# South Asian Summer Monsoon Precipitation is Sensitive to Southern Hemisphere Subtropical Radiation Changes

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# Abstract

We study the sensitivity of South Asian Summer Monsoon (SASM) precipitation to Southern Hemisphere (SH) subtropical Absorbed Solar Radiation (ASR) changes using Community Earth System Model 2 simulations. Reducing positive ASR biases over the SH subtropics impacts SASM, and is sensitive to the ocean basin where changes are imposed. Radiation changes over the SH subtropical Indian Ocean (IO) shifts rainfall over the equatorial IO northward causing 1-2 mm/day drying south of equator, changes over the SH subtropical Pacific increases precipitation over northern continental regions by 1-2 mm/day, and changes over the SH subtropical Atlantic have little effect on SASM precipitation. Radiation changes over the subtropical Pacific impacts the SASM through zonal circulation changes, while changes over the IO modify meridional circulation to bring about changes in precipitation over northern IO. Our findings suggest that reducing SH subtropical radiation biases in climate models may also reduce SASM precipitation biases.

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# Key Points:

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13	•	We test if biases in southern hemisphere shortwave radiation contributes to biases
14		in South Asian Summer Monsoon Precipitation in the CESM2.
15	•	Reducing incoming shortwave radiation in the subtropical southern hemisphere
16		reduces dry biases over continental South Asia.
17	•	This effect is mostly due to forcing in the South Pacific, with less impact from the
18		Atlantic or Indian ocean.

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#### 19 Abstract

We study the sensitivity of South Asian Summer Monsoon (SASM) precipitation to South-20 ern Hemisphere (SH) subtropical Absorbed Solar Radiation (ASR) changes using Com-21 munity Earth System Model 2 simulations. Reducing positive ASR biases over the SH 22 subtropics impacts SASM, and is sensitive to the ocean basin where changes are imposed. 23 Radiation changes over the SH subtropical Indian Ocean (IO) shifts rainfall over the equa-24 torial IO northward causing 1-2 mm/day drying south of equator, changes over the SH 25 subtropical Pacific increases precipitation over northern continental regions by 1-2 mm/day, 26 and changes over the SH subtropical Atlantic have little effect on SASM precipitation. 27 Radiation changes over the subtropical Pacific impacts the SASM through zonal circu-28 lation changes, while changes over the IO modify meridional circulation to bring about 29 changes in precipitation over northern IO. Our findings suggest that reducing SH sub-30

tropical radiation biases in climate models may also reduce SASM precipitation biases.

# 32 Plain Language Summary

Precipitation from South Asian Summer Monsoon (SASM) is of high significance 33 to the livelihoods of over a billion people. As the global climate continues to evolve, it 34 is essential to have a clear understanding of the possible future changes to the SASM. 35 However, current state-of-the-art climate models have difficulties in simulating clima-36 tological mean SASM precipitations. Here we study sensitivity of SASM precipitation 37 to subtropical southern ocean radiation as one of the possible causes of SASM precip-38 itation bias. Our experiments indicate that SASM precipitation is sensitive to southern 30 hemisphere subtropical radiation changes particularly to those in subtropical Pacific. These 40 findings imply that improving southern hemisphere subtropical radiation biases might 41 improve SASM precipitation simulations in climate models. 42

# 43 1 Introduction

The South Asian Summer Monsoon (SASM) is a highly influential monsoon sys-44 tem, known for its strength, spatial extent, and significance to the livelihoods of over a 45 billion people. Future greenhouse gas forcing is projected to increase SASM precipita-46 tion, but the magnitude and pattern of these changes remain uncertain (Katzenberger 47 et al., 2021; Kumar et al., 2023). Earth System Models (ESMs) serve as the primary means 48 of projecting these changes, but ESMs have long-standing biases in simulating the SASM's 49 climatological precipitation intensity, spatial pattern, and seasonality, among other el-50 ements (Mitra, 2021; Konda & Vissa, 2022; Rajendran et al., 2022). These biases have 51 been attributed to various sources, including errors in representation of orography (Boos 52 & Hurley, 2013), land-atmosphere interactions (Ashfaq et al., 2017), air-sea interactions 53 (Annamalai et al., 2017; Hanf & Annamalai, 2020), and sea surface temperatures over 54 the Indian Ocean (Levine et al., 2013; He et al., 2022). ESMs participating in the Cou-55 pled Model Intercomparison Project 6 (CMIP6) better simulate the SASM compared to 56 previous generations of models, but significant biases persist (Gusain et al., 2020; Choud-57 hury et al., 2021). 58

The Community Earth System Model 2 (CESM2) (Danabasoglu et al., 2020), like 59 other climate models, has systematic biases in simulating SASM precipitation. Figure 60 1a displays the June to September (JJAS) precipitation and lower tropospheric circu-61 lation biases for the 1979-2014 period compared to observations and reanalysis. CESM2 62 does not produce enough precipitation in two regions of the SASM system: (1) over the 63 central land region between  $20^{\circ}-30^{\circ}N$ , and (2) over the Indian Ocean between  $10^{\circ}S-0^{\circ}N$ with dry biases of 2-4 mm/day over both regions. Between these two dry biased areas, 65 there is a region of excessive precipitation (4-6 mm/day) in the northern Indian ocean 66 surrounding the South Asian landmass producing a meridional dry-wet-dry bias pattern. 67

- <sup>68</sup> The model also tends to overestimate precipitation over the high orography of the East-
- <sup>69</sup> ern Himalayas, Southeast Asia, and the Maritime continent.



Figure 1. Existing biases in CESM2: Ensemble mean bias of 11 members of CESM2 historical simulations from CMIP dataset in (a) JJAS precipitation, and 850 hPa wind with respect to GPCP (Adler et al., 2018) and ERA5 (Hersbach et al., 2020), respectively; (b) JJAS 200 hPa velocity potential relative to ERA5; and (c) annual mean SST relative to ERSST (Huang et al., 2015); and (d) annual mean net downward top-of-atmosphere ( $F_{TOA}$ ) radiation flux relative to CERES-EBAF (Loeb et al., 2018). Biases are calculated based on climatology of period 2004-2014 for  $F_{TOA}$  and 1979-2014 for all other variables. Boxes in panel d show the specific region where CDNC perturbations are applied for Atlantic (ATL, blue), Indian Ocean (IND, orange), Pacific (PAC, green) and Zonal (ZON, black) experiments.

SASM biases are linked to biases in the regional circulation, which partly arise from 70 local Sea Surface Temperature (SST) biases (Annamalai et al., 2017; He et al., 2022). 71 In CESM2, lower tropospheric winds are biased southeasterly and southerly over the equa-72 torial Indian ocean and parts of Arabian Sea (Figure 1a) and are biased northwesterly 73 over Indian land regions and the Bay of Bengal. Figure 1b displays the upper troposphere 74 circulation bias in terms of the 200hPa velocity potential (VP200). The VP200 bias is 75 greatest over the land regions of India and south of the equator up to 10°S. These pos-76 itive VP200 biases correspond to subsidence and weaker convergence at lower tropospheric 77 levels, thereby creating dry biases. CESM2 also displays positive SST biases in the trop-78 ics globally, which affects the Walker circulation and thus the SASM (Walker, 1923; Jain 79 et al., 2021). 80

