

1 **Regional and Teleconnected Impacts of Radiation-Topography Interaction over the**
2 **Tibetan Plateau**

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10
11 **Key Points:**

- 12 • Topography-radiation interaction over the Tibetan Plateau increases annual average near-
13 surface air temperature of the region by 0.26 K.
- 14 • Radiation-topography interaction over the Tibetan Plateau also affects the precipitation
15 patterns in South and East Asia.
- 16 • Including radiation-topography interaction overall improves the simulation of surface climate
17 over the Tibetan Plateau and Asian regions.
18

19 **Abstract**

20 Radiation-topography interaction plays an important role in surface energy balance over the
21 Tibetan Plateau (TP). However, the impacts of such interaction over the TP on climate locally
22 and in the Asian regions remain unclear. This study uses the Energy Exascale Earth System
23 Model (E3SM) to evaluate the regional and teleconnected impacts of radiation-topography
24 interaction over the TP. Land-atmosphere coupled experiments show that topography regulates
25 the surface energy balance, snow processes, and surface climate over the TP across seasons.
26 Accounting for radiation-topography interaction overall improves E3SM's performance in
27 simulating surface climate. The winter cold bias in air temperature decreases from -4.48 K to -
28 3.70 K, and the wet bias in summer precipitation is mitigated in southern TP. The TP's radiation-
29 topography interaction further reduces the South and East Asian summer precipitation biases.
30 Our results demonstrate the topographic roles in regional climate over the TP and highlight its
31 teleconnected climate impacts.

32

33 **Plain Language Summary**

34 The Tibetan Plateau (TP) is characterized by high elevation and complex topography. Interaction
35 between solar radiation and the undulating topography has important impacts on the regional
36 surface energy balance and hydrologic cycle. Here we use Earth System Model simulations to
37 show the local and remote impacts of the TP's radiation-topography interaction on the surface
38 climate of the Asian regions. Such interaction overall increases the air temperature especially in
39 winter over the TP and reduces the summer precipitation in southern TP. Teleconnectedly, the
40 interaction further alters the precipitation patterns in South and East Asia, by altering the
41 atmospheric circulation that influences moisture transport and clouds. Accounting for such
42 interaction generally improves the model performance when benchmarked against observations.
43 These findings underscore the important roles of the TP's radiation-topography interaction in
44 modulating the climate of the local and remote Asian regions.

45

46 **1 Introduction**

47 Radiation-topography interaction plays an important role in the surface energy balance [*Arnold*
48 *et al.*, 2006; *Lee et al.*, 2019]. Compared to flat terrain, rugged terrain alters direct solar
49 radiation, due to the local solar geometry, self-shadowing and cast shadowing from adjacent
50 terrain [*Dubayah and Rich*, 1995; *Olson and Rupper*, 2019]. Besides direct solar radiation, the
51 occlusion of adjacent terrain reduces diffuse radiation from sky [*Proy et al.*, 1989], while the
52 reflected radiation from adjacent terrain increases the solar radiation received by the surface, due
53 to the multi-scattering effects [*Sirguey*, 2009]. Neglecting the adjustment and redistribution of
54 solar radiation over mountainous regions due to its interaction with topography can cause large
55 uncertainties in modeling surface energy and water cycles [*Comola et al.*, 2015; *Liou et al.*,
56 2013], snow processes [*Hao et al.*, 2022], land-atmosphere interaction, and atmospheric
57 circulation and clouds [*Cai et al.*, 2023; *Gu et al.*, 2022; *Lee et al.*, 2019].

58

59 Parameterizations of radiation-topography interaction have been recently developed and
60 incorporated in a few Earth System Models (ESMs). However, nearly all the ESMs that

61 participated in the Coupled Model Intercomparison Project Phase-6 (CMIP6) adopt a simple
62 plane-parallel (PP) two-stream approximation scheme to describe the radiative transfer processes
63 with the assumption of flat surface. Based on the Monte Carlo photon tracing simulations, Lee et
64 al. (2011) developed a computationally-efficient radiative transfer parameterization (TOP) for
65 rugged terrain to consider the subgrid topographic effects on solar radiation. This TOP
66 parameterization has been recently implemented in the Community Earth System Model
67 (CESM) [Lee et al., 2019], Energy Exascale Earth System Model (E3SM) [Hao et al., 2021], and
68 Geophysical Fluid Dynamics Laboratory (GFDL) ESM [Zorzetto et al., 2023]. Such model
69 enhancement allows us to systematically explore the impacts of radiation-topography interaction
70 regionally and globally.

71

72 Radiation-topography interaction has been found to have large impacts on land surface and
73 atmospheric processes over mountainous regions such as the Tibetan Plateau (TP). Driven by
74 meteorological forcing, offline land simulations have shown non-negligible effects of
75 topography on shortwave radiation balance, surface turbulent heat flux, snow cover, and surface
76 temperature across a wide range of spatial resolutions from 1-km to 2° [Hao et al., 2021; Hao et
77 al., 2022; Zhang et al., 2022; Zorzetto et al., 2023]. However, offline land simulations neglect
78 the impacts of land-atmosphere interaction, motivating the use of land-atmosphere coupled ESM
79 experiments to investigate how radiation-topography interaction impacts both atmospheric and
80 land processes over the TP. Most ESMs tend to underestimate air temperature (T_{air}) and
81 overestimate precipitation (P) over the TP across seasons [Cui et al., 2021; Zhu and Yang, 2020].
82 The inclusion of radiation-topography interaction in land-atmosphere coupled CESM simulations
83 reduces the cold bias over the TP in winter [Lee et al., 2019], and overall decreases P across
84 seasons [Fan et al., 2019].

