



1 **Quantifying snow-darkening and atmospheric radiative effects of black carbon and**  
2 **dust on the South-Asian Monsoon and hydrological cycle: Experiments using**  
3 **variable resolution CESM**

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5 Stefan Rahimi<sup>1</sup>, Xiaohong Liu<sup>1,\*</sup>, Chenglai Wu<sup>1,2</sup>, William K. Lau<sup>3</sup>, Hunter Brown<sup>1</sup>, Mingxuan Wu<sup>1</sup>, Yun Qian<sup>4</sup>  
6

7 <sup>1</sup>Department of Atmospheric Science, University of Wyoming, 1000 E. University, 1000 E. University Ave. Laramie,  
8 WY 82071

9 <sup>2</sup>International Center for Climate and Environment Sciences, Institute of Atmospheric Physics, Chinese Academy of  
10 Sciences, Beijing, China 100029

11 <sup>3</sup>Earth System Science Interdisciplinary Center, University of Maryland, College Park, MD, 20742

12 <sup>4</sup>Atmospheric Sciences and Global Change, Pacific Northwest National Laboratory, P.O. Box 999, Richland, WA  
13 99352

14  
15 \*Correspondence to: Xiaohong Liu ([xliu6@uwyo.edu](mailto:xliu6@uwyo.edu))

16 **Abstract**

17 Black carbon (BC) and dust impart significant effects on the south-Asian monsoon  
18 (SAM), which is responsible for ~80% of the region's annual precipitation. This study  
19 implements a variable-resolution (VR) version of Community Earth System Model (CESM) to  
20 quantify the impacts of absorbing BC and dust on the SAM. This study focuses on the snow  
21 darkening effect (SDE), as well as how these aerosols interact with incoming and outgoing  
22 radiation to facilitate an atmospheric response (i.e., aerosol radiation interactions (ARI)). By  
23 running sensitivity experiments, the individual effects of SDE and ARI are quantified, and a  
24 theoretical framework is applied to assess these aerosols' impacts on the SAM. It is found that  
25 ARI of absorbing aerosols warm the atmospheric column in a belt coincident with the May-June  
26 averaged location of the subtropical jet, bringing forth anomalous upper-tropospheric (lower-  
27 tropospheric) anticyclonogenesis (cyclonogenesis) and divergence (convergence). This anomalous  
28 arrangement in the mass fields brings forth enhanced rising vertical motion across south Asia and  
29 a stronger westerly low-level jet, the latter of which furnishes the Indian subcontinent with  
30 enhanced Arabian Gulf moisture. This leads to precipitation increases of +2 mm d<sup>-1</sup> or more  
31 across much of northern India from May through August, with larger anomalies in the western  
32 Indian mountains and southern TP mountain ranges due to orographic and anabatic enhancement.



33 Across the Tibetan Plateau foothills, SDE by BC aerosols drives large precipitation anomalies of  
34  $>6 \text{ mm d}^{-1}$ , comparable to ARI of absorbing aerosols from April through August. Runoff changes  
35 accompany precipitation and Tibetan Plateau snow changes, which have consequences for south-  
36 Asian water resources.

37

### 38 1. Introduction

39

40 The South-Asian Monsoon (SAM) and Tibetan Plateau (TP) snow cover are critical to the  
41 security of water resources across India, Pakistan, and Bay of Bengal-region countries.  
42 Developing from June through early September, the thermally driven SAM provides the region  
43 with  $\sim 80\%$  of its annual precipitation (Bookhagen and Burbank, 2010; Hasson et al., 2013). This  
44 precipitation, together with seasonal snowmelt from Tibet, serves to replenish major waterways  
45 across the region. Southern Asia has a very high population density and is in a state of rapid  
46 industrialization. As a result, large amounts of black carbon (BC) particles are emitted to the  
47 atmosphere. BC can modify the premonsoonal and monsoonal system by perturbing the regional  
48 radiative balance (Flanner et al., 2007; Qian et al., 2009, 2011; Lau et al., 2010). Additionally,  
49 southern Asia's proximity to major dust emission sources makes this region's climate system  
50 susceptible to dust effects (Vinoj et al., 2014; Jin et al., 2016).

51 Various studies have shown that absorbing BC and dust (referred to collectively as BCD) can  
52 impart significant perturbations on the earth's radiative balance and climate globally (Jacobsen,  
53 2001; Koch, 2001; Flanner et al., 2007; Xu et al., 2016) and regionally (Quinn et al., 2008; Qian  
54 et al., 2009; Painter et al., 2010, 2012; Zhao et al., 2014; Jin et al., 2016; Wu et al., 2018),  
55 resulting in changes in temperature, cloud fraction, precipitation, snow cover, and runoff. BCD  
56 have been shown to have a particularly strong impact on the south-Asian monsoon (Lau et al.,  
57 2010, 2017; Qian et al., 2011; Das et al., 2015) through a variety of pathways. For instance,  
58 atmospheric BCD can increase the amount of absorbed solar energy across snow-covered regions  
59 when deposited on ice, leading to increased melting rates in a process known as the snow



60 darkening effect (SDE; Qian et al., 2015). Furthermore, atmospheric BCD aerosols absorb and  
61 scatter incoming sunlight, altering the thermodynamic structure of the atmosphere. Dust interacts  
62 with longwave radiation to alter atmospheric thermodynamics further (Seinfeld et al., 2004; Zhao  
63 et al., 2011). These aerosol-radiation interactions (ARI) describe the explicit heating/cooling of  
64 the atmosphere by attenuating aerosols (direct effects), as well as how the atmosphere circulations  
65 may be changed to influence cloud formation (semi-direct effects).

66 Southern Asia is especially susceptible to SDE and ARI during the spring and summer for  
67 several reasons. First, BCD burdens increase during this time period, contributing to stronger  
68 perturbations in the region's radiative balance. Second, the solar zenith angle is reduced, and  
69 higher intensity sunlight warms the region; the more direct sunlight amplifies the radiative  
70 perturbations brought forth by BCD. Third, the highly elevated TP remains snow-covered for  
71 large fractions of the year and lies directly north and east of BCD sources, respectively, making  
72 this region vulnerable to BCD SDE (Qian et al., 20011; Lau et al., 2017).

73 The warm season evolution of SAM is quite complex (Boos and Kuang, 2010; Wu et al.,  
74 2014). During the spring and summer, unabated heating of the Indian peninsula brings forth the  
75 establishment of the monsoon trough beneath attendant upper-tropospheric anticyclogenesis (the  
76 Tibet High). Additionally, a westerly low-level jet (WLLJ) forms, which is responsible for  
77 transporting copious amounts of moisture into south Asia from the Arabian Sea. The heating is  
78 due to the presence of the zonally-oriented mountain ranges of the southern TP, which effectively  
79 block cold air intrusions into the Indian subcontinent associated with mid-latitude cyclones (Boos  
80 and Kuang, 2010). The rising branch of the monsoonal circulation that develops is moisture laden,  
81 contributing to deep convection from June through September. This rainfall, combined with  
82 seasonal snowmelt across the TP, replenishes main waterways across southern, central, and  
83 eastern Asia, providing water resources for billions.

84 BCD have been shown to warm the TP via SDE and ARI, with maximum warming and  
85 snowmelt during the late spring (Lau et al., 2006; Lau et al., 2010; Qian et al., 2011; Vinoj et al.,



86 2014; Lau et al., 2017). Eastward dust transport from the Middle East in addition to BC transport  
87 from India warm the atmospheric column across south central Asia leading to low-level relative  
88 vorticity spin-up. The alignment of this low-level feature beneath the Tibet High brings forth an  
89 intensification of the WLLJ, and more moisture is transported from the Arabian Sea into southern  
90 Asia. The increased moisture amounts collocate with the rising branch of the SAM, and  
91 precipitation amounts are increased while surface temperatures cool (Vinoj et al., 2014; Jin et al.,  
92 2016). Furthermore, the warming associated with enhanced snow melting across Tibet enhances  
93 this circulation change by increasing the rate of column warming.

94 While many studies have attempted to model the BCD-induced perturbations on  
95 premonsoonal (May and June) and monsoonal (June through August) climate and hydrology,  
96 several opportunities for scientific understanding and advancement still exist as far as quantifying  
97 their regional climate impacts. Firstly, many SAM-aerosol studies have utilized horizontal grid  
98 spacings ( $\Delta x$ ) in excess of 100 km (e.g. Lau et al., 2010; Qian et al., 2011; Xu et al., 2016). While  
99 these grid spacings are generally adequate for resolving large-scale meteorological signatures,  
100 simulations with such coarse grid spacing may entirely fail to capture mesoscale precipitation  
101 systems whose latent energy helps to regulate the SAM. For studies that have utilized smaller grid  
102 spacings (e.g. Das et al., 2015; Jin et al., 2016; Lau et al., 2017), limited area models (LAMs)  
103 were used for computational practicality. However, LAMs require prescribed boundary  
104 conditions for both the meteorological, chemical, and aerosol fields. Uncertainties associated with  
105 these fields, in addition to inconsistencies between reanalysis and simulation physics, can lead to  
106 uncertainties in quantifying SDE and ARI effects on the SAM. In addition, LAMs by their very  
107 nature prevent two-way interactions between the small-scale features of the inner domain and the  
108 large-scale features of the reanalysis grid.

109 In this study, we employ a variable resolution (VR) version of the Community Earth System  
110 Model (CESM) to quantify the effects of BC- and dust-induced SDE and ARI on SAM dynamics.  
111 This relatively new modeling approach allows for a model grid spacing of  $0.125^\circ$  ( $\sim 14$  km) across





112 the TP, which transitions to a  $1^\circ$  mesh outside of south Asia. A suite of sensitivity experiments is  
113 conducted to quantify BCD-induced SDE and ARI changes in premonsoonal and monsoonal  
114 climate and hydrology. By implementing this VR version of CESM in this way, we are able to: i)  
115 bypass the need for boundary conditions, homogenizing the physics and chemistry  
116 parameterizations across the entire model domain, ii) decrease the model grid spacing over the  
117 most complicated terrain of southern and central, which has been identified as being critical to  
118 SAM dynamics and evolution, and iii) estimate the relative importance of the SDE compared to  
119 ARI in affecting the premonsoonal and monsoonal environment for BC and dust separately.

120 This manuscript is organized as follows. Section 2 provides a methodology for our  
121 experimental design. This is followed by an aerosol validation in section 3, in which we compare  
122 simulated to observed aerosol optical depth (AOD), as well as simulated in-snow and in-  
123 atmosphere BC concentrations to observations. Simulated meteorological and hydrological  
124 perturbations due to various effects by BC and dust are presented in section 4. A theoretical  
125 framework is presented in section 5 to unify the simulated changes in meteorology,  
126 thermodynamics, and hydrology. Concluding remarks and future work are discussed in section 6.

127

## 128 **2. Model**

### 129 **2.1 Model configuration**

130 The VR grid, which refines to  $0.125^\circ$  horizontal grid spacing ( $\sim 14$  km) across south-central  
131 Asia, transitions to  $0.25^\circ$ , followed by  $0.5^\circ$  and eventually  $1^\circ$  (Figure 1a). The region of the  
132 analysis grid that is characterized by horizontal grid spacings less than  $1^\circ$  is located  
133 approximately between  $60^\circ\text{E}$ - $120^\circ\text{E}$  and  $5^\circ\text{N}$ - $55^\circ\text{N}$ . This encompasses all of India, the Bay of  
134 Bengal, and the Arabian Sea. Most of India and the Bay of Bengal are characterized by grid  
135 resolutions greater than  $0.125^\circ$ . As in Zarzycki et al. (2013), the spectral element dynamic core is  
136 used in the Community Atmosphere Model (CAM-SE), the atmospheric component of CESM, to



137 solve the primitive hydrostatic equations on a fully unstructured quadrilateral mesh (Dennis et al.,  
138 2012). In addition to atmospheric fields, land surface fields are also treated on the same VR grid.  
139 VR and uniform (UN) 1°-resolution experiments are run to explore the sensitivity of BCD  
140 effects to model grid spacing. The number of horizontal computational grid cells increases from  
141 48,602 in the UN experiment to 114,860 in the VR experiment. CESM experiments are  
142 conducted on 30 vertical levels, however the physics time step in VR simulations is two times  
143 smaller than that used in UN experiments (15 min compared to 30 min) to avoid numerical  
144 instability. Additionally, the dynamics time step is 9 s in VR experiments and 90 s in UN  
145 experiments. The grid setup between the VR and UN experiments is identical to that used in  
146 Rahimi et al. (2018).