An emerging view is that regional tropical monsoons are local manifestations of sea-81 sonal Intertropical Convergence Zone (ITCZ) shifts, and are an important feature of an 82 energy-constrained global circulation (Bordoni & Schneider, 2008; Biasutti et al., 2018; 83 Hill, 2019; Geen et al., 2020). SASM variability in particular is strongly linked to sea-84 sonal migrations of the ITCZ over the Indian subcontinent (Gadgil, 2018; Hari et al., 2020). 85 These seasonal ITCZ migrations are in turn driven by inter-hemispheric energy gradi-86 ents, as the Hadley circulation, and thus the ITCZ shift, towards the warmer hemisphere 87 to transport excess energy to the opposite hemisphere (Kang et al., 2009; Schneider et 88 al., 2014). Thus, a systematic Absorbed Solar Radiation (ASR) bias could alter this inter-89 hemispheric energy gradient, thereby disrupting this seasonal ITCZ migration and in-90 ducing biases in the monsoon. While ASR biases over the Southern Ocean have been re-91

<sup>92</sup> duced in the latest generation of climate models (Kay et al., 2012; Bodas-Salcedo et al., <sup>93</sup> 2014; Zhao et al., 2022), positive radiation biases still exist over the subtropical South-<sup>94</sup> ern Hemisphere (SH) oceans in many ESMs (Li et al., 2013, 2020). In CESM2, the mid-<sup>95</sup> latitude SH displays a negative net downward top-of-atmosphere ( $F_{TOA}$ ) radiation bias <sup>96</sup> while the subtropical SH has a positive  $F_{TOA}$  bias between 15°-30°S (Figure 1d), which <sup>97</sup> may be due to insufficient low cloud cover (Xiao et al., 2014) and too weak stratocumu-<sup>98</sup> lus cloud feedbacks (Kim et al., 2022). This positive SH subtropical  $F_{TOA}$  bias is great-<sup>99</sup> est over the Atlantic and Pacific Oceans, but is relatively small in the Indian Ocean.

100 In the present study, we aim to address the following: does improving SH subtropical radiation biases impact ITCZ position and the simulation of the SASM? And if so, 101 by what mechanism? To test this, we perform CESM2 experiments where we decrease 102 ASR in SH regions with high ASR biases by changing the cloud albedo. We change the 103 cloud properties by prescribing the Cloud Droplet Number Concentration (CDNC) that 104 influences the cloud microphysics and radiation. Our results indicate that reducing ASR 105 biases over the subtropical SH modestly improve SASM precipitation biases over the In-106 dian subcontinent. 107

#### 108 2 Methods

To understand the effect of reducing SH ASR biases on SASM, we performed a se-109 ries of experiments with the Community Earth System model 2 (CESM2, Danabasoglu 110 et al., 2020), the latest generation ESM developed by the National Center for Atmospheric 111 Research and collaborators. Our configuration uses CAM6 for atmosphere, POP2 for ocean, 112 CLM5 for land and CICE5 for sea ice. In our experiments, we perturbed the radiative 113 fluxes in CESM2 by prescribing the in-cloud liquid CDNC at all vertical levels where liq-114 uid and mixed phase clouds exist over oceanic regions in the SH subtropics. This reduces 115 the average cloud droplet radius, increasing cloud albedo and lifetime, and thus reduc-116 ing ASR. 117

First we conducted Fixed Sea Surface Temperature (Fixed-SST) experiments to 118 identify the sensitivity of  $F_{TOA}$  fluxes to CDNC, and determined the CDNC values that 119 would minimize the ASR bias compared to observations. We then carried out five fully 120 coupled simulations with repeating year-2000 forcings - one control simulation (2000CNT) 121 without any perturbations and four experiments which introduced the identified CDNC 122 perturbations in different regions of SH subtropics  $(15^{\circ}-30^{\circ}S)$  (listed in Table 1). The 123 first three simulations introduced a perturbation in each of the Atlantic (ATL)  $[50^{\circ}W$ -124 20°E], Indian (IND) [48°-115°E], and Pacific (PAC) [150°E-80°W] oceans and the fourth 125 simulation perturbed ocean regions in the entire zonal band (ZON)  $(180^{\circ}W-180^{\circ}E)$ . We 126 set the CDNC value to 150  $\text{cm}^{-3}$  for ATL and 75  $\text{cm}^{-3}$  for the PAC, IND, and ZON ex-127 periments. When compared to existing CESM2  $F_{TOA}$  bias with EBAF (Figure 1d), the 128 applied perturbations correct most of the  $F_{TOA}$  bias in the subtropical SH Atlantic but 129 over the subtropical SH Pacific and Indian Ocean the perturbation exceeds  $F_{TOA}$  bias, 130 resulting in a negative bias (Figure S1 and Table 1). Fixed-SST experiments were run 131 for 7 years and fully coupled experiments for 150 years. We calculate the response as the 132 climatological average difference between perturbation experiment and control run for 133 the last 5 and 30 years of the fixed-SST and fully coupled simulations respectively. Sta-134 tistical significance of response is tested using the Student's t-test with p < 0.05. 135

### 2.1 ITCZ position

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We compute the ITCZ latitude as the centroid of the zonal-mean, time-mean precipitation between 30°N and 30°S (Frierson & Hwang, 2012; Voigt et al., 2016). As demonstrated by Atwood et al. (2020), regional rainbelts do not shift uniformly in response to hemispherically asymmetric forcing, therefore we also analyse regional ITCZs in the Atlantic (10°W-40°W), Indian (70°E -90°E) and Pacific (110°E-100°W) oceans.

Experiment name	Abbreviation	CDNC per- turbation applied	Perturbation Region	Global $F_{TOA}$ response in W/m2	
Control	2000CNT	0	0	0	0
Atlantic	ATL	$150 \ {\rm cm}^{-3}$	$\begin{vmatrix} 50^{\circ} W-20^{\circ} E, \\ 15^{\circ}-30^{\circ} S \end{vmatrix}$	-0.21±0.87	$\begin{vmatrix} -5.43 \ \pm 1.28 \\ (5.77 \ \pm 0.19) \end{vmatrix}$
Indian Ocean	IND	$  75 \text{ cm}^{-3}  $	48°-115°E,   15°-30°S	-0.08±0.88	$\begin{vmatrix} -5.71 \pm 1.5 \\ (-2.50 \pm 0.2) \end{vmatrix}$
Pacific	PAC	$75 \text{ cm}^{-3}$	150°E-80°W,   15°-30°S	$  -0.39 \pm 0.85$	$\begin{vmatrix} -7.27 \pm 1.21 \\ (3.31 \pm 0.3) \end{vmatrix}$
Zonal	ZON	75 cm <sup>-3</sup>	$\begin{vmatrix} 180^{\circ}W-180^{\circ}E, \\ 15^{\circ}-30^{\circ}S \end{vmatrix}$	-0.53±0.88	$\begin{vmatrix} -4.34 \pm 1.23 \\ (2.43 \pm 0.23) \end{vmatrix}$

Table 1. Details of coupled CESM2 experiments used in this study

# 142 **3 Results**

#### 143

# 3.1 Energetically constrained global precipitation response

All experiments show energy being transported towards the SH subtropics, where 144 we created an ASR depression by artificially increasing CDNC, producing increasing north-145 ward transport between 60-30 S (positive values) and southward transport between 15S-146 30N (negative values) (Figure 2a). The experiments show similar responses in terms of 147 direction of transport however the magnitude of response differs. The magnitude is pro-148 portional to the  $F_{TOA}$  anomaly e.g. ZON shows highest absolute  $F_{TOA}$  and total energy 149 transport response while as IND shows least absolute values for the both (See Table 1 150 and Figure 2a). Most of this energy transport is carried out by atmospheric transport 151 as shown by Atmospheric Energy Transport (AET) response in Figure 2b. Experiments 152 show varied ocean heat transport responses but the changes are one order less than that 153 of AET (Figure 2c). These results confirm the findings of Xiang et al. (2018) that at-154 mosphere transports energy more effectively than oceans when energy perturbations are 155 applied at lower latitudes. 156