85 Besides local impacts, changes in the elevated heating due to radiation-topography interaction
86 over the TP may influence the climate in other regions through teleconnection by excitation of
87 Rossby waves. As the highest plateau in the Earth surface with large snow cover, TP plays an
88 important role in modulating the atmospheric circulation and shaping the weather and climate
89 around the TP [Wu et al., 2014; Wu et al., 2007; Yang et al., 2020]. For example, the spring
90 surface temperature over the TP shows a lag correlation with summer P in East Asia [Xue et al.,
91 2022]. Xue et al. [2022] identified an out-of-phase oscillation between the TP and Rocky
92 Mountain surface temperature and suggested that TP may provide a substantial source of
93 subseasonal-to-seasonal predictability for P in many global regions. Changing snow cover over
94 the TP can advance/delay the onset and strengthen/weaken the intensity of the East Asian
95 Summer Monsoon, and strongly influence the South Asian Summer Monsoon precipitation [Li et
96 al., 2018; You et al., 2020]. The projected surface darkening due to reduced snowpack by global
97 warming will strengthen the elevated heat pump of the TP, and further impact the remote Asian
98 monsoon systems [Tang et al., 2023]. Likewise, the radiation-topography interaction-induced
99 changes in land surface thermal conditions over the TP are expected to regulate the transport of
100 water and heat to the Asian downstream regions and further impact the climate of these
101 surrounding regions.

102

103 This study aims to investigate the regional and teleconnected impacts of radiation-topography
104 interaction over the TP. Specifically, we used E3SM to carry out present-day 40-year land-

105 atmosphere coupled experiments using three different model configurations to isolate the impact
106 of radiation-topography interaction over the TP. Using these experiments and six benchmark
107 datasets, we evaluated the local impacts of radiation-topography interaction on the TP's surface
108 climate across seasons, followed by analysis of the remote impacts of the TP on T_{air} and P
109 patterns of the East and South Asian regions.

110 **2 Materials and Methods**

111 2.1 Radiation-topography interaction in E3SM

112 E3SM, supported by the United States Department of Energy (DOE), is a state-of-the-art fully-
113 coupled ESM developed to address the grand challenge of robust, actionable predictions of the
114 variability and change of the Earth system [Leung *et al.*, 2020]. Rooted from CESM version-1,
115 the latest version-2 of E3SM (E3SMv2) features significant developments especially in the
116 atmospheric dynamical core and physics parameterization schemes, river routing, ocean and sea
117 ice components [Golaz *et al.*, 2022]. Compared to its predecessor E3SM version-1, E3SMv2
118 shows a higher computational efficiency and improved performance in simulated clouds and P
119 patterns [Golaz *et al.*, 2022]. The E3SM Land Model version-2 (ELMv2), which originated from
120 the Community Land Model version-4.5, includes a more realistic snow albedo parameterization
121 [Dang *et al.*, 2019] and an improved land biogeochemistry representation [Burrows *et al.*, 2020]
122 for various simulation campaigns. The new optional radiation-related configurations in ELMv2
123 include the TOP parameterization [Hao *et al.*, 2021], support for multiple types of snow grain
124 shape (i.e., spherical and non-spherical), and updates to the snow albedo parameterizations to
125 account for different mixing states of snow grain and light-absorbing particles (LAP) [Hao *et al.*,
126 2023].

127 ELMv2, by default, uses the PP parameterization to calculate surface shortwave radiation
128 balance without accounting for the impacts of topographic relief. The new TOP parameterization
129 in ELMv2 can capture the subgrid topographic effects on solar radiation [Hao *et al.*, 2021]. TOP
130 represents the relationship between the topography-related factors (i.e., the grid-average cosine
131 of local solar incident angle, sky view factor, terrain configuration factor, and standard deviation
132 of elevation) and the radiation adjustments caused by subgrid topography via multiple linear
133 regression [Hao *et al.*, 2021; Lee *et al.*, 2011]. The land-only ELM simulations showed that TOP
134 has better performance in simulating surface energy balance and water cycles in the TP than PP
135 [Hao *et al.*, 2021; Hao *et al.*, 2022]. The performance of TOP in land-atmosphere coupled
136 simulations in and around the TP is evaluated in this study.

