

1 **Intensified modulation of winter aerosol pollution in China**
2 **by El Niño with short duration**

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22 **Abstract**

23 El Niño-Southern Oscillation (ENSO), a phenomenon of periodic changes in sea
24 surface temperature in the equatorial central eastern Pacific Ocean, is the strongest
25 signal of interannual variability in the climate system with a quasi-period of 2-7 years.
26 El Niño events have been shown to have important influences on meteorological
27 conditions in China. In this study, the impacts of El Niño with different durations on
28 aerosol concentrations and haze days during December-January-February (DJF) in
29 China are quantitatively examined using the state-of-the-science Energy Exascale
30 Earth System Model version 1 (E3SMv1). We find that PM_{2.5} concentrations are
31 increased by 1-2 μg m⁻³ in the northeastern and southern China and decreased by up
32 to 2.4 μg m⁻³ in central-eastern China during El Niño events relative to the
33 climatological means. Compared to long duration (LD) El Niño events, El Niño with
34 short duration (SD) but strong intensity causes northerly wind anomalies over
35 central-eastern China, which is favorable for aerosol dispersion over this region.
36 Moreover, the anomalous southeasterly winds weaken the wintertime prevailing
37 northwesterly in northeastern China and facilitate aerosol transport from South and
38 Southeast Asia, enhancing aerosol increase in northeastern China during SD El Niño
39 events relative to LD El Niño events. In addition, the modulation on haze days by SD
40 El Niño events is 2-3 times more than that by LD El Niño events in China. The
41 aerosol variations during El Niño events are mainly controlled by anomalous aerosol
42 accumulation/dispersion and transport due to changes in atmospheric circulation,
43 while El Niño-induced precipitation change has little effect. The occurrence frequency
44 of SD El Niño events has been increasing significantly in recent decades, especially
45 after 1940s, suggesting that El Niño with short duration has exerted increasingly
46 intense modulation on aerosol pollution in China over the past few decades.

47

48 **1. Introduction**

49 Since the beginning of the 21st century, China has experienced frequent events of
50 heavy haze pollution (Yang et al., 2018). The excessive aerosol concentrations during
51 the heavy haze events can cause a large decrease in atmospheric visibility (Han et al.,
52 2013) and pose significant public health hazards, such as a dramatic increase in
53 cardiovascular and respiratory diseases and associated mortality rates (Liu et al.,
54 2019). PM_{2.5} (particulate matter less than 2.5 µm in diameter) has been reported to be
55 the fifth leading risk factor for mortality, inducing 7.6% of total deaths globally in
56 2015 (Cohen et al., 2017). In order to alleviate air pollution, a comprehensive and
57 better scientific understanding of factors that can affect aerosol concentrations and
58 haze pollution in China is required.

59 Undoubtedly, the rise of anthropogenic emissions is the fundamental reason for
60 the increase in aerosol concentration and haze pollution events (Huang et al., 2014),
61 but the unfavorable meteorological condition, as one of the most important external
62 factors, has been reported to have substantial influences on haze formation (Yang et
63 al., 2016a; Wang et al., 2019, 2020a). With increasing greenhouse gases in the future
64 (2050-2099), severe winter haze events in Beijing would become 50% more frequent
65 and 80% longer in duration, compared to the historical period (1950-1999), due to an
66 accelerated warming of the lower atmosphere and weakening of the East Asian winter
67 monsoon (Cai et al., 2017). In addition, external forcings, such as Pacific Decadal
68 Oscillation (Zhao et al., 2016) and Arctic sea ice (Wang et al., 2015; Zou et al., 2020),
69 all have important impacts on aerosol concentrations and haze pollution in China. El
70 Niño-Southern Oscillation (ENSO), as another prominent climate phenomenon
71 caused by the coupled atmosphere-ocean interactions in the tropical Pacific Ocean
72 (Trenberth, 2019), is a significant signal of interannual climate change on a global
73 scale. It triggers atmospheric circulation and precipitation anomalies globally (Yang et
74 al., 2016b, 2016c) and certainly has an important impact on haze events and aerosol
75 concentrations in China by modulating the East Asian winter monsoon system (Sakai
76 and Kawamura, 2009; Wang et al., 2000; Zhang et al., 2017).