Following the energetics framework (Kang et al., 2009; Schneider et al., 2014; Dono-157 hoe et al., 2013), AET changes bring about changes in precipitations as shown in Fig-158 ure 2d. All the experiments show increase of precipitation in the tropical northern hemi-159 sphere and reductions in the tropical SH, indicating northward migration of the ITCZ. 160 Consistent with AET response, the precipitation change is larger in ZON and PAC and 161 weaker in IND and ATL. In Figure 2e and 2f, we quantify these ITCZ shifts and their 162 biases. In CESM2, the annual mean ITCZ position has northward bias of 0.25  $^{\circ}$  (Fig-163 ure 2e). However, this bias is regionally and seasonally dependent. In the context of the 164 SASM we are interested in JJAS ITCZ location in Indian Ocean, which is biased  $1.2^{\circ}$ 165 southwards (Figure 2f). All of the experiments show significant northward shifts in re-166 spective regional annual and JJAS global ITCZs (Figure 2e and 2f). Notably, IND, PAC 167 and ZON show significant northward shifts in annual and JJAS mean ITCZ over the In-168 dian Ocean. Thus, as hypothesized, our experiments show that the ITCZ responses fol-169 lowed the energetic framework. 170



Figure 2. Global energetic and precipitation response: Zonal mean response of (a) total energy transport, (b) atmospheric energy transport, (c) ocean heat transport, (d) annual mean precipitation, (e) annual and (f) JJAS ITCZ position shift in fully coupled simulation for ATL (red), IND (black), PAC (green) and ZON (blue) experiments and CESM2 existing biases (yellow in e and f panels). White dots in panels e and f indicate the statistically significant shifts of ITCZ.

### 3.2 SASM precipitation response

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We now consider how the ITCZ shifts influence SASM precipitation. Despite the 172 similar effect of IND, PAC and ZON on the JJAS Indian Ocean ITCZ, the SASM pre-173 cipitation response differ considerably in these experiments (Figure 3). Reducing ASR 174 in the PAC causes significant precipitation increases over the central continental India. 175 This stands out from the other experiments and contributes to a reduction of the JJAS 176 dry bias over this region (Figure 3c; see Figure 1a for bias). However, the increase in pre-177 cipitation of 1-2 mm/day for the PAC is only a partial compensation for the 2-4 mm/day 178 dry bias of SASM precipitation in CESM2 over this region. For the ATL experiment, 179 whose main ITCZ response is in the Atlantic, the JJAS SASM precipitation change is 180 largely non-significant over the SASM region (Figure 3a). It is also interesting to see that 181 IND experiment moves the annual and JJAS ITCZ northwards (Figure 2e and 2f) and 182 produces precipitation decrease of 1-2 mm/day (Figure 3b) south of equator thus adding 183 to JJAS SASM dry bias over there (Figure 1a). The ZON experiment is similar to IND 184 experiment: significant precipitation increase of 1-2mm/day add to existing wet bias over 185 oceanic regions surrounding Indian land and precipitation decrease of 1mm/day add to 186



Figure 3. SASM precipitation response: JJAS precipitation response (at simulation period 121-150 years) in fully coupled simulation for experiments (a) ATL, (b) IND, (c) PAC and (d) ZON. Non-significant responses (at level of significance 0.05) are masked in white.

dry biases (Figure 3d) south of the equator. It also reduces dry bias of a small portion over central India east of 80°E. To better decipher these intriguing SASM precipitation responses in different experiments we further explore the large-scale circulation changes in each of these experiments.

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# 3.3 Large-scale circulation response

Panels a-d in Figure 4 show lower tropospheric circulation and associated moisture 192 divergence for each of the experiments. For IND and ZON, the 850hPa wind responses 193 add to existing southeasterly and southerly biases in the equatorial Indian Ocean and 194 parts of the Arabian sea (Figure 4b and d, see Figure 1a for bias). The important wind 195 response of PAC is a southeasterly response over land regions of central India (Figure 196 4c) which reduces northwesterly circulation bias (Figure 1a). In tandem with lower tro-197 pospheric circulation changes, we observe that moisture convergence increases over In-198 dian land region for PAC but over the surrounding oceanic regions for ZON and IND 199 (Figure 4b,c and d). The ATL shows statistically significant increase in lower tropospheric 200 westerlies and the moisture convergence along the west coast (Figure 4a) over a small 201 region. The improved representation of lower level winds and increased moisture flux con-202 vergence over central continental India in PAC, contributes to a partial reduction of the 203 dry precipitation bias over this region. 204

The SASM regional circulation is tightly linked to the large-scale Hadley circula-205 tion in the meridional plane and the Walker circulation in the zonal plane (Goswami & 206 Chakravorty, 2017). ZON and IND, show strengthening of the meridional overturning 207 circulation with enhanced south of equator descent and north of equator ascent (Figure 208 4h,f) but the enhancement in ascending motion extends only up to 15° N. ATL shows 209 no significant circulation changes in the meridional plane (Figure 4e). The PAC is quite 210 different than others that it shows ascent over a broader latitudinal band concentrated 211 in upper levels above 500hPa and no enhanced descent is observed south of the equa-212 tor over Indian longitudes, indicating a Walker circulation change (Figure 4g). 213



Figure 4. Circulation response: Response of (a-d) Vertically integrated moisture divergence between 1000-700hPa (shading) and 850hPa winds (vectors), (e-h) local meridional overturning circulation (average  $70^{\circ}-90^{\circ}$ E), (i-l) 200hPa velocity potential and (m-p) surface temperature in fully-coupled simulations for different experiments - (a,e,i,m) ATL, (b,f,j,n) IND, (c,g,k,o) PAC and (d,h,l,p) ZON for the climatology of simulation years 121-150. Only significant (p < 0.05) changes are shown in colour.

We confirm this by analysing the VP200 response of these experiments as proxy 214 to the Walker circulation changes in the Indo-Pacific (Tanaka et al., 2004; Vecchi & So-215 den, 2007). For ZON and PAC, VP200 is reduced over the Asian continent (Figure 4k,l), 216 reducing the existing positive bias. On the other hand, for IND, the main response is 217 an increase in VP200 south of the equator over the Indian Ocean (Figure 4j) which adds 218 to the existing positive bias (Figure 1b). For ZON and PAC we see a positive VP200 re-219 sponse in the Pacific Ocean south of the equator. Increases in VP200 in the Pacific and 220 reduction over the Indian ocean are indications of a La Niña like Walker circulation re-221 sponse in these experiments. However existing strong negative VP200 biases in the north 222 Pacific and the equatorial Atlantic remained same in all of the experiments (Figure 1b). 223 The emergence of La Niña like conditions are further confirmed by the SST response, 224 which shows cooling in the eastern equatorial Pacific in PAC and ZON (Figure 40,p). 225 This reduces the positive existing SST bias in this region (Figure 1c). In summary, the 226 SASM changes in IND and ZON are brought about by changes in the meridional over-227 turning circulation over Indian longitudes, but for PAC the driving mechanism is the zonal 228 circulation change influenced by the equatorial Pacific SST. The ATL perturbation pro-229 duces negligible changes in the large-scale circulation over South Asia. Reducing ASR 230 biases over the subtropical Pacific modestly improve SASM precipitation biases over the 231 Indian subcontinent. However, the key dynamic mechanism driving these improvements 232 is not inter-hemispheric energy gradients, but instead zonal walker circulation changes 233 caused the observed improvements. Our results imply that improving southern hemi-234 sphere subtropical radiation biases might improve SASM precipitation in climate mod-235 els. 236