137 2.2 Experimental Design

138 We conducted land-atmosphere coupled present-day simulations using E3SMv2 with three
139 different configurations: 1) the default PP scheme, denoted as PP_Globe; 2) the TOP scheme for
140 the TP region and PP for the rest of the globe (Figure S1), denoted as TOP_TP; and 3) the TOP
141 scheme for the global land, denoted as TOP_Globe. For each simulation, we used the F2010
142 component set with only active land, atmosphere and river components. In the F2010
143 configuration, the solar constant, sea surface temperature, sea ice, greenhouse gas concentrations,
144 and aerosol emissions are prescribed at the 2010 level. The E3SM Atmosphere Model version-2
145 (EAMv2) was set at approximately 1° spatial resolution with 72 vertical layers. ELMv2 was
146 configured at a 0.5° spatial resolution in the satellite phenology mode driven by the satellite-

147 derived climatological leaf area index data. We ran 40-year global simulations and used the last
 148 20-year simulations for model analysis. Both the EAMv2 and ELMv2 outputs were aggregated
 149 to seasonal and annual mean. The EAMv2 outputs were resampled to 0.5° for further analysis.

150 2.3 Model Analysis and Evaluation

151 To clarify the role of radiation-topography interaction, the three model simulations described in
 152 Section 2.2 were compared in the TP and downstream over East and South Asia, across the
 153 seasons: winter (DJF), spring (MAM), summer (JJA), autumn (SON) as well as annual average.
 154 Specifically, the difference between TOP_TP and PP_Globe was used to investigate the local
 155 and remote impacts of radiation-topography interaction over the TP. We also evaluated the
 156 difference between TOP_Globe and PP_Globe to diagnose the impacts of non-TP mountainous
 157 regions. Specifically, for the local impacts, we compared the spatiotemporal differences in land
 158 surface albedo (α), surface radiation fluxes, turbulent heat fluxes, snow cover fraction (f_{sno}),
 159 snow water equivalent (SWE), T_{air} , and P . For the remote effects, we investigate the impacts on
 160 T_{air} and P in two subregions: South Asia (SA; $10\text{-}25^\circ\text{N}$, $70\text{-}100^\circ\text{E}$) and East Asia (EA; $17\text{-}49^\circ\text{N}$,
 161 $105\text{-}140^\circ\text{E}$).

162 We collected six benchmark datasets from 2005-2015 for model evaluation (Table S1): (1) the
 163 surface radiation fluxes from the Clouds and the Earth's Radiant Energy System (CERES)
 164 Energy Balanced and Filled (EBAF) Edition 4.2 [*Nasa/Larc/Sd/Asdc*, 2023], (2) latent (F_{lat}) and
 165 sensible (F_{sen}) heat fluxes from FLUXCOM [*Jung et al.*, 2019], (3) the spatially- and temporally-
 166 complete (STC) snow-covered area and grain size (STC-MODSCAG) product [*Rittger et al.*,
 167 2020], (4) the snow property inversion from remote sensing (SPIReS) product [*Bair et al.*,
 168 2021]), (5) T_{air} from the University of Delaware (UDel) v5.01 terrestrial air temperature monthly
 169 data [*Willmott and Matsuura*, 1995], and (6) P from the Global Precipitation Climatology Project
 170 (GPCP) v2.3 [*Adler et al.*, 2018; *Huffman et al.*, 1997]. We used the average of STC-
 171 MODSCAG and SPIReS as the reference values of f_{sno} . All the datasets were resampled spatially
 172 to 0.5° and temporally to multi-year average seasonal mean to be identical with the model
 173 outputs. Based on these benchmark datasets, statistical metrics including correlation coefficient
 174 (R), area-weighted mean bias, and area-weighted root-mean-square-error (RMSE) were used to
 175 evaluate the model performance. We also calculated the relative difference (δ ; unit: %) between
 176 the mean bias of TOP_TP ($\text{Bias}_{\text{TOP_TP}}$) and that of PP_Globe ($\text{Bias}_{\text{PP_Globe}}$) as $(|\text{Bias}_{\text{TOP_TP}}| -$
 177 $|\text{Bias}_{\text{PP_Globe}}|)/|\text{Bias}_{\text{PP_Globe}}|*100$.