147

## 148 **2.2 Model physics**

149 Both VR and UN simulations are conducted using identical physics. CESM version 1.2 is  
150 used with CAM version 5.3 (Neale et al., 2010) and is coupled with the Community Land Model,  
151 version 4.0 (CLM4). Coupled into CLM4 is the SNow ICE and Aerosol Radiative (SNICAR)  
152 model, which prognostically treats the snow aging process, the effects of BC and dust deposition  
153 on snowpack absorption, and meltwater scavenging (Flanner et al., 2007). Parameterized cloud  
154 microphysics from Morrison and Gettleman (2008), shallow convection from Park and Bertherton  
155 (2009), and deep convection from Zhang and McFarlane (1995) and Richter and Rasch (2008) are  
156 used. Radiation is treated using the rapid radiative transfer model from Iacano et al. (2008), and  
157 aerosol impacts are simulated using the three-mode version of the Modal Aerosol Module  
158 (MAM3), described in Liu et al. (2012).

159

## 160 **2.3 Model experiments**

161 Eight VR and four UN experiments are run to estimate the impacts of BCD-induced SDE and  
162 ARI on premonsoonal and monsoon climate fields (Table 1). Aside from the control experiments,



163 defined to be CONT-vr and CONT-un for the VR and UN control experiments, respectively, 7  
164 perturbation VR experiments are run to quantify various BCD effects. First, an experiment is run  
165 with BCD SDE turned off (noSDE-vr). This is achieved by setting the on-snow BCD deposition  
166 fluxes to zero in CLM4. Second, an experiment is run in which BCD ARI is turned off (noARI-  
167 vr). This is achieved by excluding the BCD volume from the calculation of bulk aerosol  
168 extinction, asymmetry parameter, and single scatter albedo in CAM. Third, an experiment is run  
169 in which both BCD SDE and ARI are removed (noBCD-vr). A fourth and fifth perturbation  
170 experiment are run, identical to noSDE-vr and noARI-vr, but only the BC-induced SDE and ARI  
171 are removed, respectively (noBCSDE-vr and noBCARI-vr, respectively). The sixth and seventh  
172 VR perturbation experiments are identical to noBCSDE-vr and noBCARI-vr, except dust SDE  
173 and ARI are removed, respectively (noDSDE-vr and noDARI-vr, respectively). Perturbation  
174 experiments are also run to assess BCD effects on a 1° mesh: noSDE-un, noARI-un, and noBCD-  
175 un.

176 The BCD effects on meteorological and hydrological fields can be found by subtracting the  
177 perturbation experiments from the control simulation (CONT-vr). For some variable  $x$ , the effect  
178 induced by a specific species (BC or dust) at a gridcell can be computed:

179

$$\text{Change in } x \text{ induced by SDE} \quad \text{SDE}_x = \text{CONT-vr}_x - \text{noSDE-vr}_x, \quad (1)$$

$$\text{Change in } x \text{ induced by ARI} \quad \text{ARI}_x = \text{CONT-vr}_x - \text{noARI-vr}_x, \quad (2)$$

$$\text{Change in } x \text{ induced by SDE+ARI} \quad \text{TOTAL}_x = \text{CONT-vr}_x - \text{noBCD-vr}_x, \quad (3)$$

$$\text{Change in } x \text{ induced by BC SDE} \quad \text{BCSDE}_x = \text{CONT-vr}_x - \text{noBCSDE-vr}_x, \quad (4)$$

$$\text{Change in } x \text{ induced by BC ARI} \quad \text{BCARI}_x = \text{CONT-vr}_x - \text{noBCARI-vr}_x, \quad (5)$$

$$\text{Change in } x \text{ induced by dust SDE} \quad \text{DSDE}_x = \text{CONT-vr}_x - \text{noDSDE-vr}_x, \quad (6)$$

$$\text{Change in } x \text{ induced by dust ARI} \quad \text{DARI}_x = \text{CONT-vr}_x - \text{noDARI-vr}_x, \quad (7)$$

180



181 The anomalies computed in (4) through (7) for variable  $x$  are linear, but may add together  
182 such that their combined effect is nonlinear. In other words, there are nonlinear interactions  
183 between BC and dust SDE and ARI that may be important when considering the perturbations to  
184 SAM climate. We emphasize the VR results, but we do briefly discuss their differences with UN  
185 results in Section 5.

186 Each individual experiment is run for 11 years, and the first year in each simulation is  
187 neglected in the analysis to allow for “spin up”. Climatological sea surface temperature, sea ice,  
188 and anthropogenic aerosol and precursor gas emissions for the year 2000 are used. After  
189 comparing the simulations to both gridded and point source (locations shown in Figure 1b)  
190 reference data, the means of various climate variables from the last 10-year of simulations are  
191 computed to evaluate the impacts of BCD-induced SDE and ARI across southern Asia.  
192 Furthermore, all simulated data (both VR and UN) across the region are interpolated to an  
193 identical  $0.125^\circ$  rectilinear grid for direct comparison.

194 For our analysis, we break up southern Asia into 5 distinctive subregions (Figure 1c). We  
195 consider most of India separately from the Indo-Gangetic Plain (IGP) to gain a sense of the  
196 impacts exerted on the SAM by the TP regional aerosol effects. We also divide the TP into the  
197 western TP (WTP) and eastern TP (ETP) since other studies (Lau et al., 2010; Qian et al., 2011)  
198 have found there to be noticeable differences in simulated aerosol effects between these two  
199 regions. A fifth region, the TP foothills (TPF), is also considered to explore how BCD effects  
200 may impact orographic precipitation.

201 A validation of the simulated meteorology was performed in Rahimi et al. (2018), in which  
202 CONT-vr and CONT-un were compared to surface- and satellite-based datasets. They found that  
203 there were marked improvements in the simulated temperature, precipitation, and snow coverage  
204 across the TP and its southern mountain ranges when using a VR grid. This is important when  
205 simulating the SDE, which is fundamentally dependent on the spatial distribution of snow  
206 coverage. In this study, we further evaluate CESM’s performance in simulating aerosols, as the



207 SDE and ARI are fundamentally dependent on the spatial variability and magnitude of aerosol  
208 loading.

209

### 210 **3. Evaluation of Simulated Aerosols**

#### 211 **3.1 Observational data**

212 To accurately capture the impacts of SDE and ARI on south-Asian climate, it is  
213 important for the simulations to adequately represent atmospheric aerosol characteristics, e.g.  
214 aerosol optical depth (AOD). We make use of satellite and ground-based AOD measurements, as  
215 well as point source BC measurements to evaluate the model performance. While in-atmosphere  
216 and in-snow BC measurements are available for model performance evaluation, in-snow and in-  
217 atmosphere dust measurements are lacking due to uncertainties in how dust is classified and the  
218 lack of data across southern Asia. These data and their respective uncertainties will now be  
219 presented.

220 Control simulations are compared to 1° and 0.5° Level-3 550 nm and 555 nm AOD data  
221 from the MODerate-resolution Imaging Spectroradiometer (MODIS, Platnick et al., 2015) and the  
222 Multi-angle Imaging SpectroRadiometer (MISR, MISR Science team 2015), respectively.  
223 MODIS data are averaged monthly from 2001 to 2014, while MISR data are similarly averaged  
224 from 2002 to 2014. More information on MODIS and MISR can be found in Section S1.1 and  
225 Section S1.2 of supplemental materials, respectively.

226 AOD spectral radiometers measurements from 57 aerosol robotic network (AERONET)  
227 sites across central and southern Asia are used to evaluate simulated total, coarse-, and fine-mode  
228 AOD at 500 nm (Holben et al., 1998). Results were interpolated to AERONET site locations and  
229 averaged spatially. The simulated fine-mode AOD is the summation of the accumulation- and  
230 Aitken-mode AOD variables in CESM. AERONET site locations are denoted by the black  
231 crosshatches in Figure 1b.



232 AOD information is available via the Max Planck Institute's Aerosol Climatology,  
233 version 2 (MACv2), providing climatological estimates of monthly mean AOD at 550 nm on a 1°  
234 rectilinear grid (Stevens et al., 2017). More information on the MACv2 product can be found in  
235 Section S1.3.

236 We compare our results against AOD data from the Modern-Era Retrospective analysis  
237 for Research and Applications version 2 (MERRA-2, Global Modeling and Assimilation office,  
238 2015). Monthly 550 nm MERRA-2 data are available on a 0.5° by 0.625° grid and analysis is  
239 performed using data from 1980-2017. More information on the MERRA-2 product can be found  
240 in Section S1.4.

241 Finally, simulations are compared to point atmospheric BC measurements from 13 sites  
242 discussed in He et al. (2014) and 11 sites discussed in Yang et al. (2018). The site locations are  
243 shown in Figure 1b. Additionally, 26 measurements of in-snow BC, as discussed in He et al.  
244 (2014) are used to evaluate our simulations. Locations of in-snow BC measurements are also  
245 given in Figure 1b.

246

### 247 **3.2 AOD comparisons**

248 Focusing on annual AOD averaged between 0°N-60°N and 60°E-140°E, in which lie  
249 several major BCD emissions sources, CONT-vr and CONT-un simulate AOD values of 0.147  
250 and 0.143 respectively. On the other hand, MODIS, MISR, MACv2, and MERRA-2 depict  
251 regionally higher annual AOD values of 0.285, 0.202, 0.235, and 0.216, respectively. While both  
252 CESM simulations reasonably capture the global annually averaged AOD compared to satellite  
253 observations (see Section S1.5), they do not capture the generally larger annual AOD values  
254 across south-central Asia. Because the buildup of aerosols across southern Asia has been  
255 identified to affect the premonsoonal and monsoonal properties (Lau et al., 2011; 2017), the  
256 CESM simulations' underprediction of annual AOD by almost a factor of 2 across south Asia  
257 must be kept in mind.



258           Figure 2 and Figure S1 show the spatial distribution of Asian May and June (MJ)  
259 averaged and annually averaged AOD, respectively, from simulations, satellite measurements,  
260 and aerosol reanalysis. Data reveal higher MJ AOD values across southern Asia, the Tarim Basin  
261 eastward into northern China, and eastern China. The lowest AOD values are located across the  
262 TP, Russia, and northwestern Micronesia. CONT-vr and CONT-un overpredict AOD values over  
263 Asian dust sources (the Taklamakan and Gobi deserts) and the northern Arabian Sea, and  
264 simulations underpredict AOD values across India and eastern China compared to MISR, MODIS,  
265 MACv2, and MERRA-2 data.

266           Figure 3(a-c) shows simulated versus observed annually averaged AOD from 57  
267 AERONET sites (shown in Figure 1b). CESM experiments underpredict AOD compared to  
268 AERONET measurements. Fine mode aerosols contribute to most of the AOD underprediction.  
269 CONT-vr simulates a fine mode and coarse mode mean AOD bias ( $\overline{\Delta AOD}$ ) of -0.189 and -0.062,  
270 respectively, while CONT-un simulates similar biases. Furthermore, CONT-vr slope values for  
271 the total, fine-mode, and coarse-mode best-fit lines of 0.157, 0.135, and 0.402, respectively,  
272 reveal that the underprediction of AOD increases for larger AERONET AOD values.

273           AOD underprediction by CONT-vr and CONT-un is also evident in a time series  
274 depicting the monthly variability of AOD averaged over the AERONET sites (Figure 3d).  
275 Simulations and observations generally show a similar pattern, with larger (smaller) AOD values  
276 in the warm (cold) months. CESM simulations underpredict AERONET AOD by a factor of  
277  $\sim 1.5$ . MISR, MACv2, and MERRA-2 appear to best agree with AERONET observations, while  
278 MODIS generally shows monthly AOD values that are 10-30% higher than AERONET  
279 observations. Additionally, CONT-vr, CONT-un, MERRA-2, and MISR correlate better with  
280 AERONET than MODIS and MACv2, with CONT-vr having the highest  $r$ -value of 0.874.

281           Figure 3e shows a time series of annually averaged monthly AOD interpolated to 18  
282 AERONET sites across the IGP and TPF between 25°N-30°N and 70°E-90°E. Similar to the 57-  
283 site average, both CESM simulations underpredict monthly AOD by a factor of 1.5 to 2 across the



284 IGP and TPF. However, in contrast to the 57-site average, the 18-site  $r$ -values of the CESM  
285 simulations with AERONET are notably lower (see Sec. S1.5).