#### 237 4 Discussion

In this study we investigated whether correcting an existing SH subtropical ASR 238 bias in CESM2 will drive the ITCZ northwards and thereby correct the SASM precip-239 itation biases during JJAS. Using CESM2 coupled simulations, we demonstrate that ra-240 diation biases in subtropical regions can indeed influence biases in simulations of regional 241 monsoon precipitations. In general, when we correct SH subtropical ASR biases, global 242 mean annual ITCZ shift northward following the energetic framework. Three of the ex-243 periments; IND, PAC and ZON; moved JJAS ITCZ northwards in the Indian Ocean re-244 245 ducing a southward bias. However these JJAS ITCZ corrections could not completely erase the SASM biases. The responses are regionally specific. An ASR correction in the 246 ATL region has weak impacts on the SASM precipitation. Reducing the ASR bias over 247 the subtropical southern Indian Ocean induces changes in meridional circulation over the 248 SASM region and reduces precipitation in the equatorial Indian Ocean. In contrast re-249 ducing the bias over the subtropical south Pacific induces a La Niña-like temperature 250 anomaly and circulation response thereby modestly increases the precipitation over the 251 central land region of the SASM. 252

Our experiments suggest that the mean precipitation bias over the central land of 253 the SASM region may be partially connected to the subtropical south Pacific radiation 254 biases. It is interesting to see that a reduction in sunlight reaching the PAC can induce 255 cooling in the equatorial Pacific and trigger a La Niña-like temperature and circulation 256 response. However, the  $F_{TOA}$  corrections we applied here do not improve the biases in 257 ocean regions surrounding South Asia; although the IND experiment hints to the merid-258 ional circulation anomalies that contribute to the dry bias south of equator in the In-259 dian Ocean. Previous studies (Annamalai et al., 2017; He et al., 2022) have suggested 260 that some of biases in oceanic precipitation can be explained by the local SST biases and 261 errors in air-sea interactions over the northern Indian Ocean regions. 262

These results have implications for improvements in climate model monsoon fidelity, 263 as we demonstrate that the SASM precipitation biases in CMIP6 climate models are in-264 fluenced by the  $F_{TOA}$  radiation biases over subtropical southern hemisphere, particu-265 larly in the Indian and Pacific ocean basins (Figure S8). Given that a poor representa-266 tion of marine stratocumulus clouds in climate models is the main cause of the ASR bi-267 ases in subtropics (Jian et al., 2021), it is important to improve their representations in 268 the climate models for better monsoon simulations. Coupled models used for seasonal 269 prediction exhibit similar SASM biases i.e. dry bias over Indian land region and wet bias 270 in surrounding oceanic region (Pillai et al., 2018). Major efforts are underway to under-271 stand and reduce these biases (e.g. Krishna et al. (2019); Fousiya et al. (2023)) and our 272 findings indicate improvements in subtropical stratocumulus cloud biases may also pro-273 vide improvements in seasonal prediction. Though here we presented results in context 274 of the SASM precipitation only, other monsoon regions may see benefits from reducing 275 low cloud biases. Notably, the Atlantic experiment shows reduction of the South Amer-276 ican December-February precipitation biases over Brazil (Figures S2 and S7) 277

One important caveat of this work is that the sensitivity of southern subtropical radiation changes to SASM precipitation explained here could be different in different climate models. We performed experiments in only one model, CESM2, so the results should be interpreted cautiously, to account for CESM2 idiosyncrasies (like the high climate sensitivity (Gettelman et al., 2019) and large ENSO amplitude (Planton et al., 2021)). Despite this, our findings suggest that improving SH subtropical radiation biases may have the co-benefit of reducing SASM precipitation biases as well.

# <sup>285</sup> 5 Open Research

CESM2 code modifications, model outputs and analysis scripts used to produce plots presented in the paper are available on Zenodo at https://doi.org/10.5281/zenodo .10536019.

Following are the freely available observation and climate model datasets used in 289 the study - CMIP6 models used in this study are listed in Table S1 and the data is down-290 loaded from the ESGF data portal (https://esgf-node.llnl.gov/search/cmip6) (Evring 291 et al., 2016), the ERA5 reanalysis data downloaded from the Copernicus Climate Data 292 Store at doi:10.24381/cds.6860a573 (Hersbach et al., 2020), CERES-EBAF version 293 4.2 data downloaded from https://ceres.larc.nasa.gov/data/ (Loeb et al., 2018), 294 the ERSST version 4 data downloaded from from their website at doi:10.7289/V5KD1VVF 295 (Huang et al., 2015), and monthly GPCP precipitation version 2.3 data downloaded from 296 National Centers for Environmental Information website doi:10.7289/V56971M6 (Adler 297 et al., 2018). 298

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# South Asian Summer Monsoon Precipitation is Sensitive to Southern Hemisphere Subtropical Radiation Changes

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# Key Points:

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13	•	We test if biases in southern hemisphere shortwave radiation contributes to biases
14		in South Asian Summer Monsoon Precipitation in the CESM2.
15	•	Reducing incoming shortwave radiation in the subtropical southern hemisphere
16		reduces dry biases over continental South Asia.
17	•	This effect is mostly due to forcing in the South Pacific, with less impact from the
18		Atlantic or Indian ocean.

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#### 19 Abstract

We study the sensitivity of South Asian Summer Monsoon (SASM) precipitation to South-20 ern Hemisphere (SH) subtropical Absorbed Solar Radiation (ASR) changes using Com-21 munity Earth System Model 2 simulations. Reducing positive ASR biases over the SH 22 subtropics impacts SASM, and is sensitive to the ocean basin where changes are imposed. 23 Radiation changes over the SH subtropical Indian Ocean (IO) shifts rainfall over the equa-24 torial IO northward causing 1-2 mm/day drying south of equator, changes over the SH 25 subtropical Pacific increases precipitation over northern continental regions by 1-2 mm/day, 26 and changes over the SH subtropical Atlantic have little effect on SASM precipitation. 27 Radiation changes over the subtropical Pacific impacts the SASM through zonal circu-28 lation changes, while changes over the IO modify meridional circulation to bring about 29 changes in precipitation over northern IO. Our findings suggest that reducing SH sub-30

tropical radiation biases in climate models may also reduce SASM precipitation biases.

# 32 Plain Language Summary

Precipitation from South Asian Summer Monsoon (SASM) is of high significance 33 to the livelihoods of over a billion people. As the global climate continues to evolve, it 34 is essential to have a clear understanding of the possible future changes to the SASM. 35 However, current state-of-the-art climate models have difficulties in simulating clima-36 tological mean SASM precipitations. Here we study sensitivity of SASM precipitation 37 to subtropical southern ocean radiation as one of the possible causes of SASM precip-38 itation bias. Our experiments indicate that SASM precipitation is sensitive to southern 30 hemisphere subtropical radiation changes particularly to those in subtropical Pacific. These 40 findings imply that improving southern hemisphere subtropical radiation biases might 41 improve SASM precipitation simulations in climate models. 42

# 43 1 Introduction

The South Asian Summer Monsoon (SASM) is a highly influential monsoon sys-44 tem, known for its strength, spatial extent, and significance to the livelihoods of over a 45 billion people. Future greenhouse gas forcing is projected to increase SASM precipita-46 tion, but the magnitude and pattern of these changes remain uncertain (Katzenberger 47 et al., 2021; Kumar et al., 2023). Earth System Models (ESMs) serve as the primary means 48 of projecting these changes, but ESMs have long-standing biases in simulating the SASM's 49 climatological precipitation intensity, spatial pattern, and seasonality, among other el-50 ements (Mitra, 2021; Konda & Vissa, 2022; Rajendran et al., 2022). These biases have 51 been attributed to various sources, including errors in representation of orography (Boos 52 & Hurley, 2013), land-atmosphere interactions (Ashfaq et al., 2017), air-sea interactions 53 (Annamalai et al., 2017; Hanf & Annamalai, 2020), and sea surface temperatures over 54 the Indian Ocean (Levine et al., 2013; He et al., 2022). ESMs participating in the Cou-55 pled Model Intercomparison Project 6 (CMIP6) better simulate the SASM compared to 56 previous generations of models, but significant biases persist (Gusain et al., 2020; Choud-57 hury et al., 2021). 58