178 3 Results

179 3.1 Regional impacts on surface energy balance and surface climate

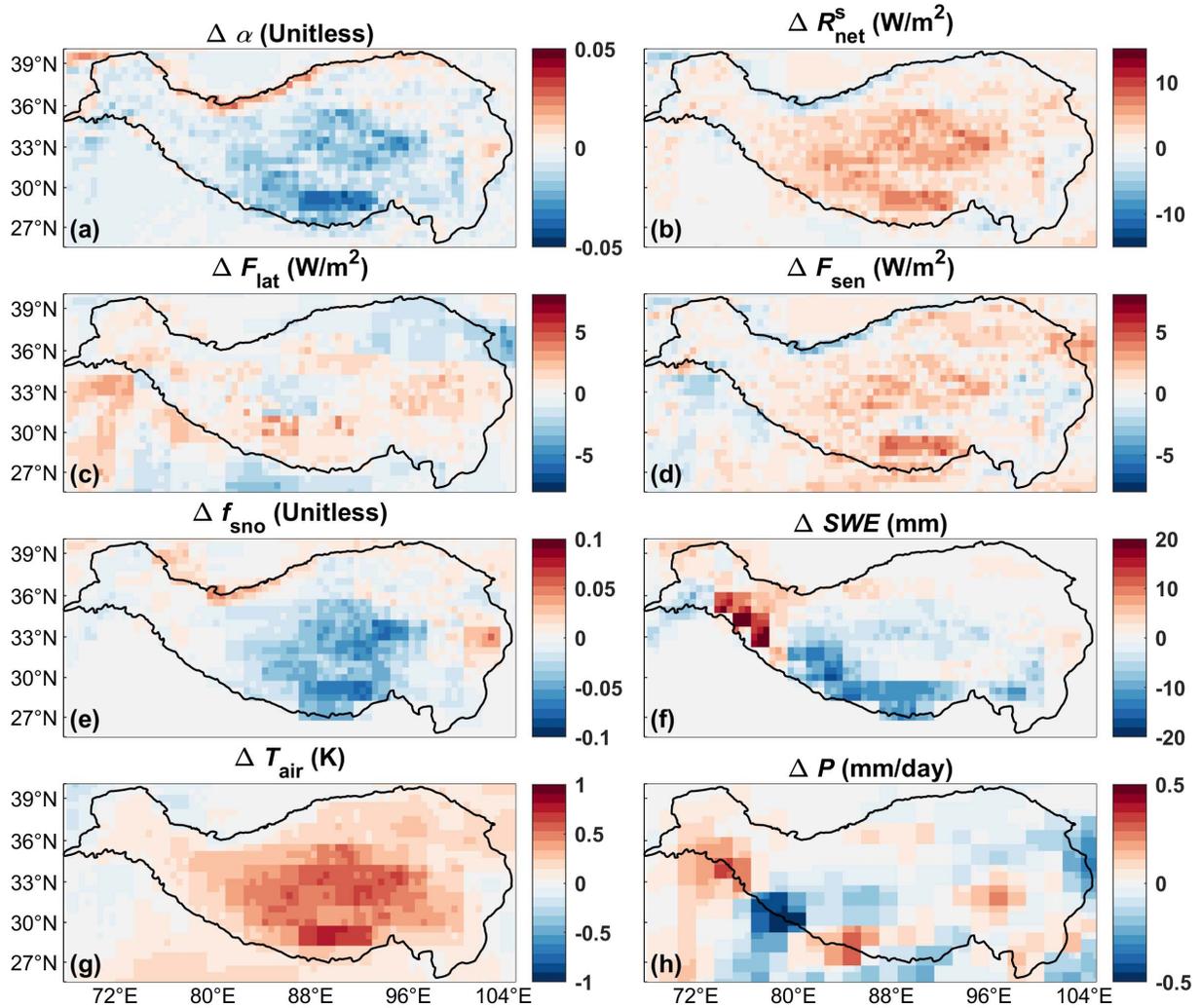
180 Radiation-topography interaction regulates the annual average surface energy balance over the
 181 TP. In TOP_TP, topography reduces α by 0.01 (mean value), especially in the central and
 182 southern TP by more than 0.05, while increasing α in the northern border of the TP (Figure 1a).
 183 TOP_TP shows lower cloud cover (Figure S2f) and thus larger downward solar radiation by 0.28
 184 W/m^2 (mean value) than PP_Globe (Figure S2a). Consequently, TP absorbs more solar radiation
 185 with the mean value of about $2.10 \text{ W}/\text{m}^2$ (Figure 1b), driven by the reduction of α and increase
 186 of downward solar radiation. Given that TP has an annual average downward solar radiation of
 187 $231 \text{ W}/\text{m}^2$ and annual average α of 0.34 in PP_Globe, the α reduction induced by topography
 188 accounts for about 96% of the increase in net solar radiation (R_{net}^S), while the downward solar

189 radiation changes only account for about 4% of the difference between TOP_PP and PP_Globe
 190 over the TP. The downward longwave radiation shows a small change of 0.19 W/m^2 in mean
 191 value (Figure S2b) responding to the cloud cover change, while the upward longwave radiation
 192 increases by 1.27 W/m^2 (mean value) (Figure S2c) associated with surface warming (Figure 1g).
 193 The change in radiation fluxes further increases both F_{lat} and F_{sen} (Figure 1c-d). F_{sen} shows a
 194 larger increase by 0.79 W/m^2 (mean value) than F_{lat} (0.18 W/m^2). For the seasonal variation,
 195 overall winter shows larger α reduction than summer (Figure S3a) due to higher snow cover and
 196 snow albedo. However, the topography-induced changes in R_{net}^s for all the four seasons are
 197 comparable (Figure S3b), because the seasonal variation of solar angle affects the available solar
 198 radiation. Although F_{sen} increases for all seasons (Figure S3d), F_{lat} shows smaller changes for all
 199 seasons and even decreases in autumn (Figure S3c). Besides the TP mean changes, the spatial
 200 patterns of the differences in radiative fluxes show larger seasonal variations (Figures S4-S7).