286 The underestimation of AOD by the CESM experiments may be tied to biased  
287 precipitation simulated by CONT-un and CONT-vr (Rahimi et al., 2018). This could have the  
288 effect of overpredicting wet scavenging of aerosols. This idea is reinforced when looking at point  
289 BC measurements across India, where simulated BC wet deposition dominates over dry  
290 deposition (Figure S2), and simulated BC concentrations are much smaller than observations  
291 (Figure 4a); simulations may be washing out too many aerosols. Finally, incorrect emissions  
292 maps may further contribute to the simulated bias in aerosol amounts (Zhao et al., 2011; Fan et al.,  
293 2018).

294

### 295 3.3 Surface BC comparisons

296 Figure 4a depicts an almost unanimous underprediction of in atmospheric BC  
297 concentrations across the 24 measurement sites by the CESM experiments, with average aerosol  
298 biases of  $-2.77 \mu\text{g m}^{-3}$  and  $-2.76 \mu\text{g m}^{-3}$  for CONT-vr and CONT-un, respectively. The largest  
299 underpredictions occur over urban sites such as Delhi (observed  $13.5 \mu\text{g m}^{-3}$ ), Dibrugarh  
300 (observed  $8.9 \mu\text{g m}^{-3}$ ), and Lhasa (observed  $3.7 \mu\text{g m}^{-3}$ ). Averages of observations depict a mean  
301 concentration across all sites of  $3.24 \mu\text{g m}^{-3}$ , while CONT-vr and CONT-un underpredict this  
302 value by a factor of 6.9 and 6.8, respectively. Additionally, several sites see a simulated  
303 underprediction of BC concentrations by more than a factor of 10.

304 The widespread underprediction of atmospheric BC does not necessarily translate to an  
305 underprediction of in-snow BC mixing ratio as seen in Figure 4b. CONT-vr and CONT-un  
306 simulate a bias of  $-10.4 \mu\text{g kg}^{-1}$  and  $+22.7 \mu\text{g kg}^{-1}$ , respectively, when comparing to the station-  
307 averaged BC mixing ratio of  $54.6 \mu\text{g kg}^{-1}$ . This indicates that CONT-vr is more comparable to  
308 observations magnitude-wise. It is also noteworthy that several sites showcase a CONT-vr





309 simulated in-snow BC mixing ratio that is an order of magnitude different from that simulated in  
310 CONT-un. These large differences in simulated in-snow BC between the VR and UN  
311 experiments can be attributed to large meteorological and terrain differences between the two  
312 experiments, especially for He et al. (2014) sites across the Himalaya Mountains. For instance,  
313 Fig. S2 shows that at East Rongbuk (28.02°N, 86.96°E), CONT-vr simulates the terrain height to  
314 be more than 2.5 km higher than CONT-un, culminating in lower monthly temperatures and snow  
315 water equivalent (SWE) increases of more than 300 mm compared to CONT-un. Despite the  
316 increased SWE at Rongbuk, CONT-vr simulates tens of millimeters less precipitation than  
317 CONT-un owing to the smaller south-to-north upslope zone simulated in the VR experiment  
318 (Rahimi et al., 2018). The smaller VR-simulated precipitation correlates with 100-300  $\mu\text{g m}^{-2} \text{d}^{-1}$   
319 less wet-scavenged BC compared to the CONT-un experiment. The decreased wet deposition,  
320 coupled with the larger SWE amounts in CONT-vr thus favors lower in-snow BC mixing ratios  
321 than CONT-un.

322

#### 323 4. Climatic Effects of BC and Dust

324 Evaluation of the aerosol SDE and ARI is performed by examining the differences of the  
325 VR perturbations from the VR control experiment as discussed in Section 2.3. Changes in 2-  
326 meter temperature, snow water equivalent (SWE), radiative effect (RE), cloud coverage, specific  
327 humidity, precipitation, and runoff are discussed in this section to motivate the application of a  
328 simple theoretical dynamical framework that describes the impacts of BCD on premonsoonal and  
329 monsoonal climate. Only results of the VR simulations are discussed in this section, and a brief  
330 comparison of VR and UN results is given in Section 5.

331

#### 332 4.1 Temperature

333 Together, BCD contribute to statistically significant (SS;  $t$ -values in excess of 0.9) ARI-  
334 and SDE-induced 2-meter temperature ( $T_2$ ) changes during MJ and July and August (JA) across



335 Tibet and south Asia. Warming in excess of 3°C across the WTP and 1.3°C across the ETP is  
336 shown in Figure 5a. Meanwhile, the collective impacts (ARI+SDE) of BCD contribute to cooling  
337 across most of India due to cloud coverage increases (to be discussed in section 4.4), with SS  
338 values of -0.7°C to -1.2°C across western India and the IGP region. By JA, BCD effectuates T2  
339 patterns similar to those in MJ across southern Asia, as shown in Fig. S3a. However, the areas of  
340 cooling characterizing much of India are shifted north and west to include most of Pakistan, and  
341 the areas of warming characterizing a majority of the TP are much reduced compared to MJ,  
342 especially across the southern TP.

343 The sign and magnitude of T2 changes vary as a function of effect type (SDE or ARI)  
344 and by species (BC or dust). BCD SDE-induced T2 warming can exceed ARI T2 warming over  
345 complex terrain, as indicated by our results. Values of +2°C are simulated across the WTP  
346 mountain chains (Figure 5b), such as the western Himalaya, Kunlun, Karakoram, and Hindu-  
347 Kush Mountains during MJ. Within the WTP, SDE-induced warming reaches 0.5°C to 1.3°C.  
348 Additionally, BC generally contributes to more simulated SDE warming than dust across the  
349 WTP during MJ and JA, while dust contributes to a majority of the SDE-induced warming across  
350 the ETP during MJ.

351 BCD ARI drive a majority of the simulated T2 changes across southern Asia during MJ  
352 (Figure 5c) and JA, with the largest T2 changes occurring during MJ. While MJ changes in T2  
353 associated with ARI are brought forth by both BCD collectively, the TP area has more expansive  
354 T2 changes as a result of dust-induced ARI (Figure 5g); a much larger swath of 1.3°C to 2°C  
355 warming occurs across the WTP compared to the simulated BC-induced ARI (Figure 5e). Across  
356 the IGP, dust ARI brings forth cooling of more than 0.7°C during MJ, while BCD ARI cool  
357 portions of central and southern India by 0.5°C to 0.7°C during MJ. By JA, dust ARI (BC ARI)  
358 contributes to most of the cooling (warming) across the IGP and Pakistan (northern TP), with  
359 simulated T2 changes of from -0.2°C to -0.7°C (+0.2 to +1.3°C). With these effects in mind, it  
360 should be noted that BC is underestimated compared to surface observations, while dust, which



361 comprises a significant component of the coarse mode AOD, is significantly overestimated  
362 compared to AERONET observations.

363 While understanding the BCD-induced changes in T2 is important from an anthropogenic  
364 perspective, these changes are inadequate when examining the influence of BCD on SAM  
365 dynamics. This is because BCD-induced SAM changes depend on the thermal characteristics of  
366 the tropospheric column. For this reason, section 5 will make use of the 300-700 hPa mean  
367 column temperature differences instead of T2 differences when examining the circulation  
368 changes brought about by BCD effects on the SAM.

369

#### 370 **4.2 SWE**

371 BCD effects contribute to large reductions in SWE across the TP and TPF from April  
372 through June. Peak BCD-induced SDE plus ARI reductions in SWE of 75 mm are simulated in  
373 May across the WTP (Figure 6a), while peak reductions of 25 mm occur in April across the ETP  
374 and TPF (Figure 6b, 6c). Even though the largest regionally averaged T2 warming results from  
375 BCD-induced ARI, the largest reductions in SWE are due to simulated BCD-induced SDE. By  
376 June, BCD SDE contributes to SWE reductions across the WTP of greater than 50 mm, while  
377 BCD ARI contributes to reductions in SWE of 10 mm or less for all months.

378 BC SDE drives a majority of SWE changes across the TP with reductions of greater than  
379 30 mm or more across the WTP from March through June, but other effects are important too. BC  
380 ARI, dust ARI, and dust SDE all contribute to reduced SWE in excess of 10 mm from March  
381 through June across the WTP. Compared to the WTP, BCD effects across the ETP and TPF bring  
382 forth smaller reductions in SWE, but BC SDE still contributes to the largest effects on SWE.  
383 Additionally, the largest reductions in SWE are typically found along the western and southern  
384 TP periphery (Figure 7).

385 BCD effects that lead to changes in TP area SWE can directly impact runoff, which  
386 replenishes main waterways across the region. It is found that BCD SDE drive runoff increases



387 (decreases) from February through June (June through September) across the WTP and ETP, with  
388 peak runoff increases (decreases) of  $1.2 \text{ mm d}^{-1}$  ( $1 \text{ mm d}^{-1}$ ) across the WTP occurring in May  
389 (July) (see Fig. S4). The peak runoff increases across the WTP (ETP) correlate with maximum  
390 BCD SDE reductions in SWE of 75 mm (25 mm), which occur in May (April).

391

### 392 **4.3 Radiative effect**

393 All-sky direct radiative effect (DRE) and in-snow radiative effect (ISRE) diagnostics are  
394 computed online in CONT-vr. The all-sky DRE is computed by subtracting a diagnostically  
395 computed DRE without aerosols from the online DRE computed with aerosols present. ISRE is  
396 computed in the SNICAR code via a similar method. BCD have distinctly different radiative  
397 properties, and while both aerosols contribute to positive ISRE for all months across the TP  
398 region, their DREs are more complicated. BCD combines to incite a generally positive DRE  
399 across south Asia during MJ and JA, and the pattern of the DRE is similar during these time  
400 periods, respectively. The spatial distribution of BCD DRE during MJ is shown in Figure 8, while  
401 that for JA is shown in Fig. S5. Positive dust-induced DRE values of  $+4$  to  $+9 \text{ W m}^{-2}$  are  
402 simulated across the Tarim basin and the Gobi desert.

403 BC is simulated to exert unanimously positive DREs across southern and central Asia  
404 during MJ and JA. BC induces a DRE of between  $+1 \text{ W m}^{-2}$  and  $+4 \text{ W m}^{-2}$  across India during  
405 MJ (Figure 8a), with the magnitude weakening by JA (Fig. S5a). The highest BC DRE values of  
406 nearly  $+2 \text{ W m}^{-2}$  are found across the IGP in northern India, while DRE values of  $+0.5 \text{ W m}^{-2}$  or  
407 less are found across the TP during MJ into JA.

408 Dust is simulated to exert both positive and negative DREs across southern and central  
409 Asia during MJ and JA (Figure 8b). Dust effectuates a DRE of nearly  $+10 \text{ W m}^{-2}$  ( $+2 \text{ W m}^{-2}$ )  
410 during MJ south of the Gobi desert across the Ghar desert (north of the TP across the Tarim  
411 Basin). Dust-induced DRE values of between  $+1.5$  and  $+4 \text{ W m}^{-2}$  are also simulated across the  
412 central and northern Ghat Mountains of India during MJ. Meanwhile, dust induces a DRE of



413 around  $+1 \text{ W m}^{-2}$  and  $-1 \text{ W m}^{-2}$  across the TP and east-central India, respectively, during MJ.  
414 Areas with negative dust DRE values are typically characterized by low surface albedo; dust  
415 brightens the planetary albedo and thus cools the TOA. During JA, dust incites a DRE that is  
416 small (less than  $+1 \text{ W m}^{-2}$ ) across east-central India (Fig. S5b). Also during JA, dust-induced  
417 DRE values in excess of  $+4 \text{ W m}^{-2}$  are simulated across the northern Arabian Sea as dust is  
418 transported eastwards from Saudi Arabia by lower tropospheric westerlies.

419 Despite the most prominent SWE reductions occurring due to BC SDE, diagnostically  
420 computed ISRE values across the WTP indicate that BC and dust contribute to similar regionally  
421 averaged seasonal values of between  $+1 \text{ W m}^{-2}$  and  $+3 \text{ W m}^{-2}$  from March through June (Figure  
422 8c-e). Together, BC and dust contribute to an annual maximum ISRE of  $+7 \text{ W m}^{-2}$  across the WTP  
423 in May, while this maximum occurs in March across the ETP ( $+2.8 \text{ W m}^{-2}$ ) and TPF ( $+2.2 \text{ W m}^{-2}$ ).  
424 The ISRE maxima occur in boreal spring for all three TP regions as the solar elevation angle and  
425 south Asia BC and dust burdens increase during this time.