The Community Earth System Model 2 (CESM2) (Danabasoglu et al., 2020), like 59 other climate models, has systematic biases in simulating SASM precipitation. Figure 60 1a displays the June to September (JJAS) precipitation and lower tropospheric circu-61 lation biases for the 1979-2014 period compared to observations and reanalysis. CESM2 62 does not produce enough precipitation in two regions of the SASM system: (1) over the 63 central land region between  $20^{\circ}-30^{\circ}N$ , and (2) over the Indian Ocean between  $10^{\circ}S-0^{\circ}N$ with dry biases of 2-4 mm/day over both regions. Between these two dry biased areas, 65 there is a region of excessive precipitation (4-6 mm/day) in the northern Indian ocean 66 surrounding the South Asian landmass producing a meridional dry-wet-dry bias pattern. 67

- <sup>68</sup> The model also tends to overestimate precipitation over the high orography of the East-
- <sup>69</sup> ern Himalayas, Southeast Asia, and the Maritime continent.



Figure 1. Existing biases in CESM2: Ensemble mean bias of 11 members of CESM2 historical simulations from CMIP dataset in (a) JJAS precipitation, and 850 hPa wind with respect to GPCP (Adler et al., 2018) and ERA5 (Hersbach et al., 2020), respectively; (b) JJAS 200 hPa velocity potential relative to ERA5; and (c) annual mean SST relative to ERSST (Huang et al., 2015); and (d) annual mean net downward top-of-atmosphere ( $F_{TOA}$ ) radiation flux relative to CERES-EBAF (Loeb et al., 2018). Biases are calculated based on climatology of period 2004-2014 for  $F_{TOA}$  and 1979-2014 for all other variables. Boxes in panel d show the specific region where CDNC perturbations are applied for Atlantic (ATL, blue), Indian Ocean (IND, orange), Pacific (PAC, green) and Zonal (ZON, black) experiments.

SASM biases are linked to biases in the regional circulation, which partly arise from 70 local Sea Surface Temperature (SST) biases (Annamalai et al., 2017; He et al., 2022). 71 In CESM2, lower tropospheric winds are biased southeasterly and southerly over the equa-72 torial Indian ocean and parts of Arabian Sea (Figure 1a) and are biased northwesterly 73 over Indian land regions and the Bay of Bengal. Figure 1b displays the upper troposphere 74 circulation bias in terms of the 200hPa velocity potential (VP200). The VP200 bias is 75 greatest over the land regions of India and south of the equator up to 10°S. These pos-76 itive VP200 biases correspond to subsidence and weaker convergence at lower tropospheric 77 levels, thereby creating dry biases. CESM2 also displays positive SST biases in the trop-78 ics globally, which affects the Walker circulation and thus the SASM (Walker, 1923; Jain 79 et al., 2021). 80

An emerging view is that regional tropical monsoons are local manifestations of sea-81 sonal Intertropical Convergence Zone (ITCZ) shifts, and are an important feature of an 82 energy-constrained global circulation (Bordoni & Schneider, 2008; Biasutti et al., 2018; 83 Hill, 2019; Geen et al., 2020). SASM variability in particular is strongly linked to sea-84 sonal migrations of the ITCZ over the Indian subcontinent (Gadgil, 2018; Hari et al., 2020). 85 These seasonal ITCZ migrations are in turn driven by inter-hemispheric energy gradi-86 ents, as the Hadley circulation, and thus the ITCZ shift, towards the warmer hemisphere 87 to transport excess energy to the opposite hemisphere (Kang et al., 2009; Schneider et 88 al., 2014). Thus, a systematic Absorbed Solar Radiation (ASR) bias could alter this inter-89 hemispheric energy gradient, thereby disrupting this seasonal ITCZ migration and in-90 ducing biases in the monsoon. While ASR biases over the Southern Ocean have been re-91

<sup>92</sup> duced in the latest generation of climate models (Kay et al., 2012; Bodas-Salcedo et al., <sup>93</sup> 2014; Zhao et al., 2022), positive radiation biases still exist over the subtropical South-<sup>94</sup> ern Hemisphere (SH) oceans in many ESMs (Li et al., 2013, 2020). In CESM2, the mid-<sup>95</sup> latitude SH displays a negative net downward top-of-atmosphere ( $F_{TOA}$ ) radiation bias <sup>96</sup> while the subtropical SH has a positive  $F_{TOA}$  bias between 15°-30°S (Figure 1d), which <sup>97</sup> may be due to insufficient low cloud cover (Xiao et al., 2014) and too weak stratocumu-<sup>98</sup> lus cloud feedbacks (Kim et al., 2022). This positive SH subtropical  $F_{TOA}$  bias is great-<sup>99</sup> est over the Atlantic and Pacific Oceans, but is relatively small in the Indian Ocean.

100 In the present study, we aim to address the following: does improving SH subtropical radiation biases impact ITCZ position and the simulation of the SASM? And if so, 101 by what mechanism? To test this, we perform CESM2 experiments where we decrease 102 ASR in SH regions with high ASR biases by changing the cloud albedo. We change the 103 cloud properties by prescribing the Cloud Droplet Number Concentration (CDNC) that 104 influences the cloud microphysics and radiation. Our results indicate that reducing ASR 105 biases over the subtropical SH modestly improve SASM precipitation biases over the In-106 dian subcontinent. 107

#### 108 2 Methods

To understand the effect of reducing SH ASR biases on SASM, we performed a se-109 ries of experiments with the Community Earth System model 2 (CESM2, Danabasoglu 110 et al., 2020), the latest generation ESM developed by the National Center for Atmospheric 111 Research and collaborators. Our configuration uses CAM6 for atmosphere, POP2 for ocean, 112 CLM5 for land and CICE5 for sea ice. In our experiments, we perturbed the radiative 113 fluxes in CESM2 by prescribing the in-cloud liquid CDNC at all vertical levels where liq-114 uid and mixed phase clouds exist over oceanic regions in the SH subtropics. This reduces 115 the average cloud droplet radius, increasing cloud albedo and lifetime, and thus reduc-116 ing ASR. 117

First we conducted Fixed Sea Surface Temperature (Fixed-SST) experiments to 118 identify the sensitivity of  $F_{TOA}$  fluxes to CDNC, and determined the CDNC values that 119 would minimize the ASR bias compared to observations. We then carried out five fully 120 coupled simulations with repeating year-2000 forcings - one control simulation (2000CNT) 121 without any perturbations and four experiments which introduced the identified CDNC 122 perturbations in different regions of SH subtropics  $(15^{\circ}-30^{\circ}S)$  (listed in Table 1). The 123 first three simulations introduced a perturbation in each of the Atlantic (ATL)  $[50^{\circ}W$ -124 20°E], Indian (IND) [48°-115°E], and Pacific (PAC) [150°E-80°W] oceans and the fourth 125 simulation perturbed ocean regions in the entire zonal band (ZON)  $(180^{\circ}W-180^{\circ}E)$ . We 126 set the CDNC value to 150  $\text{cm}^{-3}$  for ATL and 75  $\text{cm}^{-3}$  for the PAC, IND, and ZON ex-127 periments. When compared to existing CESM2  $F_{TOA}$  bias with EBAF (Figure 1d), the 128 applied perturbations correct most of the  $F_{TOA}$  bias in the subtropical SH Atlantic but 129 over the subtropical SH Pacific and Indian Ocean the perturbation exceeds  $F_{TOA}$  bias, 130 resulting in a negative bias (Figure S1 and Table 1). Fixed-SST experiments were run 131 for 7 years and fully coupled experiments for 150 years. We calculate the response as the 132 climatological average difference between perturbation experiment and control run for 133 the last 5 and 30 years of the fixed-SST and fully coupled simulations respectively. Sta-134 tistical significance of response is tested using the Student's t-test with p < 0.05. 135

### 2.1 ITCZ position

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We compute the ITCZ latitude as the centroid of the zonal-mean, time-mean precipitation between 30°N and 30°S (Frierson & Hwang, 2012; Voigt et al., 2016). As demonstrated by Atwood et al. (2020), regional rainbelts do not shift uniformly in response to hemispherically asymmetric forcing, therefore we also analyse regional ITCZs in the Atlantic (10°W-40°W), Indian (70°E -90°E) and Pacific (110°E-100°W) oceans.