201 The increasing R_{net}^s overall decreases annual average f_{sno} over the TP (Figure 1e). The spatial
 202 pattern of the change in annual average f_{sno} is consistent with that of α (Figure 1a,e), attributed to
 203 the positive snow albedo feedback [Thackeray and Fletcher, 2016] where the darkened snow
 204 absorbs more solar radiation, accelerate snow aging and melt, and thus reduces f_{sno} , which further
 205 reduces α . Different from f_{sno} , SWE increases in the western TP and decreases in the central and
 206 southern TP (Figure 1f). Although topography can accelerate snow melt (Figure S2e), the
 207 increasing snowfall compensates the loss of snow masses over the western TP (Figure S2f). The
 208 snow-atmosphere interaction complicates the snow changes caused by topography compared to
 209 the offline land simulations. At the seasonal scale, similar to α , f_{sno} shows larger differences in
 210 cold seasons and the smallest changes in summer due to the smaller snow cover (Figure S3e).
 211 Similarly, SWE shows the largest decrease by 4.3 mm in winter (Figure S3f).

212 The increase in F_{sen} leads to higher annual average near-surface (i.e., 2 m) T_{air} over the whole TP
 213 (Figure 1g) by a mean value of 0.26 K. The increase in T_{air} for winter is more pronounced with a
 214 mean value of 0.78 K, while there is a slight decrease in the western and Southern TP for
 215 summer (Figures S3g and S6). The changing T_{air} and F_{lat} affect the water and heat exchange
 216 between land and atmosphere, and eventually affect the regional P . Overall the central and
 217 northern TP shows a decrease in annual average P , while the western TP shows an increase
 218 (Figure 1h). The seasonal differences in P are well correlated with the differences in clouds
 219 (Figure S2f). Topography reduces the summer P by 0.1 mm/day (mean value), especially in the
 220 central and southern TP (Figure 2h), related to the changing winds and cloud cover (Figures S2
 221 and S8). The topography-induced increase in winter T_{air} and decrease in summer P are expected
 222 to reduce the cold and wet bias of E3SM in the TP.

223



224

225 **Figure 1. Regional impacts of radiation-topography interaction on annual (a-d) surface**
 226 **energy balance, (e-f) snow, (g) air temperature (T_{air}) and (h) precipitation (P) over the TP.**
 227 For each panel, the impacts are represented by the difference (Δ) between TOP_TP and
 228 PP_Globe.

229

230 Including radiation-topography interaction overall improves the E3SM model performance in
 231 simulating surface energy balance, snow processes, and surface climate over the TP. For the
 232 annual scale, TOP_TP generally shows similar correlation, but smaller mean bias and RMSE
 233 with/than PP_Globe (Table 1). For example, δ of α is -7.4% which means that the mean bias of
 234 TOP_TP reduces 7.4% compared to that of PP_Globe, while the mean biases of R_{net}^s , f_{sno} and T_{air}
 235 reduces 23.2%, 6.7% and 11.1%, respectively. Seasonally (Table S2), for α , both TOP_TP and
 236 PP_Globe show high correlation (≥ 0.59) with CERES across seasons, but large positive mean
 237 bias and large RMSE, especially in winter. TOP_TP reduces about 5% and 6% of the mean
 238 biases compared to PP_Globe for winter and spring. For R_{net}^s , TOP_TP improves the correlation
 239 with CERES from 0.62 (PP_Globe) to 0.68 in winter and from 0.39 (PP_Globe) to 0.53 in
 240 autumn. TOP_TP shows slightly higher positive mean biases in summer, but lower negative
 241 mean biases and smaller RMSEs than PP_Globe in other seasons. For F_{lat} , both TOP_TP and

242 PP_Globe are similarly well correlated with FLUXCOM, but show large overestimations in non-
 243 winter seasons, although TOP_TP shows slightly lower negative mean biases in winter than
 244 PP_Globe. For F_{sen} , both TOP_TP and PP_Globe show low R values, high negative mean biases
 245 and large RMSEs especially in the warm seasons. Compared to PP_Globe, TOP_TP shows
 246 higher correlations in winter and autumn, and smaller negative mean biases across seasons than
 247 PP_Globe. F_{lat} and F_{sen} generally show opposite mean biases, implying that there are large
 248 uncertainties in partitioning the turbulent heat fluxes in E3SM. For f_{sno} , TOP_TP shows slightly
 249 better performance for all the three metrics. For T_{air} , TOP_TP shows similar R values with
 250 PP_Globe but reduces the cold bias especially in the cold seasons. For example, the negative
 251 mean bias of winter T_{air} decreases from -4.57 K to -3.79 K. For P , TOP_TP generally has similar
 252 mean biases and RMSEs, but higher R values than PP_Globe. For example, R increases from
 253 0.65 to 0.72 in spring, and TOP_TP also slightly reduces the summer wet bias.

254

255 Comparing TOP_Globe and PP_Globe produces results that are spatially (Figure S9) and
 256 temporally (Figure S10) similar to the comparison between TOP_TP and PP_Globe, despite
 257 some differences in the magnitude of the statistical metrics. The corresponding evaluation results
 258 are shown in Table S3.

