (FO) simulation (Yeager et al. 2018), approximating the observed AMOC timeseries. Blue solid curve shows the ensemble mean of LENS, with blue shading showing the range of the ensemble spread. Orange solid curve shows the ensemble mean of SOPACE, with orange shading showing the range of the ensemble spread. The dashed blue curve shows the first ensemble member of LENS, which was used to initialize all 20 members of SOPACE. The 1975-2016 climatology is used to calculate the monthly anomaly. (b)

Box and whisker plot of AMOC trend (1979-2013) from LENS and SOPACE ensemble members. Orange lines show the median values, boxes show the middle 50% members, whiskers show the 5 to 95% members, and circles show the outliers.



**Figure S3**. Box-and-whisker plot of SST trends averaged over (a) tropical South Atlantic (gray box in Figure 1f) and (b) the Southern Ocean (40S-65S) for each model ensemble. Green stars show the EM values. Orange lines show the median values. Gray horizontal lines show the observed values from various data sets: NOAA Extended Reconstruction SSTs version 3b and 5 (ERSSTv3b & ERSSTv5, Huang et al. 2017), Centennial in situ Observation-Based Estimates (COBE, Ishii et al. 2005), and the Hadley Centre Global Sea Ice and Sea Surface Temperature (HadISST, Rayner et al. 2003).



**Figure S4.** Terms in the ocean mixed layer heat budget based on monthly trends during 1979–2013 from SOPACE-internal over the tropical eastern Pacific and Atlantic. (a) SST trend, (b) trend in mixed layer heat storage (left hand side of equation (1)), (c) net longwave flux trend, (d) net shortwave flux trend, (e) sensible heat flux trend, (f) latent heat flux trend, (g) trend due to ocean dynamics, and (h) trend in liquid water path (contours show the climatological LWP). Positive flux is downward and warms the SST.



**Figure S5.** Latent heat budget based on monthly trends during 1979—2013 from SOPACE-internal over the tropical eastern Pacific and Atlantic due to (a) SST damping, (b) near-surface wind speed changes, (c) near-surface relative humidity changes, and (d) air-sea temperature difference changes. See Equation S6 for details.



**Figure S6.** Ekman advection based on monthly trends during 1979—2013 from SOPACEinternal over the tropical eastern Pacific and Atlantic, expressed as an equivalent surface heat flux trend. Area near the equator (5°S–5°N) is masked out.