Experiment name	Abbreviation	CDNC per- turbation applied	Perturbation Region	$ \begin{array}{ c c } Global \\ F_{TOA} \text{ response in} \\ W/m2 \end{array} $	
Control	2000CNT	0	0	0	0
Atlantic	ATL	$150 \ {\rm cm}^{-3}$	$\begin{vmatrix} 50^{\circ} W-20^{\circ} E, \\ 15^{\circ}-30^{\circ} S \end{vmatrix}$	-0.21±0.87	$\begin{vmatrix} -5.43 \ \pm 1.28 \\ (5.77 \ \pm 0.19) \end{vmatrix}$
Indian Ocean	IND	$  75 \text{ cm}^{-3}  $	48°-115°E,   15°-30°S	-0.08±0.88	$\begin{vmatrix} -5.71 \pm 1.5 \\ (-2.50 \pm 0.2) \end{vmatrix}$
Pacific	PAC	$75 \text{ cm}^{-3}$	150°E-80°W,   15°-30°S	$  -0.39 \pm 0.85$	$\begin{vmatrix} -7.27 \pm 1.21 \\ (3.31 \pm 0.3) \end{vmatrix}$
Zonal	ZON	75 cm <sup>-3</sup>	$\begin{vmatrix} 180^{\circ}W-180^{\circ}E, \\ 15^{\circ}-30^{\circ}S \end{vmatrix}$	-0.53±0.88	$\begin{vmatrix} -4.34 \pm 1.23 \\ (2.43 \pm 0.23) \end{vmatrix}$

Table 1. Details of coupled CESM2 experiments used in this study

# 142 **3 Results**

#### 143

# 3.1 Energetically constrained global precipitation response

All experiments show energy being transported towards the SH subtropics, where 144 we created an ASR depression by artificially increasing CDNC, producing increasing north-145 ward transport between 60-30 S (positive values) and southward transport between 15S-146 30N (negative values) (Figure 2a). The experiments show similar responses in terms of 147 direction of transport however the magnitude of response differs. The magnitude is pro-148 portional to the  $F_{TOA}$  anomaly e.g. ZON shows highest absolute  $F_{TOA}$  and total energy 149 transport response while as IND shows least absolute values for the both (See Table 1 150 and Figure 2a). Most of this energy transport is carried out by atmospheric transport 151 as shown by Atmospheric Energy Transport (AET) response in Figure 2b. Experiments 152 show varied ocean heat transport responses but the changes are one order less than that 153 of AET (Figure 2c). These results confirm the findings of Xiang et al. (2018) that at-154 mosphere transports energy more effectively than oceans when energy perturbations are 155 applied at lower latitudes. 156

Following the energetics framework (Kang et al., 2009; Schneider et al., 2014; Dono-157 hoe et al., 2013), AET changes bring about changes in precipitations as shown in Fig-158 ure 2d. All the experiments show increase of precipitation in the tropical northern hemi-159 sphere and reductions in the tropical SH, indicating northward migration of the ITCZ. 160 Consistent with AET response, the precipitation change is larger in ZON and PAC and 161 weaker in IND and ATL. In Figure 2e and 2f, we quantify these ITCZ shifts and their 162 biases. In CESM2, the annual mean ITCZ position has northward bias of 0.25  $^{\circ}$  (Fig-163 ure 2e). However, this bias is regionally and seasonally dependent. In the context of the 164 SASM we are interested in JJAS ITCZ location in Indian Ocean, which is biased  $1.2^{\circ}$ 165 southwards (Figure 2f). All of the experiments show significant northward shifts in re-166 spective regional annual and JJAS global ITCZs (Figure 2e and 2f). Notably, IND, PAC 167 and ZON show significant northward shifts in annual and JJAS mean ITCZ over the In-168 dian Ocean. Thus, as hypothesized, our experiments show that the ITCZ responses fol-169 lowed the energetic framework. 170



Figure 2. Global energetic and precipitation response: Zonal mean response of (a) total energy transport, (b) atmospheric energy transport, (c) ocean heat transport, (d) annual mean precipitation, (e) annual and (f) JJAS ITCZ position shift in fully coupled simulation for ATL (red), IND (black), PAC (green) and ZON (blue) experiments and CESM2 existing biases (yellow in e and f panels). White dots in panels e and f indicate the statistically significant shifts of ITCZ.

### 3.2 SASM precipitation response

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We now consider how the ITCZ shifts influence SASM precipitation. Despite the 172 similar effect of IND, PAC and ZON on the JJAS Indian Ocean ITCZ, the SASM pre-173 cipitation response differ considerably in these experiments (Figure 3). Reducing ASR 174 in the PAC causes significant precipitation increases over the central continental India. 175 This stands out from the other experiments and contributes to a reduction of the JJAS 176 dry bias over this region (Figure 3c; see Figure 1a for bias). However, the increase in pre-177 cipitation of 1-2 mm/day for the PAC is only a partial compensation for the 2-4 mm/day 178 dry bias of SASM precipitation in CESM2 over this region. For the ATL experiment, 179 whose main ITCZ response is in the Atlantic, the JJAS SASM precipitation change is 180 largely non-significant over the SASM region (Figure 3a). It is also interesting to see that 181 IND experiment moves the annual and JJAS ITCZ northwards (Figure 2e and 2f) and 182 produces precipitation decrease of 1-2 mm/day (Figure 3b) south of equator thus adding 183 to JJAS SASM dry bias over there (Figure 1a). The ZON experiment is similar to IND 184 experiment: significant precipitation increase of 1-2mm/day add to existing wet bias over 185 oceanic regions surrounding Indian land and precipitation decrease of 1mm/day add to 186



Figure 3. SASM precipitation response: JJAS precipitation response (at simulation period 121-150 years) in fully coupled simulation for experiments (a) ATL, (b) IND, (c) PAC and (d) ZON. Non-significant responses (at level of significance 0.05) are masked in white.

dry biases (Figure 3d) south of the equator. It also reduces dry bias of a small portion over central India east of 80°E. To better decipher these intriguing SASM precipitation responses in different experiments we further explore the large-scale circulation changes in each of these experiments.