77 The ENSO cycle is composed of warm-phase (i.e., El Niño) and cool-phase (i.e.,
78 La Niña) of sea surface temperatures (SSTs) over the tropical eastern Pacific Ocean,
79 which further cause precipitation, atmospheric circulation and temperature anomalies
80 in much of the tropics and subtropics. Such changes also affect the spatiotemporal

81 distribution of aerosols in China (Feng et al., 2017, 2020; Sun et al., 2018; Yang et al.,
82 2014; Zhao et al., 2018; Zhu et al., 2012; Wang et al., 2020b). During a strong El
83 Niño event in 2015/2016, PM_{2.5} concentrations in winter were observed to increase by
84 20-100 µg/m³ in eastern China compared to that in 2014, which was attributed to the
85 weakened wind speed in the North China Plain during the El Niño event (Chang et al.,
86 2016; Wang et al., 2020a). PM_{2.5} concentrations in southern China were also
87 decreased by about 20 µg/m³ during the 2015/2016 El Niño event, which was
88 attributed to an enhanced precipitation and aerosol wet scavenging over this region.
89 Many studies counted haze days based on atmospheric visibility and found that El
90 Niño events could induce more (fewer) winter haze days in northern (southern) China
91 (Gao and Li, 2015; Li et al., 2017). In addition to surface observations, several studies
92 have also analyzed the relationship between ENSO events and aerosol loading based
93 on aerosol optical depth (AOD) data from satellite retrievals (Jeoung et al., 2014; Sun
94 et al., 2018). Jeoung et al. (2014) analyzed the combined AOD data of MODIS
95 (Moderate Resolution Imaging Spectroradiometer), MISR (Multi-angle Imaging
96 SpectroRadiometer) and AERONET (Aerosol Robotic Network) and found that
97 during the warm phase of ENSO, the fine-mode AOD increased in eastern coastal
98 areas but decreased in some inland areas of China. Sun et al. (2018) studied the
99 influence of ENSO events on the interannual variation of wintertime aerosol in China
100 using AOD data (1980-2016) from MERRA-2 reanalysis and found that AOD in the
101 North China Plain increased significantly during El Niño events, with a 15%
102 increment in the Beijing-Tianjin-Hebei region compared to the long-term average.
103 They also pointed out that AOD increased in eastern and southern China and
104 decreased in southwestern China during El Niño events.

105 Although observational data showed that aerosols in China were largely
106 perturbed during El Niño events, the individual impacts of atmospheric circulation
107 and precipitation anomalies associated with El Niño could not be simply extracted out
108 with observations alone. Numerical simulations have been used to isolate the
109 individual impacts of El Niño on aerosols in China through a superposed SST
110 perturbation method and explore the underlying mechanisms (Yu et al., 2019; Zhao et
111 al., 2018). Based on an aerosol-climate coupling model, Zhao et al. (2018) suggested
112 that El Niño increased the seasonal mean aerosol concentration in southern China in
113 winter, which is mainly due to the increased aerosol transport from South and
114 Southeast Asia. Using the same model, Yu et al. (2019) showed that, relative to the

115 climatological mean, wintertime surface aerosol concentrations in northeastern and
116 southeastern China (central and southwestern China) increased (decreased) during El
117 Niño events, which was mainly attributed to anomalies in near-surface winds and the
118 resulting aerosol mass flux divergences. Sun et al. (2018) used the aerosol-climate
119 model CAM5 to simulate the impact of ENSO events on the interannual variability of
120 AOD in China and found that El Niño events led to an increase in AOD in central and
121 eastern China. They suggested that the change in AOD was mainly dominated by the
122 change in meridional winds.