426 The largest ISRE values occur in MJ compared to JA, with the mountains on the southern  
427 and western TP periphery being characterized by ISRE values greater than  $+10 \text{ W m}^{-2}$ , locally  
428 (see Fig. S6). Furthermore, similar to the T2 patterns in Figure 5, there is a strong heterogeneity  
429 to the ISRE field across the TP relative to that in the 1-degree experiments (not shown).

430

#### 431 **4.4 Cloud coverage and moisture**

432 BCD effects bring forth responses in the mass fields which impact simulated SAM cloud  
433 fraction (CF) and specific humidity ( $q$ ). Driven primarily by BC and dust ARI,  $q$  increases in  
434 excess of  $1 \text{ g kg}^{-1}$  are simulated from mid-March through June across India, the IGP, and TPF;  
435 moisture changes peak during June across the IGP ( $+3.5 \text{ g kg}^{-1}$ ) and TPF ( $+2.5 \text{ g kg}^{-1}$ ) (Figure 9).  
436 BCD-induced SDE contribute to smaller  $q$  changes from April through July across the IGP and  
437 TPF of  $+0.7 \text{ g kg}^{-1}$  and  $+0.8 \text{ g kg}^{-1}$ , respectively. Interestingly, the  $q$  changes from dust-induced  
438 SDE and BC-induced SDE across the TPF and IGP do not add linearly during June (see Figure



439 **9cc.e**) Focusing specifically on the IGP, as the  $q$  changes are similar between the IGP and TPF,  
440 dust (BC) SDE contributes to a  $q$  change of  $+0.3 \text{ g kg}^{-1}$  ( $+1.2 \text{ g kg}^{-1}$ ), but we see a total (i.e., BC  
441 SDE + dust SDE)  $q$  change of  $+0.7 \text{ g kg}^{-1}$ . This result could be due to an increase in precipitation  
442 due to the combined effects of BCD SDE, which would act to deplete the available water vapor,  
443 locally. The positive  $q$  changes just discussed are dramatically reduced from July through  
444 September. Across the WTP and ETP,  $q$  changes of  $+1 \text{ g kg}^{-1}$  are simulated in June, while smaller  
445  $q$  changes are simulated for the spring and summer months across both regions.

446 Specific humidity increases due to BCD effects correlate reasonably well with increases  
447 in CF, especially across south Asia from April through September (see Fig. S7). An increase in  
448 CF, primarily driven by BCD ARI, is simulated across India, the IGP, and TPF from April  
449 through June. Net peak CF increases occur in May across India (12%) and June across the IGP  
450 (15%) and TPF (16%) as seen in Fig. S7. CF increases of 10% or more due to aerosol effects  
451 across southern Asia during MJ have been noted previously (Lau et al., 2010). By September, the  
452 CF increases vanish across India, the IGP, and TPF. Across the ETP, simulations indicate that CF  
453 increases of 3% are due to BCD SDE in June, with slightly larger increases (as much as 7%) in  
454 June and July across the WTP (see Fig. S7) also due to BCD SDE.

455 Compared to SDE, BCD ARI generally brings forth the largest changes in CF across  
456 south Asia from May through August. The spatial distributions of MJ (JA) CF changes are shown  
457 in Figure 10 (Fig. S8). CF increases of 7% or more across India and CF decreases of 5-10% or  
458 more across the central and northern TP are simulated (Figure 10a). The Arabian Sea, western  
459 Ghat Mountains, and TPF are characterized by the largest positive CF changes, which can exceed  
460 15%. Furthermore, while BC ARI patterns are similar to those of dust ARI during MJ, the  
461 magnitudes of dust-induced ARI changes are generally larger, especially across the Arabian Sea  
462 and the ETP (Figure 10e,g).

463 While SDE effect magnitudes are generally smaller compared to those induced by BCD  
464 ARI, BC SDE drives MJ CF changes of +5% to +15% across the IGP and southwestern TP (Fig.



465 10d). Meanwhile, CF reductions of 2% to 5% are simulated due to BC SDE across northern TP  
466 during MJ, making the spatial pattern of CF changes induced by BC SDE (Figure 10d) similar to  
467 those induced by BCD ARI (Figure 10c). In addition, dust SDE contributes to increases in CF of  
468 4% to 7% across the southern TP during MJ (see Figure 10f).

469

#### 470 **4.5 Precipitation**

471 BCD effects contribute to almost unanimously increased precipitation across southern  
472 Asia during premonsoonal months, as seen in Figure 11a, with BCD collectively contributing to  
473 values in excess of +6 mm d<sup>-1</sup> across the eastern Arabia Sea, the eastern Bay of Bengal, and the  
474 TPF. Elsewhere, BCD contributes to SS changes of between +1 mm d<sup>-1</sup> and +4 mm d<sup>-1</sup> across  
475 India, and changes in MJ precipitation of around +1 mm d<sup>-1</sup> over the southern and eastern TP. It  
476 seems as though the large-scale pattern in MJ precipitation changes is regulated by dust ARI  
477 (Figure 11g). However, precipitation changes of between +1.5 mm d<sup>-1</sup> and +5 mm d<sup>-1</sup> are  
478 simulated across the Bay of Bengal and Arabian Sea associated with BC SDE (Figure 11d).

479 From July through August, precipitation increases of 1 mm d<sup>-1</sup> are simulated across India,  
480 driven primarily by dust ARI. Meanwhile, dust ARI-driven precipitation increases of 2 mm d<sup>-1</sup>  
481 are simulated through September across the IGP (see Figure 12). Additionally, BCD SDE  
482 contributes to slight increases of 0.5 mm d<sup>-1</sup> or less across the IGP from May through August.  
483 Across India however, with the exception of July, BCD-induced SDE contributes to decreased  
484 precipitation from March through October, which a peak reduction in June at 1.2 mm d<sup>-1</sup>.

485 The TPF region is characterized by the largest precipitation increases relative to the other  
486 subregions due to an enhancement of BCD effects by the complex terrain, with BCD effects  
487 bringing forth increases in precipitation from April through August (see Figure 12c). BCD  
488 contribute to unanimous SDE- and ARI-induced precipitation increases in June across the TPF,  
489 with dust and BC ARI (both +4.5 mm d<sup>-1</sup>) and BC SDE (+4.3 mm d<sup>-1</sup>) dominating the enhanced



490 precipitation; the collective impacts of BCD enhance precipitation by more than  $7 \text{ mm d}^{-1}$  in June.

491 Dust SDE contributes to smaller MJ anomalies of less than  $+1 \text{ mm d}^{-1}$  across the TPF.

492 BCD ARI-induced precipitation differences drive changes in runoff across India and the  
493 IGP, with runoff increases induced from April through October. Maximum precipitation increases  
494 of  $2.5 \text{ mm d}^{-1}$  and  $2 \text{ mm d}^{-1}$  occur a month ahead of runoff increases of  $2 \text{ mm d}^{-1}$  and  $1.6 \text{ mm d}^{-1}$   
495 across India and the IGP, respectively (Fig. S4). The maximum precipitation (runoff) increase  
496 occurs in June (July) across India, while the maximum precipitation (runoff) increase occurs in  
497 July (August) across the IGP. In contrast to India and the IGP, the maximum precipitation  
498 increases in TPF driven by BCD ARI occur in the same month (June) as runoff, with precipitation  
499 (runoff) increases of  $7 \text{ mm d}^{-1}$  ( $6 \text{ mm d}^{-1}$ ). The reasons for the differences in runoff/precipitation  
500 phase between India/IGP and the TPF may be the result of larger runoff effects associated with  
501 the SDE across the TPF compared to the IGP and India, which contributes to runoff increases of  
502 more than  $3.5 \text{ mm d}^{-1}$  in June (Fig. S4c,e).

503

## 504 **5. Nature of the simulated changes**

505 The simulated changes to south-Asian climate introduced by BCD are the result of direct  
506 aerosol interactions with sunlight and outgoing terrestrial radiation, which leads to circulation  
507 changes brought about by stability and thermodynamic modifications of the atmospheric column.  
508 Furthermore because this study did not attempt to isolate the near-field and far-field aerosol  
509 effects on the SAM, we restrict our attention to the combined near- and far-field BCD effects  
510 represented in overall circulation and meteorology perturbations across the region.

511

### 512 **5.1 Dynamical impacts of BCD on the SAM**

513 The changes induced by the combination of BCD effects on the premonsoonal and  
514 monsoonal meteorology can be examined by considering thermal vorticity,  $\zeta_T$  (Bluestein 1992).  
515 Analogous to thermal wind,  $\zeta_T$  is defined to be the difference between the upper-level





516 geostrophic vorticity ( $\zeta_{g,above}$ ) and the lower-level geostrophic vorticity ( $\zeta_{g,below}$ ) within a  
517 column:

$$\zeta_T \equiv \zeta_{g,above} - \zeta_{g,below}, \quad (8)$$

518 and,

$$\zeta_g = \frac{1}{f_0} \nabla_p^2 \Phi, \quad (9)$$

519 where  $f_0$  is the constant Coriolis parameter and  $\Phi \equiv gz$  is the local geopotential height. Taking  
520 the derivative of Eq. (9) with respect to independent variable  $t$ , we find the local tendency in  $\zeta_g$  to  
521 be:

$$\frac{\partial \zeta_g}{\partial t} = \frac{1}{f_0} \nabla_p^2 \frac{\partial \Phi}{\partial t}. \quad (10)$$

522 Substituting Eq. (10) into Eq. (8), we get:

$$\frac{\partial \zeta_T}{\partial t} = \frac{1}{f_0} \nabla_p^2 \left( \left. \frac{\partial \Phi}{\partial t} \right|_{above} - \left. \frac{\partial \Phi}{\partial t} \right|_{below} \right) = \frac{1}{f_0} \nabla_p^2 \frac{\partial \Delta \Phi}{\partial t}, \quad (11)$$

523 where  $\Delta \Phi \equiv \Phi_{above} - \Phi_{below}$  is the average column thickness. Here, variable  $t$  can be time, but  
524 because we are applying Eq. (11) to independently run experiments,  $t$  is more accurately an  
525 independent variable denoting case. For simplicity, we define all variables that are subject to the  
526 operator  $\partial/\partial t$  to be a tendency. Eq. (11) states that the thermal vorticity tendency is proportional  
527 to the Laplacian of the layer thickness tendency, which is proportional to the mean layer  
528 temperature. This section will thus utilize the 300-to-700 hPa column averaged temperature to  
529 evaluate circulation changes.

530 For synoptic-scale flows, it can be shown via scale analysis that the local vorticity  
531 tendency is dominated by the stretching of earth's vorticity, or:

$$\frac{\partial \zeta_g}{\partial t} = -\delta f_0, \quad (12)$$

532 where  $\delta \equiv \frac{\partial u}{\partial x} + \frac{\partial v}{\partial y}$ . Eq. (11) can be written as:



$$\frac{1}{f_0} \nabla_p^2 \frac{\partial \Delta \Phi}{\partial t} = -f_0 (\delta_{above} - \delta_{below}) = -f_0 \Delta \delta. \quad (13)$$

533 In the above series of equations,  $u$  and  $v$  are the zonal ( $x$ ) and meridional ( $y$ ) components of the  
 534 3-dimensional wind field. The upper- and lower-tropospheric divergence of the horizontal wind  
 535 field is given by  $\delta_{above}$  and  $\delta_{below}$ , respectively, and  $\Delta \delta \equiv \delta_{above} - \delta_{below}$ .