191

# 3.3 Large-scale circulation response

Panels a-d in Figure 4 show lower tropospheric circulation and associated moisture 192 divergence for each of the experiments. For IND and ZON, the 850hPa wind responses 193 add to existing southeasterly and southerly biases in the equatorial Indian Ocean and 194 parts of the Arabian sea (Figure 4b and d, see Figure 1a for bias). The important wind 195 response of PAC is a southeasterly response over land regions of central India (Figure 196 4c) which reduces northwesterly circulation bias (Figure 1a). In tandem with lower tro-197 pospheric circulation changes, we observe that moisture convergence increases over In-198 dian land region for PAC but over the surrounding oceanic regions for ZON and IND 199 (Figure 4b,c and d). The ATL shows statistically significant increase in lower tropospheric 200 westerlies and the moisture convergence along the west coast (Figure 4a) over a small 201 region. The improved representation of lower level winds and increased moisture flux con-202 vergence over central continental India in PAC, contributes to a partial reduction of the 203 dry precipitation bias over this region. 204

The SASM regional circulation is tightly linked to the large-scale Hadley circula-205 tion in the meridional plane and the Walker circulation in the zonal plane (Goswami & 206 Chakravorty, 2017). ZON and IND, show strengthening of the meridional overturning 207 circulation with enhanced south of equator descent and north of equator ascent (Figure 208 4h,f) but the enhancement in ascending motion extends only up to 15° N. ATL shows 209 no significant circulation changes in the meridional plane (Figure 4e). The PAC is quite 210 different than others that it shows ascent over a broader latitudinal band concentrated 211 in upper levels above 500hPa and no enhanced descent is observed south of the equa-212 tor over Indian longitudes, indicating a Walker circulation change (Figure 4g). 213



Figure 4. Circulation response: Response of (a-d) Vertically integrated moisture divergence between 1000-700hPa (shading) and 850hPa winds (vectors), (e-h) local meridional overturning circulation (average  $70^{\circ}-90^{\circ}$ E), (i-l) 200hPa velocity potential and (m-p) surface temperature in fully-coupled simulations for different experiments - (a,e,i,m) ATL, (b,f,j,n) IND, (c,g,k,o) PAC and (d,h,l,p) ZON for the climatology of simulation years 121-150. Only significant (p < 0.05) changes are shown in colour.

We confirm this by analysing the VP200 response of these experiments as proxy 214 to the Walker circulation changes in the Indo-Pacific (Tanaka et al., 2004; Vecchi & So-215 den, 2007). For ZON and PAC, VP200 is reduced over the Asian continent (Figure 4k,l), 216 reducing the existing positive bias. On the other hand, for IND, the main response is 217 an increase in VP200 south of the equator over the Indian Ocean (Figure 4j) which adds 218 to the existing positive bias (Figure 1b). For ZON and PAC we see a positive VP200 re-219 sponse in the Pacific Ocean south of the equator. Increases in VP200 in the Pacific and 220 reduction over the Indian ocean are indications of a La Niña like Walker circulation re-221 sponse in these experiments. However existing strong negative VP200 biases in the north 222 Pacific and the equatorial Atlantic remained same in all of the experiments (Figure 1b). 223 The emergence of La Niña like conditions are further confirmed by the SST response, 224 which shows cooling in the eastern equatorial Pacific in PAC and ZON (Figure 40,p). 225 This reduces the positive existing SST bias in this region (Figure 1c). In summary, the 226 SASM changes in IND and ZON are brought about by changes in the meridional over-227 turning circulation over Indian longitudes, but for PAC the driving mechanism is the zonal 228 circulation change influenced by the equatorial Pacific SST. The ATL perturbation pro-229 duces negligible changes in the large-scale circulation over South Asia. Reducing ASR 230 biases over the subtropical Pacific modestly improve SASM precipitation biases over the 231 Indian subcontinent. However, the key dynamic mechanism driving these improvements 232 is not inter-hemispheric energy gradients, but instead zonal walker circulation changes 233 caused the observed improvements. Our results imply that improving southern hemi-234 sphere subtropical radiation biases might improve SASM precipitation in climate mod-235 els. 236

#### 237 4 Discussion

In this study we investigated whether correcting an existing SH subtropical ASR 238 bias in CESM2 will drive the ITCZ northwards and thereby correct the SASM precip-239 itation biases during JJAS. Using CESM2 coupled simulations, we demonstrate that ra-240 diation biases in subtropical regions can indeed influence biases in simulations of regional 241 monsoon precipitations. In general, when we correct SH subtropical ASR biases, global 242 mean annual ITCZ shift northward following the energetic framework. Three of the ex-243 periments; IND, PAC and ZON; moved JJAS ITCZ northwards in the Indian Ocean re-244 245 ducing a southward bias. However these JJAS ITCZ corrections could not completely erase the SASM biases. The responses are regionally specific. An ASR correction in the 246 ATL region has weak impacts on the SASM precipitation. Reducing the ASR bias over 247 the subtropical southern Indian Ocean induces changes in meridional circulation over the 248 SASM region and reduces precipitation in the equatorial Indian Ocean. In contrast re-249 ducing the bias over the subtropical south Pacific induces a La Niña-like temperature 250 anomaly and circulation response thereby modestly increases the precipitation over the 251 central land region of the SASM. 252

Our experiments suggest that the mean precipitation bias over the central land of 253 the SASM region may be partially connected to the subtropical south Pacific radiation 254 biases. It is interesting to see that a reduction in sunlight reaching the PAC can induce 255 cooling in the equatorial Pacific and trigger a La Niña-like temperature and circulation 256 response. However, the  $F_{TOA}$  corrections we applied here do not improve the biases in 257 ocean regions surrounding South Asia; although the IND experiment hints to the merid-258 ional circulation anomalies that contribute to the dry bias south of equator in the In-259 dian Ocean. Previous studies (Annamalai et al., 2017; He et al., 2022) have suggested 260 that some of biases in oceanic precipitation can be explained by the local SST biases and 261 errors in air-sea interactions over the northern Indian Ocean regions. 262

These results have implications for improvements in climate model monsoon fidelity, 263 as we demonstrate that the SASM precipitation biases in CMIP6 climate models are in-264 fluenced by the  $F_{TOA}$  radiation biases over subtropical southern hemisphere, particu-265 larly in the Indian and Pacific ocean basins (Figure S8). Given that a poor representa-266 tion of marine stratocumulus clouds in climate models is the main cause of the ASR bi-267 ases in subtropics (Jian et al., 2021), it is important to improve their representations in 268 the climate models for better monsoon simulations. Coupled models used for seasonal 269 prediction exhibit similar SASM biases i.e. dry bias over Indian land region and wet bias 270 in surrounding oceanic region (Pillai et al., 2018). Major efforts are underway to under-271 stand and reduce these biases (e.g. Krishna et al. (2019); Fousiya et al. (2023)) and our 272 findings indicate improvements in subtropical stratocumulus cloud biases may also pro-273 vide improvements in seasonal prediction. Though here we presented results in context 274 of the SASM precipitation only, other monsoon regions may see benefits from reducing 275 low cloud biases. Notably, the Atlantic experiment shows reduction of the South Amer-276 ican December-February precipitation biases over Brazil (Figures S2 and S7) 277

One important caveat of this work is that the sensitivity of southern subtropical radiation changes to SASM precipitation explained here could be different in different climate models. We performed experiments in only one model, CESM2, so the results should be interpreted cautiously, to account for CESM2 idiosyncrasies (like the high climate sensitivity (Gettelman et al., 2019) and large ENSO amplitude (Planton et al., 2021)). Despite this, our findings suggest that improving SH subtropical radiation biases may have the co-benefit of reducing SASM precipitation biases as well.

# <sup>285</sup> 5 Open Research

CESM2 code modifications, model outputs and analysis scripts used to produce plots presented in the paper are available on Zenodo at https://doi.org/10.5281/zenodo .10536019.