123 Some studies focused on the effects of different spatial types (e.g., East Pacific
124 and Central Pacific El Niño, Kao and Yu (2009)) and intensities of El Niño events on
125 aerosol concentrations in China (e.g., Yu et al., 2019). Yu et al. (2019) found that, due
126 to the difference in atmospheric circulation between two types of El Niño, Central
127 Pacific El Niño events resulted in a larger increase in aerosol burden in southern
128 China than East Pacific El Niño events. They also indicated that a moderate El Niño
129 event led to an increase in seasonal mean near-surface aerosol concentrations
130 throughout eastern China in winter, while a strong or weak El Niño event brought
131 about a significant decrease in aerosol concentrations in northern China.

132 Apart from the spatial types with different intensities, El Niño can also be
133 categorized as short duration (SD) and long duration (LD) according to the length of
134 their decay period (Boo et al., 2004; Chen et al., 2012; Guo and Tan, 2018). These two
135 temporal types of El Niño events have been confirmed to have different impacts on
136 the SSTs, vertical wind shear, relative humidity and precipitation in South China Sea
137 and Philippine Sea (Guo and Tan, 2018; Wu et al., 2019). The El Niño events with
138 different durations are likely to have different impacts on the aerosol distribution in
139 China. However, few studies explore the different impacts of SD and LD El Niño
140 events on aerosol concentrations and haze days in China, as well as the associated
141 mechanisms, which are essential for air pollution control in the near future.

142 In this study, the effects of SD and LD El Niño events on wintertime aerosols in
143 China are investigated by using the state-of-the-science Energy Exascale Earth
144 System Model (E3SM). The data, model, and analysis methods used in this research
145 are presented in Section 2. The influences of different durations of El Niño events on
146 aerosols over China and the mechanisms involved are analyzed in Section 3.
147 Summary of the main results and discussion of the implications for future research are
148 provided in Section 4.

149 **2. Data and Methods**

150 **2.1 Data**

151 We use the following datasets in this study.

152 (1) The merged Hadley-NOAA/OI SST and sea ice concentration (SIC) datasets
153 from 1870 to 2017 with a horizontal resolution of $1^\circ \times 1^\circ$ (Hurrell et al., 2008)
154 are used to obtain the climatological mean SST and SIC pattern and the
155 anomalies of SST during SD and LD El Niño events.

156 (2) Monthly mean emissions of aerosols and their precursors in 2014 from the
157 CMIP6 (Coupled Model Intercomparison Project Phase 6) (Hoesly et al.,
158 2018; van Marle et al., 2017) with emissions in China replaced by MEIC
159 (multi-resolution emission inventory for China) emission inventory are used
160 as input datasets in model simulations.

161 (3) Hourly observations of PM_{2.5} concentrations at 1657 stations over China
162 from December 2014 to February 2015 derived from the China National
163 Environmental Monitoring Centre (CNEMC) are applied to evaluate the
164 model performance.

165 (4) Monthly averaged ERA5 reanalysis data from 1950 to 2017 (Hersbach et al.,
166 2020) provided by European Centre for Medium-Range Weather Forecasts
167 (ECMWF) are used to evaluate the simulated meteorological parameters
168 during El Niño events.

169 **2.2 SD and LD El Niño events**

170 Here we first describe how the LD and SD El Niño events are defined. The year
171 in which El Niño developed is denoted by year⁰ and the months of that year are
172 denoted by Jan⁰, Feb⁰, ..., and Dec⁰, while the following year and months are year¹
173 and Jan¹, Feb¹, ..., and Dec¹, respectively. Niño 3.4 index is detrended SST anomaly
174 over the Niño 3.4 region (170°W-120°W, 5°S-5°N). El Niño event is firstly identified
175 when a 3-month running mean Niño 3.4 index is greater than 0.75°C in any month
176 from Oct⁰ to Feb¹ of its developing phase. If the Niño 3.4 index is higher than 0.5°C
177 in any month from Oct¹ to Feb² of its decaying phase, the El Niño event is an LD El
178 Niño event; otherwise, it is an SD El Niño event (Wu et al., 2019).