536 Using Eq. (13), we can link BCD-induced temperature changes to circulation changes  
 537 during premonsoonal months. Figure 13a shows simulated MJ anomalies in 300-to-700 hPa  
 538 column averaged temperature. BCD-induced columnar warming of between 1°C and 3°C occurs  
 539 in a belt between 20°N and 40°N, within a zone of climatologically maximized west-southwesterly  
 540 upper-tropospheric flow (see Fig. S9c). These strong upper tropospheric winds are responsible for  
 541 transporting dust downstream of major emission sources such as northern Africa and the Middle  
 542 East. The warming in this belt is also due to changes in clouds through evaporation and regional  
 543 circulation changes (Figure 5a). Under this warming scenario,  $\frac{\partial \Delta \Phi}{\partial t} > 0$  (i.e., the layer expands). If  
 544 it is assumed that the geopotential field is the linear combination of sinusoidal functions, the  
 545 Laplacian of a positive quantity will contribute to a negative thermal vorticity tendency by Eq.  
 546 (11). Hence, a warming/expanding column will bring about  $\Delta \delta > 0$  via Eq. (13), and upper-  
 547 tropospheric anticyclonic  $\zeta_g$  by Eq. (12) (and  $\Phi$  rises; Figure 13a,d) will tend to be generated  
 548 atop lower-tropospheric cyclonic  $\zeta_g$  by Eq. (12) (and  $\Phi$  falls). Via continuity, these changes in  
 549 the vorticity field lead to changes in vertical motion under the assumption of incompressibility.  
 550 As aerosols heat the atmosphere,  $\frac{\partial \Delta \Phi}{\partial t} > 0$ ,  $\frac{\partial \zeta_T}{\partial t} < 0$ , and  $\Delta \delta > 0$ . In an atmosphere where  
 551 divergence increases with increasing height ( $\frac{\partial \delta}{\partial z} > 0$ ), the continuity equation subject to  
 552 incompressibility is:

$$\frac{\partial \delta}{\partial z} = -\nabla_{x,y}^2 w, \quad (14)$$



553 where  $\nabla_{x,y}^2$  is the one-dimensional Laplacian and  $w$  is the vertical component to the 3-D velocity  
554 vector. Under such conditions,  $w$  must be positive if it is assumed that  $w(x,y,z,t)$  is a linear  
555 combination of sinusoidal functions. BCD warming brings forth an enhancement of the rising  
556 branch of the thermally direct monsoon circulation.

557 The application of Eqs. (13) and (14) to our results couples the total (BCD ARI and SDE)  
558 thermodynamic changes to the circulation changes during MJ. Maximal column averaged  
559 temperature anomalies (Figure 13a) are collocated with an upper-tropospheric anticyclonic  
560 anomaly in the mass field (Figure 13d) across southwest into southern Asia during MJ, with 250  
561 hPa pressure surface height rises of between 6 and 9 dm across south-central Asia (Figure 13a).

562 The response in the mass field due to BCD ARI and SDE is not confined only to the  
563 upper troposphere, where the geostrophic approximation is most applicable. In fact, BCD-induced  
564 cyclonic changes in the 850 hPa flow are simulated (contour fill, Figure 13b), with an  
565 intensification in the WLLJ of as much as  $5 \text{ m s}^{-1}$ , extending from eastern Africa, bifurcating the  
566 Arabian Sea, and protruding into southeast Asia during MJ. The magnitude of WLLJ  
567 intensification is smaller across India ( $2 \text{ m s}^{-1}$  to  $4 \text{ m s}^{-1}$ ), but its magnitude across the Bay of  
568 Bengal is more comparable to the intensified westerlies in the Arabian Sea. To this feature's north,  
569 there are simulated BCD-induced easterlies of  $-1 \text{ m s}^{-1}$  across eastern Iran into Russia, but the  
570 most pronounced low-level flow changes lie across the open sea. This is to be expected, as the  
571 lower tropospheric geostrophic assumption is more erroneous over land compared to oceans due  
572 to the far greater surface friction over land. More generally, friction may explain why the 850 hPa  
573 vorticity feature is more diffuse than the 250 hPa vorticity feature during MJ across south-central  
574 Asia.

575 The BCD-induced anomalies in the mass fields, discussed by invoking thermal vorticity  
576 arguments, lead to changes in the vertical motion pattern, water vapor budget, and precipitation  
577 patterns across south-Asia during MJ. 850 hPa  $q$  increases across south-Asia of  $1 \text{ g kg}^{-1}$  to  $2 \text{ g kg}^{-1}$   
578 are simulated (Figure 13c) due to a stronger WLLJ (Figure 13b). Meanwhile, BCD warming



579 brings enhanced rising vertical motion by Eq. (14) across south Asia. These increases in  $w$  and  $q$   
580 correlate with precipitation increases of  $+2 \text{ mm d}^{-1}$  across India during MJ, while precipitation  
581 increases of greater than  $6 \text{ mm d}^{-1}$  are simulated across the western Ghats and TPF (see Figure  
582 13c). The latter increases may be due to increased orographic effects, as low-level upslope flow  
583 over the Ghats is enhanced by a strengthened WLLJ, and upslope flow over the TPF is enhanced  
584 by stronger cyclonic flow across India (Figure 13b). Positive precipitation anomalies increase in  
585 magnitude towards the east across the Arabian Sea and Bay of Bengal as eastward-moving  
586 precipitation systems gain sensible and latent heat from the open waters.

587 Into JA, the belt of maximum 300-to-700 hPa column heating shifts north (Fig. S11a)  
588 along with the subtropical jet. Meanwhile, the area of WLLJ intensification shifts north and  
589 shrinks significantly compared to MJ (Fig. S11b). A precipitation dipole is simulated across the  
590 western mountains of India during JA, with increases (decreases) in excess of  $4 \text{ mm d}^{-1}$  located  
591 further north (south) in the mountain chain. Precipitation increases greater than  $4 \text{ mm d}^{-1}$  are also  
592 simulated across the western IGP and the southern half of Pakistan (Fig. S11c), with the  
593 maximum BCD precipitation increases tending to occur on the eastern nose of the WLLJ anomaly  
594 (as in MJ).

595 It is noted in Section 4.5, when compared to other subregions, TPF precipitation is more  
596 sensitive to BC SDE, in addition to BC ARI and dust ARI (see Figure 12). This could indicate the  
597 presence of the elevated heat pump effect, which develops in proximity to the TP as anomalous  
598 BCD-induced heating of the TP leads to enhanced anabatic upslope flow. This leads to stronger  
599 precipitation across the TPF compared to other subregions, in addition to the enhanced southerly  
600 cyclonic flow incident on the TPF during MJ.

601

## 602 **5.2 Dominant species and effects contributing to SAM alterations**

603 Different species (BC or dust) and different radiative effects (SDE or ARI) regulate the  
604 BCD impact on the SAM and premonsoonal meteorology. Figure 14 depicts the vertical structure



605 of BCD effect-wise changes in temperature,  $w$ , CF, and  $q$  across northern India during MJ,  
606 averaged horizontally between 25°N to 30°N and 75°E to 80°E (north-central India). The  
607 combined effects of BCD SDE and ARI contribute to low-level cooling (1.2°C) beneath middle-  
608 and upper-tropospheric warming (as large as 2.1°C), which increases thermodynamic stability.  
609 Above 3 km, dust ARI drives atmospheric warming as large as 1.5°C. BC ARI contributes to  
610 atmospheric warming as large as 1.1°C in a pattern similar to that of dust ARI. The BC SDE  
611 contributes to middle-tropospheric temperature changes that are similar in sign to dust ARI but  
612 half the magnitude, while dust SDE contributes the least to atmospheric heating compared to  
613 other effects (Fig. 14a) during MJ. Finally, the lower-tropospheric cooling is the result of cloud  
614 increases through  $w$  and  $q$  increases.

615 Accompanying the BCD-induced warming of the 3-to-15 km layer is an increase in  
616 tropospheric  $w$  up to 17 km during MJ. As with temperature, BCD ARI drives  $w$  increases that  
617 peak around 12 km at 0.5 cm s<sup>-1</sup>. Dust ARI accounts for a large portion of the total  $w$  increase,  
618 with BC ARI-induced  $w$  increases ~30% less. Meanwhile, BC SDE contributes to slightly  
619 stronger upward vertical velocities than BC ARI below 11 km, while the reverse is true above this  
620 altitude. This result may indicate the presence of the elevated heat pump effect (Lau et al., 2010)  
621 associated with strong warming on the southern TP periphery. BC SDE-induced warming of the  
622 TP periphery may induce locally strong solenoidal circulations due to horizontal density  
623 variations between the warming TP and the adjacent free atmosphere. These circulations could  
624 manifest as lower-tropospheric anabatic branches of rising vertical motion across northern India,  
625 which may be stronger than the larger-scale thermally-direct rising motions induced by BC ARI.  
626 As a result, BC ARI would induce positive  $w$  anomalies throughout the tropospheric column  
627 across northern India, while BC SDE may generate rising vertical motion only in a layer of the  
628 atmosphere in which BC SDE-induced quasihorizontal temperature gradients could develop in  
629 proximity to the TP. As for temperature, dust SDE contributes to relatively small upward vertical  
630 velocity enhancements across northern India.



631            Interestingly, BCD ARI by itself corresponds with larger upward vertical velocity  
632 changes than those associated with the combined effects (SDE and ARI) of BCD, even though the  
633 combined effects lead to the strongest middle-tropospheric warming during MJ. This is because  
634 the combined effects of BCD greatly increase the thermodynamic stability of the atmosphere  
635 across the northern India such that  $w$  increases are depressed. Hydrostatically, warming of the  
636 mean atmospheric column tends to initiate a thermally direct rising bubble. However, if the  
637 heating is vertically non-uniform within the column, as is the case depicted in Figure 14a,  
638 changes in atmospheric thermodynamic stability may actually reduce the buoyancy of the rising  
639 bubble. This increased atmospheric stability can be tied to negative CF anomalies as well,  
640 because as the atmosphere becomes more stratified, increases in turbulent mixing and  
641 entrainment may incite the evaporation of clouds. Across northern Tibet, dust ARI and BC ARI  
642 actually stratify the atmosphere so much that CF anomalies approaching -10% are simulated  
643 during MJ (see Figure 10c). This leads to enhanced surface warming that persists into the  
644 monsoonal period.

645            The BCD-induced effects on MJ upward vertical velocities, driven primarily by dust ARI  
646 and to a lesser extent BC ARI and BC SDE, correlate positively with CF anomalies. This is  
647 especially true for upper-tropospheric clouds (Fig. 14c). CF increases of 2% are effectuated  
648 mainly by dust ARI below 4 km, while BC SDE primarily drives CF increases in excess of +6%  
649 around 17 km. The increase in upper-tropospheric clouds is the result of increased convective  
650 precipitation, while the low-level CF increases result from increases in boundary layer moisture  
651 of nearly  $3 \text{ g kg}^{-1}$  (Figure 14d) coupled with vertical velocity increases of  $0.1 \text{ cm s}^{-1}$  below 5 km  
652 and increased low-level thermodynamic stability.

653

### 654 **5.3 CONT-vr versus CONT-un effects**

655            VR experiments are comparable with UN experiments in simulating BCD effects (Figure  
656 15). These experiments simulate warming (cooling) across the TP (India) during MJ. However,



657 VR experiments reveal a larger area of warming  $>1.3^{\circ}\text{C}$  across the central and eastern TP (Figure  
658 15a) due to BCD effects compared to UN experiments (Figure 15d). Additionally, UN simulated  
659 BCD effects bring about stronger cooling across India and Pakistan (Figure 15d). These  
660 differences between VR and UN experiments result from a stronger BCD SDE in the UN  
661 experiments. Aerosol-induced snow melting is much stronger in the UN experiments (not shown),  
662 leading to a much stronger warming of the WTP than in the VR experiments. This brings about a  
663 stronger columnar warming in the UN experiments that leads to stronger Arabian Gulf moisture  
664 flow into India, and higher CF increases in the UN versus the VR experiments during MJ. The  
665 BCD ARI warming across the TP in the VR experiments is lower compared to the UN  
666 experiments because the VR experiments can capture mesoscale BCD heating patterns.

667

## 668 **6. Summary and Conclusions**

669 Implementing a variable resolution (VR) version of CESM allowed for a relatively high-  
670 resolution evaluation of the impacts of BC and dust on the south-Asian monsoon. With a  
671 horizontal grid spacing of  $0.125^{\circ}$  across the TP and the rest of central Asia, VR simulations were  
672 able to capture the horizontally heterogeneous warming induced by BCD on TP snowpack.  
673 Results indicated that BCD effects, driven mainly by BCD ARI, lead to an enhancement of the  
674 SAM through a radiative-dynamical feedback that enhances precipitation in MJ. Precipitation  
675 increases of greater than  $2\text{ mm d}^{-1}$  across central and northern India, with larger precipitation  
676 increases of more than  $6\text{ mm d}^{-1}$  are simulated over the Ghat Mountains and TPF due to  
677 orographic enhancement in MJ. Into JA, precipitation increases shift west and north, with  
678 precipitation increases of more than  $4.5\text{ mm d}^{-1}$  being simulated across the western IGP and  
679 southern Pakistan. Runoff increases follow precipitation increases across India and the IGP.  
680 Across the WTP and ETP, however, runoff changes are modulated primarily by BC SDE-induced  
681 snowmelt, with runoff increases (decreases) prior to (after) June; a majority of the simulated  
682 SWE reductions are due to BC SDE. Across the TPF, precipitation enhancements due to BC SDE



683 are comparable in magnitude to BCD ARI, indicating that BCD effects may be enhanced over  
684 regions where the TP can act as an elevated heat source.