Following are the freely available observation and climate model datasets used in 289 the study - CMIP6 models used in this study are listed in Table S1 and the data is down-290 loaded from the ESGF data portal (https://esgf-node.llnl.gov/search/cmip6) (Evring 291 et al., 2016), the ERA5 reanalysis data downloaded from the Copernicus Climate Data 292 Store at doi:10.24381/cds.6860a573 (Hersbach et al., 2020), CERES-EBAF version 293 4.2 data downloaded from https://ceres.larc.nasa.gov/data/ (Loeb et al., 2018), 294 the ERSST version 4 data downloaded from from their website at doi:10.7289/V5KD1VVF 295 (Huang et al., 2015), and monthly GPCP precipitation version 2.3 data downloaded from 296 National Centers for Environmental Information website doi:10.7289/V56971M6 (Adler 297 et al., 2018). 298

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# Supporting Information for South Asian Summer Monsoon Precipitation is Sensitive to Southern Hemisphere Subtropical Radiation Changes

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# Contents of this file

- 1. Description of observations and reanalysis data
- 2. Figures S1 to S8
- 3. Table S1

# Description of observations and reanalysis data

Monthly observation data used in this study includes precipitation from the Global Precipitation Climatology Project version 2.3 (GPCP; Adler et al. (2018)), Sea Surface Temperature (SST) data from Extended Reconstructed Sea Surface Temperature (ERSST),

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version 4 (Huang et al., 2015) and Top Of Atmosphere (TOA) radiation fluxes from Clouds and the Earth's Radiant Energy System Energy Balanced and Filled (CERES-EBAF) version 4.2 (Loeb et al., 2018). Gridded reanalyzed upper-air circulation products used in the study were obtained from the European Centre for Medium-Range Weather Forecasts (ECMWF) ERA5 (Hersbach et al., 2020). To calculate biases in the CESM2 we used 11 historical simulations from CMIP6 archive. The biases are calculated as the climatological mean difference between model and observations for the period 1979-2014, except for TOA radiation fluxes where we use climatology of 2004-2014, as the CERES instrument was launched in 1999.

# CMIP6 Data

We used historical monthly data from 11 ensemble members of CESM2 (Danabasoglu et al., 2020) participating in CMIP6 to analyse the existing CESM2 biases. To calculate multimodel regression of SASM precipitation biases onto net TOA biases we used 32 CMIP6 (Eyring et al., 2016) ESMs as listed in table S1.

# **Figure Descriptions**

Figures S1 to S8 provide additional support and insights on the findings detailed in the main text. The  $F_{TOA}$  response of fixing CDNC to 150 cm<sup>-3</sup> for Atlantic and 75 cm<sup>-3</sup> in other experiments is shown in Figure S1. Fixing CDNC produces a reduction of 4-12 Wm<sup>-2</sup> in  $F_{TOA}$  radiation over the SH subtropical region (panels e-h Figure S1). Figure S2 shows the existing biases in the CESM2 with GPCP for different monsoon regions of the world. The CESM2 have biases in simulating climatological mean precipitation for other monsoon regions as well. Figures S3 to S7 show precipitation response of the radiation reduction experiments over other monsoon regions. For West African monsoon

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during May-June and July to September, the Atlantic experiment improves the Sahel precipitation bias by small amount while as other experiments have negligible effects (Figures S3 and S4). All the experiments show worsening of biases of southern hemisphere monsoons such as Australian monsoon and East/South African monsoon in December-February, (Figures S5 and S6) with exception of South American monsoon for which the Atlantic experiment shows improvement in precipitation biases over Brazil (Figure S7). Figure S8 depicts the relationship between Root Mean Square Errors (RMSE, calculated with GPCP for precipitation and CERES-EBAF for radiation) in net TOA and SASM precipitation for CMIP6 multimodel ensemble. The models having higher RMSE in SASM precipitation show errors in SH subtropical net TOA as well.

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CMIP6 model	Ensemble member	CMIP6 model	Ensemble member
ACCESS-CM2	r1i1p1f1	E3SM-1-1	r1i1p1f1
ACCESS-ESM1-5	r1i1p1f1	E3SM-1-1-ECA	r1i1p1f1
AWI-CM-1-1-MR	r1i1p1f1	EC-Earth3-Veg-LR	r1i1p1f1
AWI-ESM-1-1-LR	r1i1p1f1	FGOALS-g3	r1i1p1f1
BCC-CSM2-MR	r1i1p1f1	GFDL-CM4	r1i1p1f1
BCC-ESM1	r1i1p1f1	GFDL-ESM4	r1i1p1f1
CESM2	r1i1p1f1	GISS-E2-1-G-CC	r1i1p1f1
CESM2-FV2	r1i1p1f1	INM-CM4-8	r1i1p1f1
CESM2-WACCM	r1i1p1f1	IPSL-CM6A-LR	r1i1p1f1
CESM2-WACCM-FV2	r1i1p1f1	MIROC-ES2L	r1i1p1f2
CMCC-CM2-SR5	r1i1p1f1	MIROC6	r1i1p1f1
CNRM-CM6-1	r1i1p1f2	MPI-ESM1-2-LR	r1i1p1f1
CNRM-CM6-1-HR	r1i1p1f2	MRI-ESM2-0	r1i1p1f1
CNRM-ESM2-1	r1i1p1f2	NESM3	r1i1p1f1
CanESM5-CanOE	r1i1p2f1	NorESM2-MM	r1i1p1f1
TaiESM1	r1i1p1f1	SAM0-UNICON	r1i1p1f1

Table S1.CMIP6 models used in this study



Figure S1. Fixed-SST response: Applied CDNC perturbations (cm<sup>-3</sup>; a-d) and annual mean changes in  $F_{TOA}$  radiation (W m<sup>-2</sup>; e-h) in fixed-SST simulations over different regions - (a,e) Atlantic, (b,f) Indian Ocean, (c,g) Pacific and (d,h) Zonal. Only significant (p < 0.05) changes are shown in colour.



Figure S2. Existing biases in CESM2: Ensemble mean bias of CESM2 Precipitation with GPCP (mm/day) for a. June-September (JJAS) over South Asia, b. May-June (MJ) over West Africa, c. July-September (JAS) over West Africa, d. December-February (DJF) over Australia, e. DJF over South Africa and f. December-March (DJFM) over South America.



Figure S3. West African precipitation response: MJ precipitation response (mm/day) (at simulation period 121-150 years) in fully coupled simulation for experiments a. Atlantic, b. Indian Ocean, c. Pacific and d. Zonal. Non-significant responses (at level of significance 0.05) are masked in white.





Figure S4. West African precipitation response: JAS precipitation response (mm/day) (at simulation period 121-150 years) in fully coupled simulation for experiments a. Atlantic, b. Indian Ocean, c. Pacific and d. Zonal. Non-significant responses (at level of significance 0.05) are masked in white.

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**Figure S5.** Australian monsoon precipitation response: DJF precipitation response (mm/day) (at simulation period 121-150 years) in fully coupled simulation for experiments a. Atlantic, b. Indian Ocean, c. Pacific and d. Zonal. Non-significant responses (at level of significance 0.05) are masked in white.





Figure S6. East African monsoon precipitation response: DJF precipitation response (mm/day) (at simulation period 121-150 years) in fully coupled simulation for experiments a. Atlantic, b. Indian Ocean, c. Pacific and d. Zonal. Non-significant responses (at level of significance 0.05) are masked in white.

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Figure S7. South American monsoon precipitation response: DJFM precipitation response (mm/day) (at simulation period 121-150 years) in fully coupled simulation for experiments a. Atlantic, b. Indian Ocean, c. Pacific and d. Zonal. Non-significant responses (at level of significance 0.05) are masked in white.



Figure S8. Regression of net TOA Root Mean Square Errors (RMSE) (W  $m^{-2}$ ) and SASM precipitation RMSE (mm day<sup>-1</sup>) in the CMIP6 multi-model ensemble evaluated when SASM precipitation RMSE is two standard deviations above the multi-model mean.