179 Figure 1 shows the time series of the Niño3.4 indices calculated based on the
180 Hadley-NOAA/OI data for the period 1870-2017. According to the definition above,
181 totally 30 El Niño events are identified in this time period, with 22 SD El Niño events

182 (1877/1878, 1885/1886, 1888/1889, 1896/1897, 1902/1903, 1911/1912, 1923/1924,
183 1925/1926, 1930/1931, 1951/1952, 1957/1958, 1963/1964, 1965/1966, 1972/1973,
184 1982/1983, 1991/1992, 1994/1995, 1997/1998, 2002/2003, 2006/2007, 2009/2010,
185 2015/2016 as developing phase) and 8 LD El Niño events (1899/1900, 1904/1905,
186 1913/1914, 1918/1919, 1939/1940, 1968/1969, 1976/1977, 1986/1987 as developing
187 phase). The temporal evolution of the Niño3.4 indices during the developing and
188 decaying phases of SD and LD El Niño events is shown in Figure 2. During the
189 developing phase from Jul⁰ to Feb¹, due to the fast accumulation of ocean heat content
190 and rapid adjustments of the surrounding seas to the tropical Pacific Ocean warming
191 (Wu et al., 2019), the Niño3.4 indices are higher in SD El Niño events, but the SST
192 anomaly decreases rapidly in the decaying phase, compared to those in LD events.

193 **2.3 Model description and experimental design**

194 To quantify the influence of El Niño with various durations on aerosols in China,
195 the U.S. Department of Energy (DOE) E3SM version 1 (E3SMv1) is utilized in this
196 study, which includes atmosphere, land, ocean, sea ice and river components (Golaz et
197 al., 2019). It was branched from the CESM1 (Community Earth System Model) but
198 has been updated substantially since. The E3SM Atmosphere Model version 1
199 (EAMv1) is a descendant of the well-known CAM5.3 (Community Atmosphere
200 Model version 5.3) (Rasch et al., 2019). It includes considerable upgrades to aerosols,
201 turbulence, chemistry and cloud related processes. EAMv1 provides various options
202 of spatial resolution. In this study, the horizontal spatial resolution of approximately
203 1° and 30 vertical layers from the surface to 3.6 hPa are used in the model
204 configuration. The model simulates aerosols including sulfate, black carbon (BC),
205 primary organic aerosol (POA), secondary organic aerosols (SOA), sea salt, and
206 mineral dust in the four-mode Modal Aerosol Module (MAM4) (Wang et al., 2020).

207 The following numerical experiments are conducted in this study. A “CLIM”
208 experiment is driven by climatological average of monthly SST and SIC over
209 1870-2017 and integrated for 20 years. Two sets of sensitivity experiments, “SD” and
210 “LD”, are respectively driven by the monthly SST representing composite SD and LD
211 El Niño events. The monthly SSTs representing SD (LD) El Niño events are produced
212 through superposing the average of monthly SSTs from Jul⁰ to Jun¹ of the 22 SD (8
213 LD) El Niño events selected in Sec. 2.2 on top of the climatological monthly SST
214 over 60°S-60°N. Each set of sensitivity experiment has 3 ensemble members with
215 different initial conditions branched from different years of the CLIM experiment.

216 Each member of the sensitivity experiments is run for 8 years with first 3 years used
217 for spin-up and the last 5 years used for analysis. The differences in the monthly and
218 daily mean model fields between SD, LD and CLIM are used to analyze the effects of
219 duration of El Niño events on aerosols. To understand the potential mechanism of El
220 Niño impacts on aerosol pollution in China, two additional experiments are also
221 conducted. The “SD_emis” experiment is the same as the first ensemble member of
222 SD experiment, except that the emissions of aerosols and precursor gases from South
223 and Southeast Asia are turned off. The “CLIM_emis” experiment is same as the
224 “SD_emis” experiment but driven by climatological average of monthly SST and SIC
225 over 1870-2017. All other external forcings, including insolation, greenhouse gas
226 concentrations, and emissions of aerosol and precursor are kept at present-day
227 conditions (year 2014), In brief, the experiments performed are as follows (Table 1).