685 The precipitation increases across south Asia during MJ, and across western India and  
686 southern Pakistan during JA, occur as BCD warm the atmospheric column by 1°C to 3°C over a  
687 large belt coincident with the subtropical jet extending from northeast Africa through to the TP  
688 region. This results in a large upper-level (lower-level) divergence and anticyclonic (convergence  
689 and cyclonic) feature in the wind fields. On the southern side of the low-level cyclonic feature, an  
690 intensified WLLJ extends from the horn of Africa through the eastern Bay of Bengal, bringing  
691 moisture to south Asia during the premonsoonal months. During JA, the WLLJ extends into the  
692 northern Arabian Sea. BCD-induced changes in the vertical gradient of the horizontal divergence,  
693 coupled with increased moisture across the region, brings forth stronger onset to monsoonal  
694 precipitation and increases overall monsoonal precipitation yields through increases in  $w$ . BCD  
695 effects are amplified across the western Ghat Mountains and the TPF, due to orographic and  
696 anabatic effects.

697 The results of this study agree well with Lau et al. (2017), suggesting that ARI contribute  
698 to the largest circulation changes during the premonsoonal and monsoonal periods. Furthermore,  
699 Vinoj et al. (2014) and Jin et al. (2016) concluded that monsoonal precipitation is positively  
700 correlated with dust transport from the Arabian Peninsula; absorbing dust initiates convergence  
701 across the Middle East, which drives moisture transport into south Asia on time scales of a week.  
702 Our results reinforce this conclusion. Dust ARI, as well as BC ARI, contribute to large increases  
703 in moisture, CF, precipitation, and upward vertical motion during the premonsoonal and  
704 monsoonal periods. Additionally, cloud reductions and precipitation increases due to semi-direct  
705 effects are simulated in our study with magnitudes comparable to Lau et al. (2010), although we  
706 simulate higher precipitation anomalies across the TPF and western India. Finally, in agreement  
707 with Das et al. (2015) we found that absorbing dust warms the atmospheric column downstream





708 of major emission sources, contributing to anomalous upper-level (lower-level) ridging  
709 (troughing) and an intensification of the WLLJ.

710           There are significant questions that remain regarding the impact of BCD on SAM and  
711 premonsoonal meteorology. First, both UN and VR experiments performed rather poorly in  
712 simulating the magnitudes of BC both in-snow and in-atmosphere. Specifically, atmospheric BC  
713 was underestimated compared to surface observations by simulations, while dust upstream of  
714 India was oversimulated compared to satellite measurements by simulations. That being said,  
715 there are improvements in VR experiments: CONT-vr reduced the average bias of the in-snow  
716 BC concentration by more than a factor of 2 compared to the CONT-un and better represented  
717 AOD seasonality in AERONET measurements compared to CONT-un. Second, there is a  
718 pronounced lack of in-situ dust measurements, both in-snow and in-atmosphere. This makes a  
719 model validation of dust aerosol quite difficult. Although our results indicate that BCD ARI are  
720 main drivers of precipitation changes across south Asia during the warm season, the scope of  
721 these results could change as more data become available for model evaluation. Third,  
722 simulations were conducted with prescribed sea-surface temperatures, so longer-term ocean-  
723 atmosphere feedbacks were not considered in the context of the aerosol effects. That being said,  
724 the heterogeneity of the SDE-induced meteorological anomalies across the TP brought forth by  
725 the use of a VR model improves significantly over its coarser resolution counterparts, making the  
726 approach of using a VR global model beneficial when examining the climate system across  
727 regions in which the topography is highly variable. Thus, this approach has significant utility in  
728 other areas in which complex terrain may be a critical regulator of regional climate.

729           An opportunity exists for these simulations to be conducted without prescribed SSTs, as  
730 ocean-atmosphere feedbacks may affect the interseasonal and interannual variability of the  
731 monsoon. These feedbacks may depress or enhance the various BCD effects discussed here.  
732 Additionally, it has been shown that monsoon intensity correlates with precipitation and wavetrain  
733 patterns far downstream of the Asian continent (Lau and Weng, 2002). Examining how this



734 telconnection's sensitivity varies with the loading of light-absorbing aerosols may shed light on  
735 the importance of pollution in affecting far-field climate.

736

#### 737 **Author Contribution**

738 Stefan Rahimi and Chenglai Wu set up and ran the simulations. Xiaohong Liu advised all  
739 analyses and provided financial support. William K. M. Lau and Yun Qian collaborated in  
740 conceptualization. Mingxuan Wu helped in the analyses of dust variables. Hunter Brown assisted  
741 in clarifying the overarching messages in this manuscript.

742

#### 743 **Acknowledgements**

744 We thank Dr. Chun Zhao for his personal communication. We also warmly acknowledge  
745 the scientists responsible for the development and processing of aerosol reference data. MODIS  
746 data can be found at: [https://ladsweb.nascom.nasa.gov/api/v1/productPage/product=MOD08\\_M3](https://ladsweb.nascom.nasa.gov/api/v1/productPage/product=MOD08_M3),  
747 MISR data can be found at <https://10dup05.larc.nasa.gov/MISR/cgi-bin/MISR/main.cgi>, and  
748 MERRA-2 data can be accessed at <https://gmao.gsfc.nasa.gov/reanalysis/MERRA-2/>. We also  
749 acknowledge the scientists responsible for the maintenance of the AERONET and MACv2  
750 datasets. We also acknowledge Colin Zarzycki and Paul Ulrich for their help in setting up the VR  
751 grid.

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965 **Tables**

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**Table 1. List of VR simulations and the BCD effects they include.**

<i>Experiment name</i>	<i>BC effects in snow</i>	<i>BC effects in atmosphere</i>	<i>Dust effects in snow</i>	<i>Dust effects in atmosphere</i>
<b>CONT-vr</b>	<b>Yes</b>	<b>Yes</b>	<b>Yes</b>	<b>Yes</b>
<b>noSDE-vr</b>	<b>No</b>	<b>Yes</b>	<b>No</b>	<b>Yes</b>
<b>noARI-vr</b>	<b>Yes</b>	<b>No</b>	<b>Yes</b>	<b>No</b>
<b>noBCSDE-vr</b>	<b>No</b>	<b>Yes</b>	<b>Yes</b>	<b>Yes</b>
<b>noBCARI-vr</b>	<b>Yes</b>	<b>No</b>	<b>Yes</b>	<b>Yes</b>
<b>noDSDE-vr</b>	<b>Yes</b>	<b>Yes</b>	<b>No</b>	<b>Yes</b>
<b>noDARI-vr</b>	<b>Yes</b>	<b>Yes</b>	<b>Yes</b>	<b>No</b>
<b>noBCD-vr</b>	<b>No</b>	<b>No</b>	<b>No</b>	<b>No</b>
<b>CONT-un</b>	<b>Yes</b>	<b>Yes</b>	<b>Yes</b>	<b>Yes</b>
<b>noSDE-un</b>	<b>No</b>	<b>Yes</b>	<b>No</b>	<b>Yes</b>
<b>noARI-un</b>	<b>Yes</b>	<b>No</b>	<b>Yes</b>	<b>No</b>
<b>noBCD-un</b>	<b>no</b>	<b>No</b>	<b>no</b>	<b>No</b>

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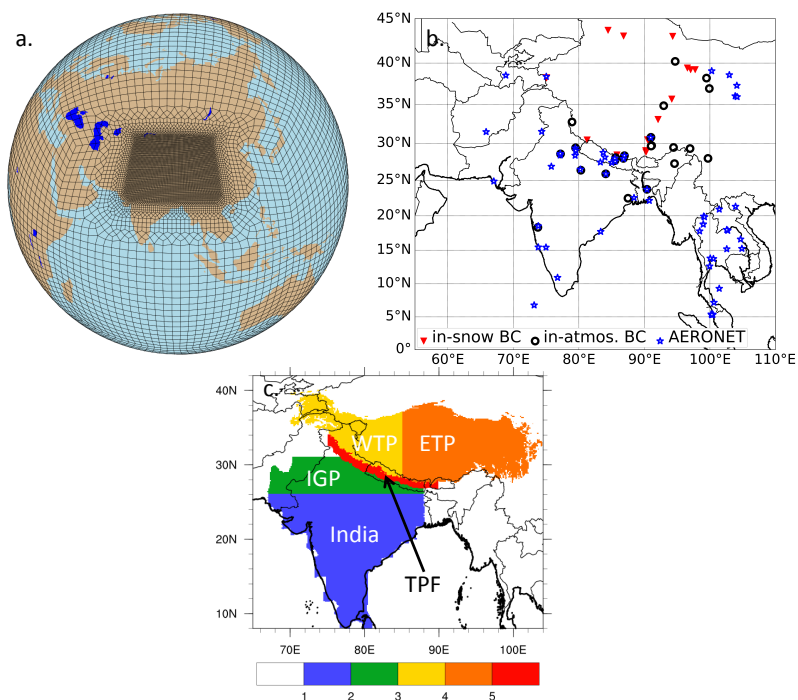
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### Figures



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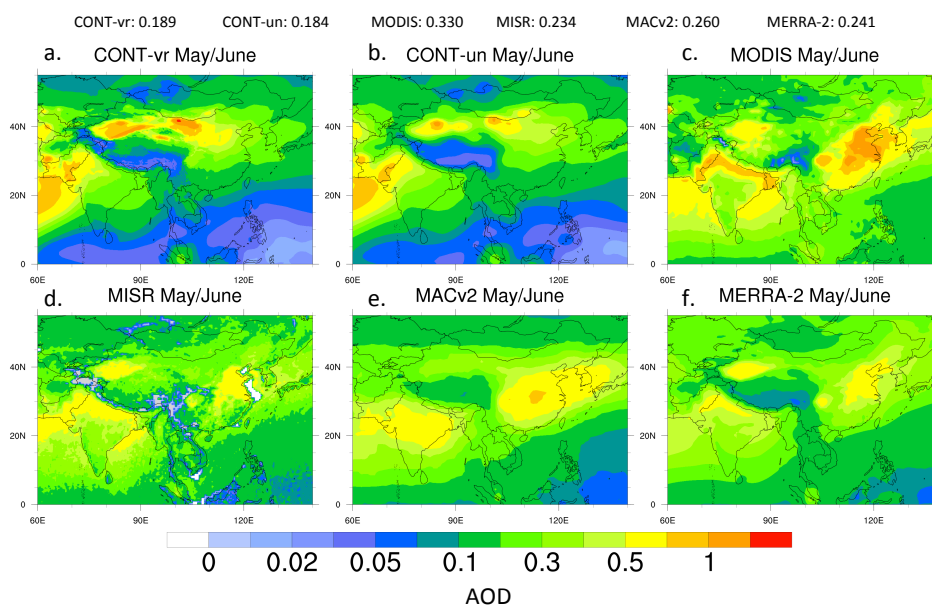
Figure 1. The variable-resolution (VR) grid points are shown in (a), and locations for point-source surface-based aerosol measurements are shown in (b). The analysis subregions are shown in panel (c).

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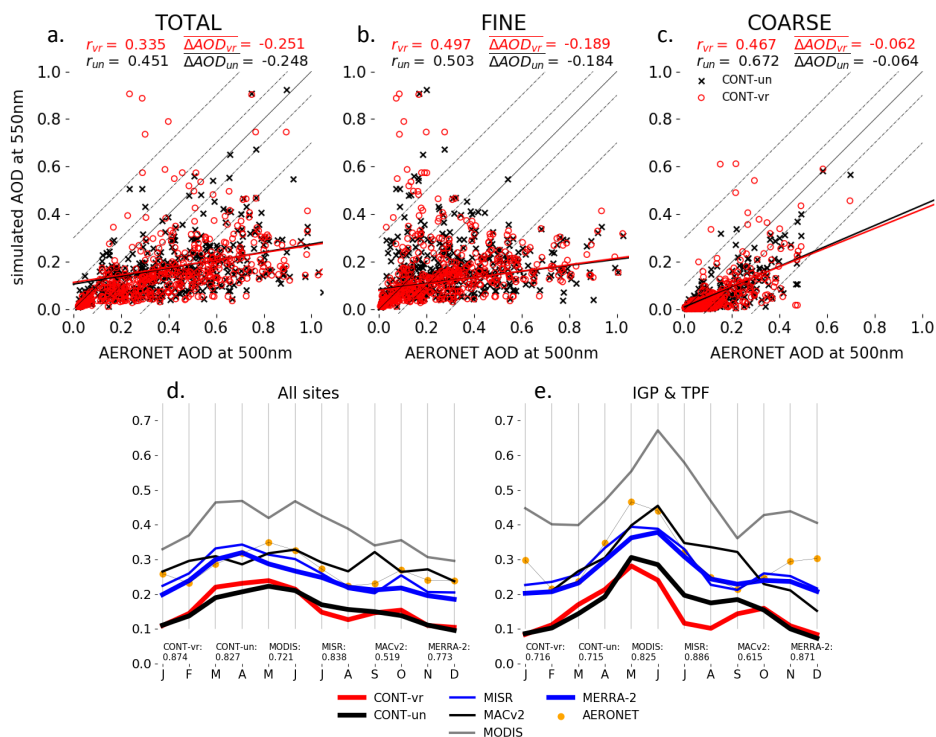
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Figure 2. Panels (a), (b), (c), (d), (e), and (f) depict May-June-averaged AOD values across south-Asia for CONT-vr, CONT-un, MODIS, MISR, MACv2, and MERRA-2, respectively. AOD averages from these respective data between 0°N-60°N and 60°E-140°E are given at the top.