259

260 **Table 1. Statistical metrics of E3SM simulated annual surface energy balance, snow**
 261 **variables, air temperature (T_{air}) and precipitation (P) against the benchmark datasets over**
 262 **the TP for both PP_Globe and TOP_TP.** The sources of the benchmark datasets are indicated
 263 in the second column. The corresponding metrics across seasons are listed in Table S2.

| Variable | Benchmark dataset | $R_{\text{PP_Glo}}$ be | $R_{\text{TOP_TP}}$ | $\text{Bias}_{\text{PP_Glo}}$ obe | $\text{Bias}_{\text{TOP_TP}}$ | $\text{RMSE}_{\text{PP_Glo}}$ obe | $\text{RMSE}_{\text{TOP_TP}}$ | δ (%) |
|---|------------------------|----------------------------|----------------------|---------------------------------------|--------------------------------|---------------------------------------|--------------------------------|-----------------|
| Land surface albedo (α , Unitless) | CERES-EBAF Edition 4.2 | 0.70 | 0.72 | 0.11 | 0.10 | 0.14 | 0.13 | -7.4 |
| Net solar radiation (R_{net}^s , W/m ²) | CERES-EBAF Edition 4.2 | 0.15 | 0.20 | -9.03 | -6.94 | 24.41 | 22.88 | -23.2 |
| Latent heat flux (F_{lat} , W/m ²) | FLUXCOM | 0.81 | 0.81 | 15.69 | 15.98 | 17.36 | 17.58 | 1.9 |
| Sensible heat flux (F_{sen} , W/m ²) | FLUXCOM | -0.10 | -0.07 | -19.60 | -18.74 | 23.41 | 22.52 | -4.4 |
| Snow cover fraction (f_{sno}) | STC-MODSCAG and SPIRES | 0.37 | 0.40 | 0.22 | 0.20 | 0.28 | 0.27 | -6.7 |
| Air temperature (T_{air} , K) | UDel v5.01 | 0.56 | 0.57 | -2.36 | -2.10 | 5.69 | 5.50 | -11.1 |
| Precipitation (P , mm/day) | GPCP v2.3 | 0.80 | 0.81 | 1.46 | 1.44 | 1.88 | 1.86 | -1.2 |

264

265 3.2 Teleconnected impacts on East and South Asian air temperature and precipitation

266

267 Including radiation-topography interaction over the TP overall reduces the bias of T_{air} in the land regions of SA and EA. PP_Globe and TOP_TP show similarly high correlations with the UDel

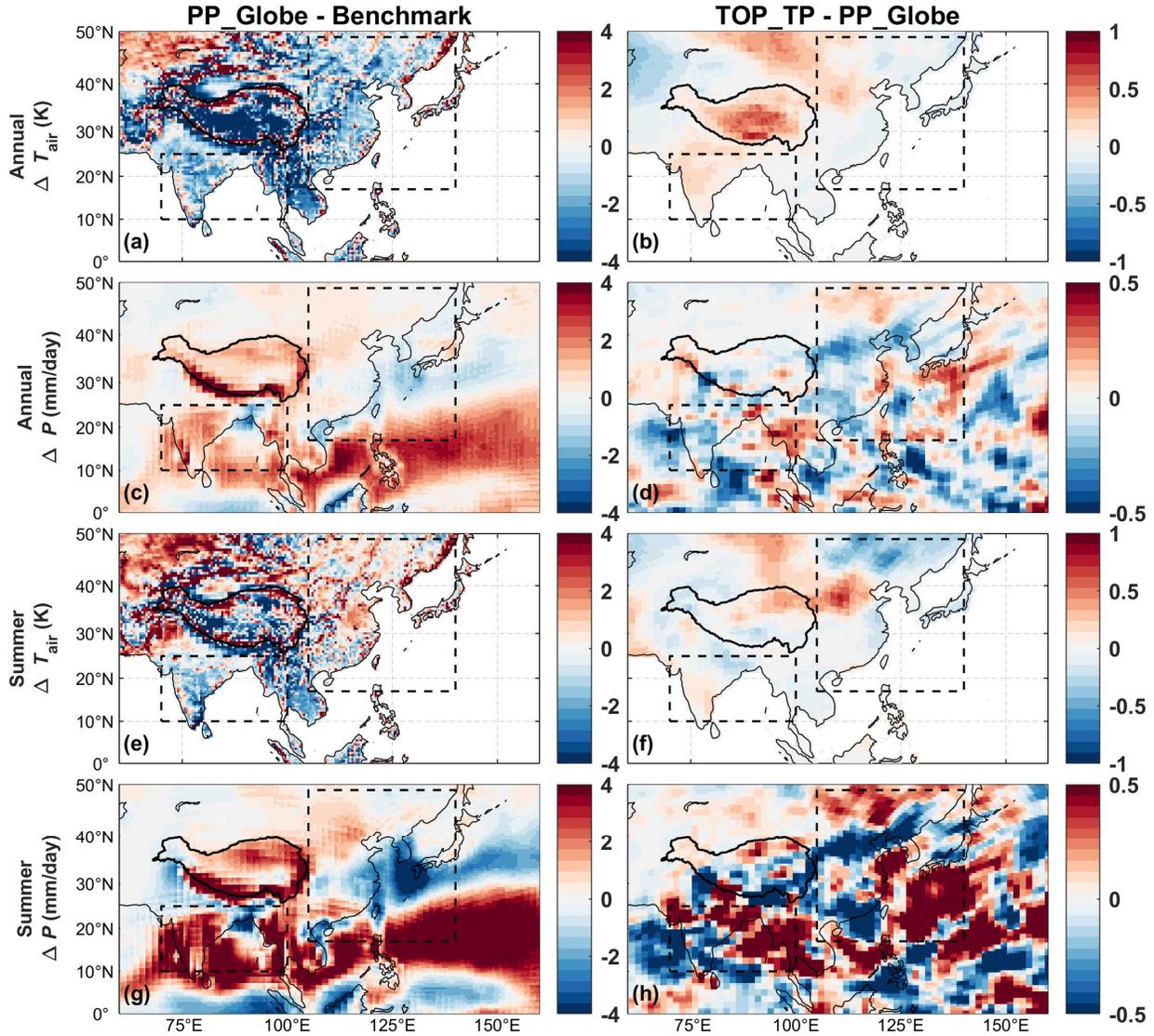
268 data across seasons with R values ≥ 0.70 and ≥ 0.89 , respectively in SA and EA (Table S4).
 269 PP_Globe shows cold biases in annual T_{air} over the TP's surrounding Asian regions (Figure 2a).
 270 For annual average in SA, PP_Globe has a cold bias of -1.28 K, while in EA, the cold bias is -
 271 0.90 K. For summer, PP_Globe has a cold bias of -0.89 K in SA, but a warm bias of +0.54 K in
 272 EA. Compared to PP_Globe, TOP_TP increases the annual and summer T_{air} in India, but shows
 273 small changes in other SA regions. The winter in SA shows larger reductions of T_{air} bias by 0.24
 274 K than other seasons (Table S4). In EA, TOP_TP increases annual T_{air} in north China, but
 275 reduces annual T_{air} in other EA regions. For summer, TOP_TP reduces the warm biases in
 276 northeast Asia, but increase the biases in north China. The summer warm bias in EA reduces
 277 from +0.54 K of PP_Globe to +0.49 K of TOP_TP.