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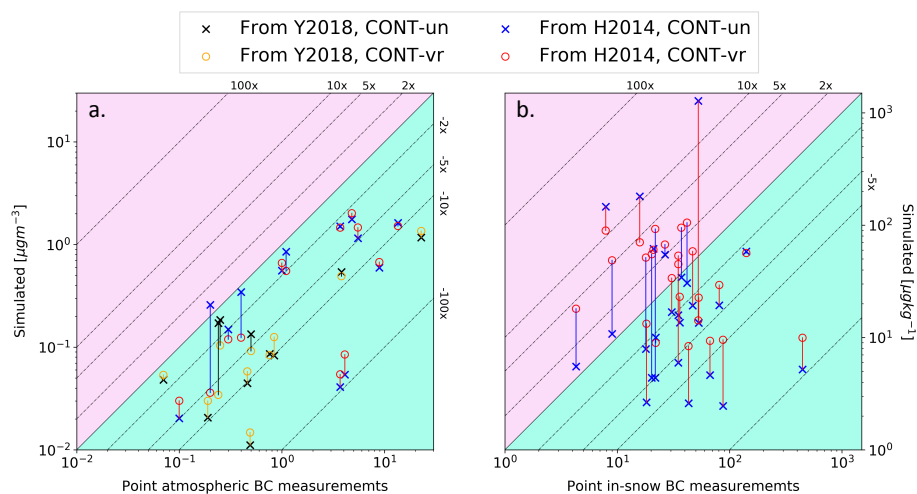
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Figure 3. Measurements from AERONET compared to CESM simulations for (a) total-, (b) fine-, and (c) coarse-mode annually averaged AOD. Pearson correlation ( $r$ ) values between simulations and observations are given, as are mean AOD differences ( $\overline{\Delta AOD}$ ). A best-fit line for the scatter data between CONT-vr (CONT-un) is plotted in red (black). The thin black line is the 1-to-1 curve. Panels (d) and (e) show the mean monthly variability of AOD averaged at all 57 AERONET sites and only for sites between 25°N-30°N and 70°E-90°E, respectively.  $r$ -values between various simulations/observations and AERONET are also given in panels (d) and (e).



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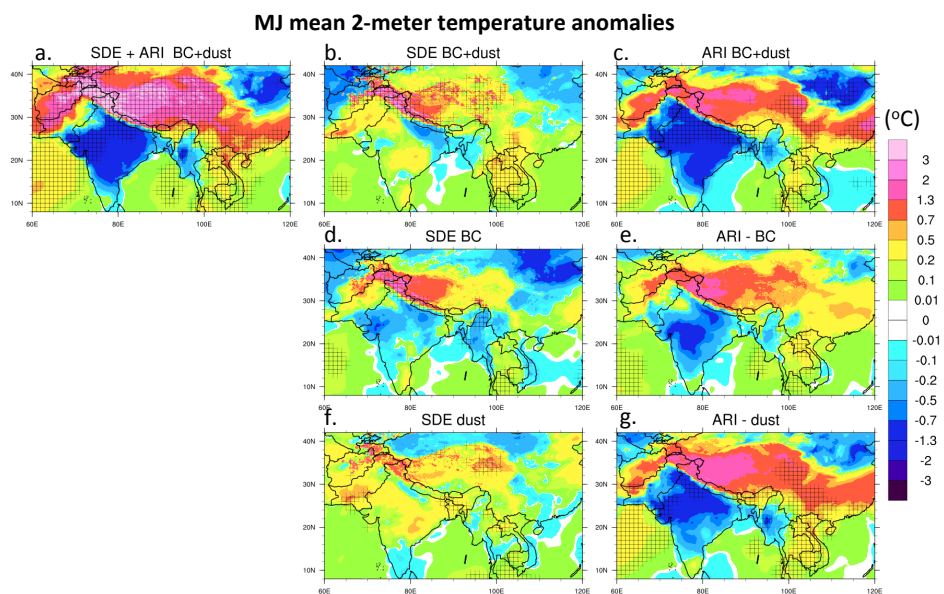
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**Figure 4.** Observed versus simulated (a) atmospheric and (b) in-snow BC. Observations are summarized in He et al. (2014; H2014) and Ye et al. (2018; Y2018). Pink- (sea foam-) colored areas denote areas of where simulations overpredict (underpredict) BC. The vertical lines connect identical observation points from CONT-vr and CONT-un, while the color of each line indicates which experiment is closer to observed BC measurements. The thin solid black diagonal represents the 1-to-1 curve, while the thin dashed diagonals represent factors of underprediction or overprediction by CESM experiments.

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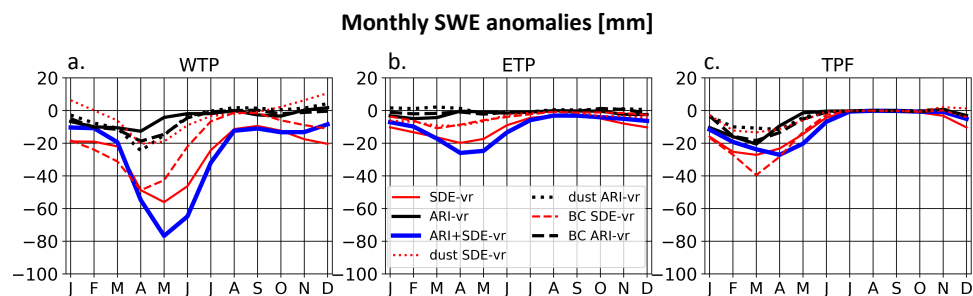
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Figure 5. May/June mean 2-meter temperature ( $T_2$ ) anomalies due to: (a) BCD-induced SDE+ARI, (b) BCD-induced SDE, (c) BCD-induced ARI (d) BC-induced SDE, (e) BC-induced ARI, (f) dust-induced SDE, and (g) dust-induced ARI. Hatching marks denote areas with  $t$ -values of 0.9 and greater, which have been interpolated to a  $1^\circ$  mesh for presentation. It is noted that, inside the  $1^\circ$  zone, there might be as many as 64 grid points that are characterized by statistically significant values.

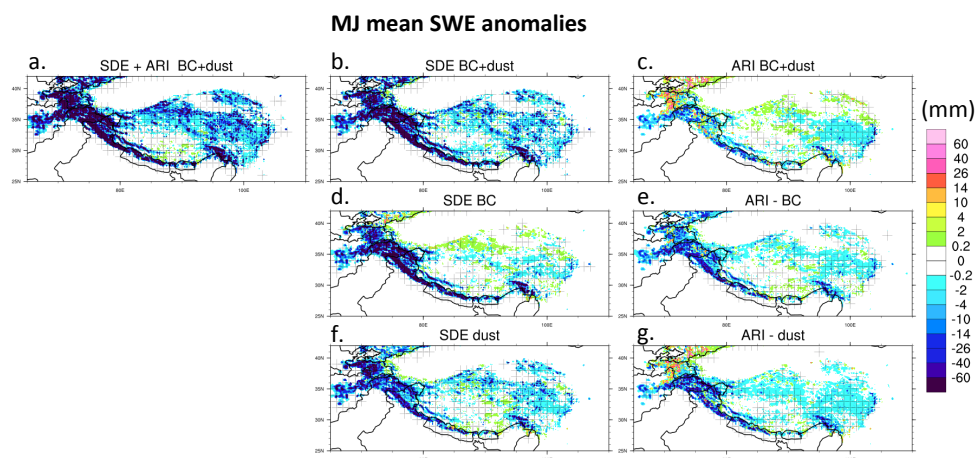
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Figure 6. Monthly time series of snow water equivalent (SWE) changes (millimeters) due to BCD effects across (a) the WTP, (b) ETP, and (c) TPF.

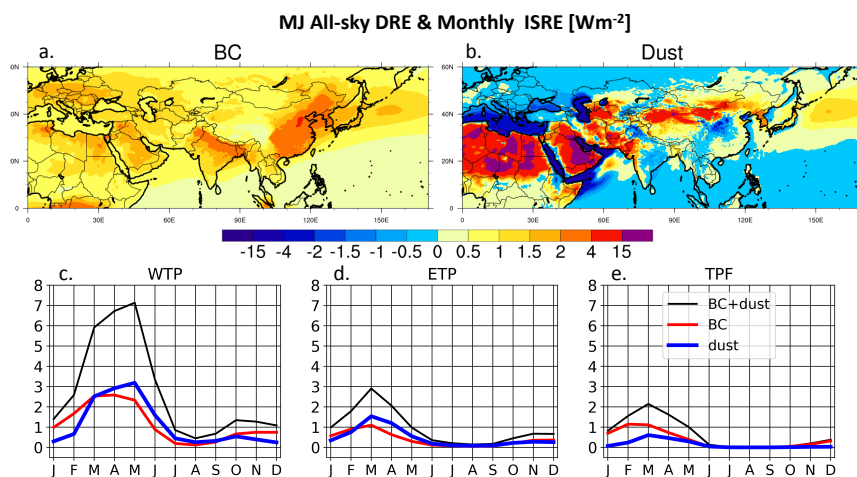
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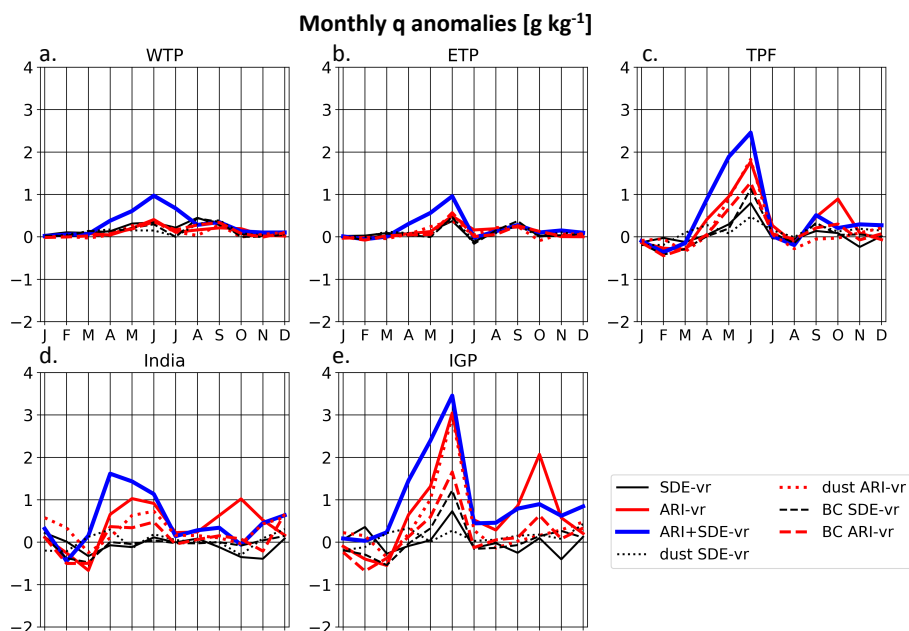
Figure 7. Same as in Figure 5, but for snow water equivalent (SWE).