278
 279 Radiation-topography interaction over the TP affects the P patterns in EA and SA, possibly
 280 through its influence on the atmospheric circulation. Overall the impacts of such interaction on
 281 annual and summer P show very heterogeneous spatial patterns in the Asian regions (Figure
 282 2d,h). In India and East China, TOP_TP overall shows smaller annual average P than PP_Globe
 283 (Figure 2d). In summer, TOP_TP reduces P in India, but increases it in the eastern regions of SA
 284 (Figure 2h). This is because the topography-induced wind anomaly weakens the climatological
 285 westerly wind and associated water transport from the Arabian Sea, while intensifying water
 286 transport from the continent to the Bay of Bengal (Figure S8). The summer P difference between
 287 TOP_TP and PP_Globe shows a tripolar structure of “north decrease–middle increase–south
 288 decrease” in East China (Figure 2h), which is a dominant pattern of P natural variability of the
 289 region [Xue *et al.*, 2023]. The tripolar pattern of changes in P is associated with the pattern wind
 290 changes at 850-hPa which enhance convergence of water vapor transport to central China while
 291 water vapor is diverged northward and southward in northern and southern China (Figure S8).

292
 293 Including the TP's radiation-topography interaction overall improves the P simulations in SA
 294 and EA. This is already apparent in Figure 2c,d for annual P and Figure 2g,h for summer P , as
 295 the differences between TOP_TP and PP_Globe generally have opposite signs compared to the
 296 difference between PP_Globe and the benchmark data. More specifically, PP_Globe shows high
 297 correlations to the GPCP especially in cold seasons (Table S4) for both SA and EA. PP_Globe
 298 overall overestimates annual P in SA with the wet mean bias of +0.95 mm/day, while it shows a
 299 small positive mean bias of +0.18 mm/day in EA. In summer, PP_Globe has larger
 300 overestimations over most SA regions with a mean bias of +2.53 mm/day, and the difference
 301 between PP_Globe and GPCP shows heterogeneous spatial distribution in EA (Figure 2g).
 302 Specifically, the difference in East China shows a “north wet– middle cold– south wet” tripolar
 303 structure in East China. Large positive and negative differences are found in the ocean regions of
 304 EA, while the land regions of EA generally show a small deviation from GPCP. By contrast,
 305 TOP_TP shows higher R values of 0.66 and 0.83 respectively in summer and autumn than
 306 PP_Globe (0.62 and 0.74, respectively) in SA, while TOP_TP has similar R values with
 307 PP_Globe across seasons in EA. TOP_TP overall reduces the wet bias in India (Figure 2d,h).
 308 Note that the summer spatial patterns between the difference of TOP_TP and PP_Globe and the
 309 difference between of PP_Globe and GPCP are opposite in East China (Figure 2g-h), which
 310 demonstrates that TOP_TP reduces the summer P biases in East China with the tripolar structure.

311
 312 Although the differences between TOP_Globe and PP_Globe overall show similar patterns to
 313 that between TOP_TP and PP_Globe, there are some large differences especially in India (Figure

314 S11). These demonstrate that the radiation-topography interactions in non-TP regions also affect
 315 the climate of the Asian regions.
 316



317

318 **Figure 2. Teleconnected impacts of radiation-topography interaction on annual and**
 319 **summer air temperature (T_{air}) and precipitation (P).** Panels (a,c,e,f) are the differences
 320 between PP_Globe and Benchmark datasets, and Panels (b,d,f,h) are the differences between
 321 TOP_TP and PP_Globe. For each panel, the black solid line is the boundary of TP, and the two
 322 black dashed lines are the boundaries of South and East Asia regions defined in Section 2.3.