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1166 **Figure 8. Diagonally-computed direct radiative effect (DRE) for (a) BC and, (b) dust**  
1167 **during MJ. Panels (c), (d), and (e) show the in-snow radiative effect (ISRE) for BCD across**  
1168 **the WTP, ETP, and TPF, respectively.**

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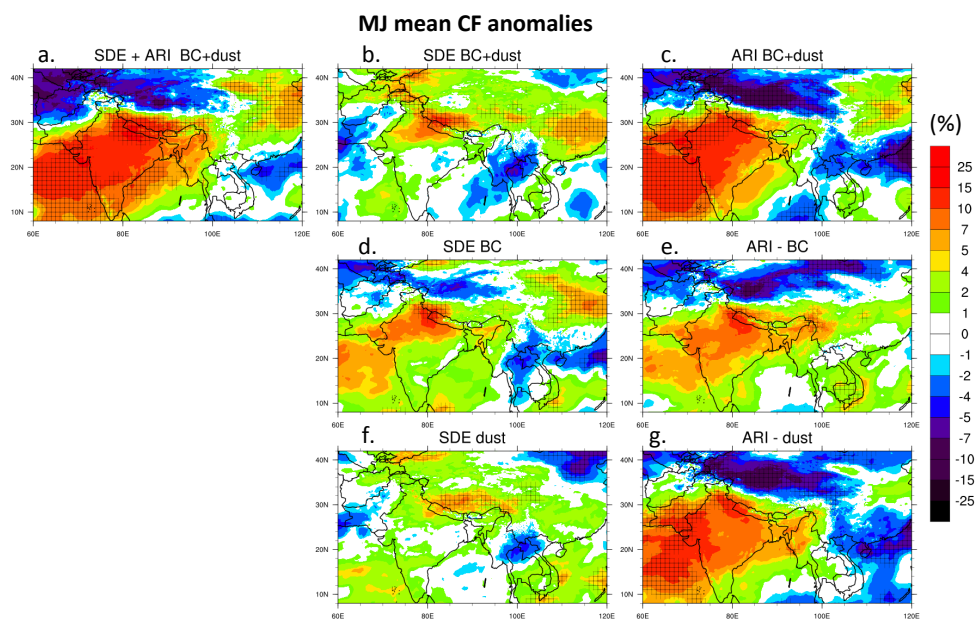


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Figure 9. Monthly time series of specific humidity ( $q$ ) changes in  $\text{g kg}^{-1}$  due to BCD effects across (a) WTP, (b) the ETP, (c) TPF, (d) India, and (e) IGP.

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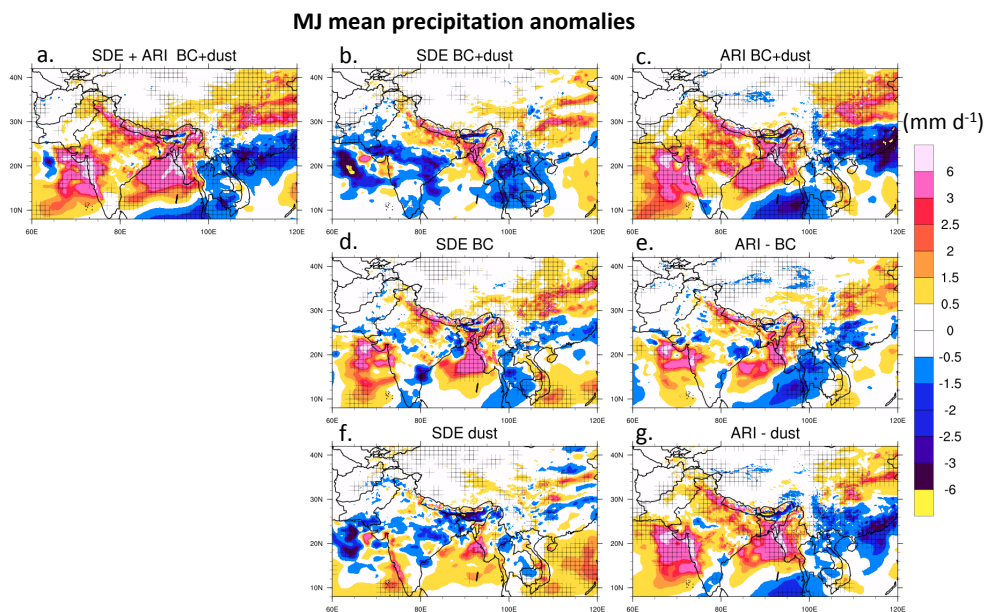




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Figure 10. Same as in Figure 5, but for cloud fraction (CF) (%).

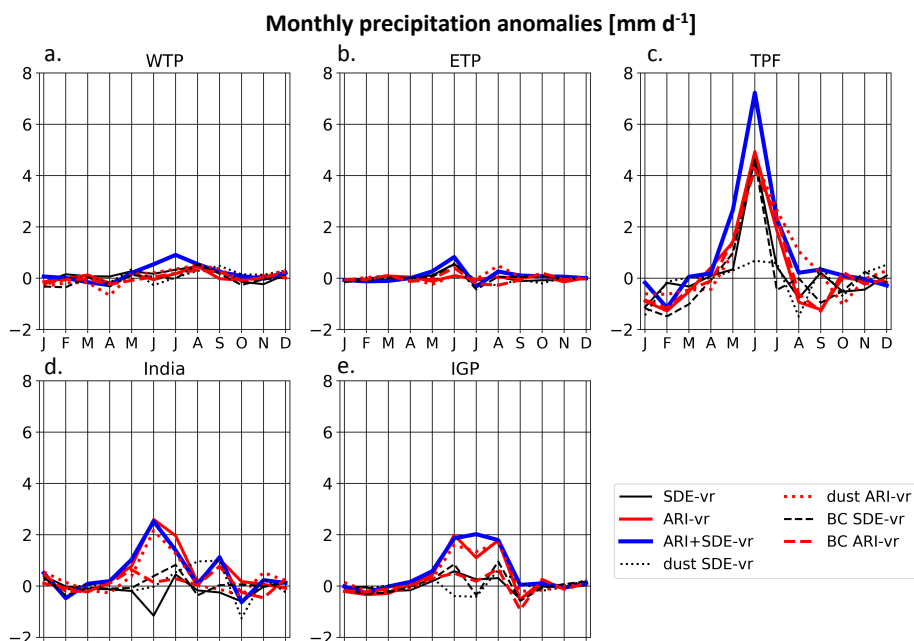
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Figure 11. Same as in Figure 5, but for precipitation rate (mm d<sup>-1</sup>).

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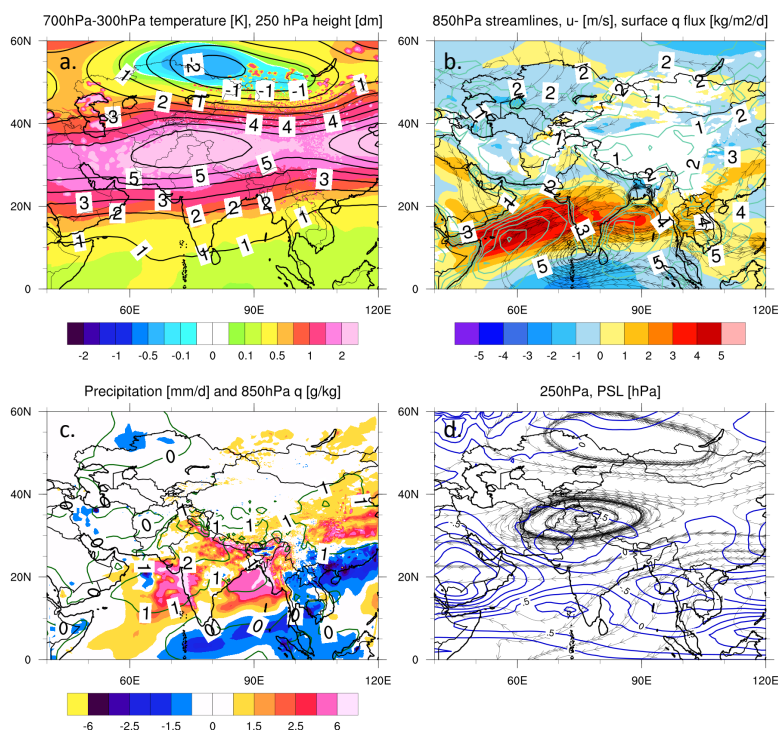
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Figure 12. Same as in Figure 9, but for precipitation rate ( $\text{mm d}^{-1}$ ). Note that the y-axis in (e) is different from panels (a)-(d).

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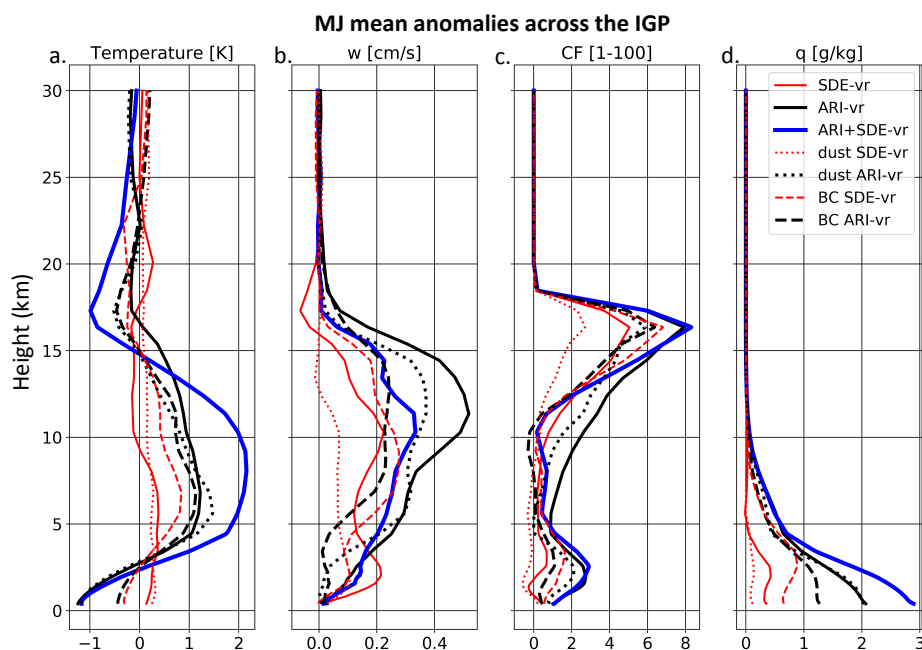
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Figure 13. MJ averaged BCD-induced anomalies in (a) 700-300 hPa column temperatures and 250 hPa heights, (b) 850 hPa  $u$ -anomalies (color fill), streamlines, and surface moisture flux (aquamarine contours), (c) precipitation rate (color fill) and 850 hPa specific humidity (dark green contours), and (d) 250 hPa streamlines and surface pressure.

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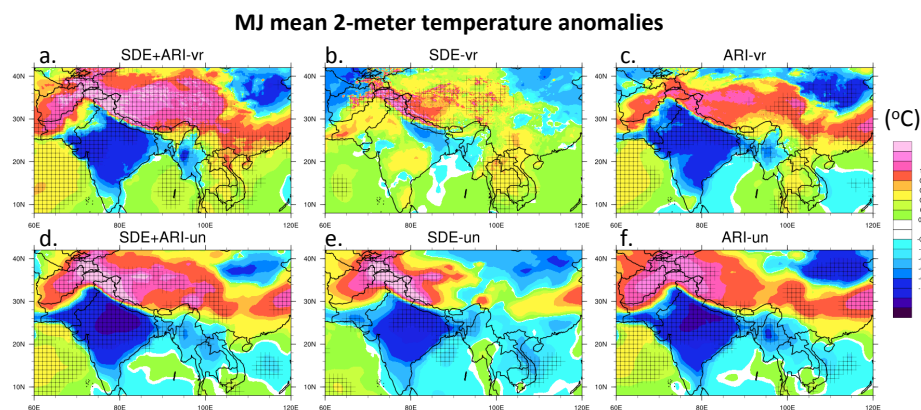
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Figure 14. Vertical profiles of BCD effect-wise anomalies in (a) temperature, (b) vertical velocity, (c) CF, and (d)  $q$  across the IGP during MJ, averaged horizontally between 25°N to 30°N and 75°E to 80°E. Red (black) curves depict ARI (SDE) effects, while the solid blue curve depicts the combined effects of BCD. Dashed lines represent BC-induced anomalies, while dotted lines represent dust-induced anomalies.

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**Figure 15.** Anomalous 2-meter temperatures during May-June due BCD (a) SDE+ARI, (b) SDE, and (c) ARI in the VR experiment. Panels (d)-(f) are the same as in (a)-(c) except for the UN experiments.

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