323 4 Discussion and Conclusions

324 Radiation-topography interaction plays an important role in regulating regional climate. By using
 325 land-atmosphere coupled sensitivity experiments based on E3SM, we demonstrate that radiation-
 326 topography interaction can influence the TP's surface energy balance by reducing α and f_{sno}
 327 (Figure 1), which is consistent with the offline ELM simulations [Hao *et al.*, 2021]. Such

328 interaction further warms the regional near-surface atmosphere, modifies the clouds and affect
329 the local P (Figure 1). Accounting for such interaction in E3SM shows reduced cold biases over
330 the whole TP (Figure 2) especially in winter, which is in line with Lee et al. (2019) and Fan et al.
331 (2019). The interaction between radiation and topography is also expected to affect the glacier
332 evolution over the TP by accelerating glacier melt and retreat [Kraaijenbrink et al., 2017; Tang
333 et al., 2023].

334 Radiation-topography interaction over the TP could further affect the East and South Asian
335 climate. Due to the important role of TP as an “elevated heat pump”, the topography-induced
336 albedo change can affect the wind and moisture transport over SA and EA, and thus redistribute
337 P (Figure 2) over the Asian regions [Tang et al., 2023]. Specifically, the TP’s albedo change can
338 affect the intensities and movement of the South Asian High and West Pacific Subtropical High
339 [Tang et al., 2023] through the Rossby wave trains [Wang et al., 2008]. Our simulations show
340 that these changes are manifested in the iconic tripolar pattern change in P in East China which
341 is a dominant pattern of P variability in the region. The land surface temperature anomalies over
342 the TP have significant impacts on the East Asian summer monsoon precipitation [Diallo et al.,
343 2022]. Besides, the snow cover change in winter over the TP is also linked to the variation of
344 summer P in the downstream regions of China [Li et al., 2018; You et al., 2020]. All of these can
345 contribute to the change of regional seasonal P patterns and timing. However, the nonlinear
346 responses of Asian climate to the topography-induced albedo change and associated dominant
347 pathways need further investigations. It is noted that non-TP mountainous regions can also
348 contribute to the change of P patterns in the Asian regions (Figure S11), which needs further
349 analysis.

350 There are still large systemic biases in simulating surface climate over the TP and surrounding
351 regions, despite improved E3SM model performance against the benchmark datasets after
352 accounting for radiation-topography interaction. Such issues have been found in all the ESMs
353 (including E3SM) participating in the “Impact of Initialized Land Surface Temperature and
354 Snowpack on Subseasonal to Seasonal Prediction” project [Xue et al., 2021]. The large cold and
355 wet biases imply that there are some additional important physical processes over the TP but are
356 not well represented or even missing in E3SM and other ESMs. For example, the coupling of
357 convection to the large-scale environment needs to be improved to reduce the P biases in E3SM
358 [Zheng et al., 2019]. The T_{air} and P biases in E3SM further contribute to the uncertainties in
359 snowpack simulations [Brunke et al., 2021]. Besides, the LAP deposition over the TP shows
360 large impacts on the TP’s snow cover [Sarangi et al., 2020] and Asian monsoon climate [Qian et
361 al., 2011]. However, there is still limited knowledge on the snow grain shape and mixing state
362 between LAP and snow grain over the TP, which has been demonstrated to have large impacts
363 on TP’s energy balance and water cycle [Hao et al., 2023; He et al., 2018]. Better considering
364 the snow-aerosol-radiation interaction is necessary to reduce the uncertainties in simulating
365 climate over the TP and surrounding regions.

366 Our findings underscore the important regional and teleconnected impacts of radiation-
367 topography interaction over the TP. Improved understanding of the topographic roles stresses the
368 significance of parameterizing such important physical processes in CMIP6 models for future
369 climate projections. Neglecting such interaction will bias the simulations and projections of
370 surface energy balance, snow processes and surface climate over complex terrain and
371 surrounding regions.

372 **Acknowledgments**

373 This research was supported by the U.S. Department of Energy (DOE), Office of Science, Office
 374 of Biological and Environmental Research, Earth System Model Development program area, as
 375 part of the Climate Process Team projects. The reported research was conducted at the Pacific
 376 Northwest National Laboratory (PNNL), which is operated for the DOE by the Battelle
 377 Memorial Institute under contract DE-AC05-76RL01830. This research used the computational
 378 resources of the National Energy Research Scientific Computing Center, and DOE's Biological
 379 and Environmental Research Earth System Modeling program's Compy computing cluster
 380 located at PNNL.

381

382 **Open Research**

383 The codes of E3SMv2 are publicly available at <https://github.com/E3SM-Project> and in this
 384 study we used the git commit 8c716b9 of E3SM. Codes and data to reproduce all results and plot
 385 all figures are available at <https://doi.org/10.5281/zenodo.8327334>.

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