- 228 1. CLIM: control simulation driven by climatological SST.
- 229 2. SD: sensitivity simulation to quantify the impacts of El Niño events with
230 short duration on aerosols in China. Same as CLIM except for the imposed
231 SST pattern of short duration El Niño (Fig. 3a).
- 232 3. LD: sensitivity simulation to quantify the impacts of El Niño events with
233 long duration on aerosols in China. Same as CLIM except for the imposed
234 SST pattern of long duration El Niño (Fig. 3b).
- 235 4. SD_emis: sensitivity simulation to quantify the role of regional transport of
236 aerosols from South and Southeast Asia on aerosols in China during El Niño
237 events with short duration. Same as SD except that the emissions of aerosols
238 and precursor gases from South and Southeast Asia are turned off.
- 239 5. CLIM_emis: sensitivity simulation to serve as the baseline for SD_emis.
240 Same as SD_emis except for the use of climatological SST.

241 **2.4 Model evaluation**

242 To evaluate the model performance in simulating aerosol concentration and
243 distribution in China, the simulated December-January-February (DJF) mean surface
244 PM_{2.5} (sum of sulfate, BC, POA and SOA in model simulation) concentrations from
245 the CLIM experiment is compared with the observed PM_{2.5} concentrations in Fig. 4.
246 The model well reproduced the spatial distribution of wintertime aerosols in China,
247 with high aerosol concentrations in eastern China (e.g., North China Plain, Fenwei
248 Plain and Yangtze River Delta) and southwestern China (e.g., Sichuan Basin) and low
249 aerosol levels in western China. The spatial correlation coefficient (R) between the

250 E3SMv1 EAMv1 simulation and observations for near-surface PM_{2.5} concentrations is
251 +0.43. However, the model underestimates the PM_{2.5} concentrations in China, with a
252 normalized mean bias (NMB) of -65.74% compared to the observed values, which
253 was also reported in many studies using the CESM1 model (e.g., Yang et al., 2017a,
254 b). The discrepancy could be due to many factors, including the lack of nitrate and
255 ammonium aerosols in the model, strong wet scavenging simulated at the mid- and
256 high latitudes, and less transformation from gas to particles. In addition, we focus on
257 anthropogenic aerosols. If natural dust is considered in the modeled PM_{2.5} calculation,
258 the NMB will drop to -6.38%. Nevertheless, the aerosol concentrations in EAMv1
259 simulations are closer to the observations than previous ENSO-aerosol studies (e.g.,
260 Yu et al., 2019; Zhao et al., 2018) and the composite differences are analyzed in this
261 study rather than the climatological mean concentration. We don't expect the
262 systematic low biases in PM_{2.5} concentrations affect our study on the impact of El
263 Niño events. However, we should note that the aerosol changes in China during
264 SD/LD El Niño events in the real world could be larger than the simulated values
265 here.

266

267 **3. Results**

268 **3.1 Impacts of SD and LD El Niño events on aerosol concentrations**

269 Figure 5 show the absolute and relative impacts of the two types of El Niño
270 events with different durations on the simulated DJF mean near-surface
271 concentrations and column burdens of PM_{2.5} in China. The effects of the SD and LD
272 El Niño events on near-surface aerosol concentrations over China are similar in the
273 spatial pattern distribution, with increases in the northeastern and southern China by
274 about 1-2 $\mu\text{g m}^{-3}$ (5-15% compared to the climatological mean) and decreases in
275 central-eastern China during El Niño events relative to the climatological averages.
276 This spatial pattern of aerosol changes is in accordance with previous modeling
277 studies (Feng et al., 2016; Yu et al., 2019; Zhao et al., 2018). However, the modeling
278 results are not exactly the same as the observed PM_{2.5} changes, which show increases
279 in PM_{2.5} over northeastern China and the North China Plain and slightly decreases in
280 southern China during the 2015/2016 El Niño event (Chang et al., 2016). The
281 discrepancy between the model simulations and observations can be attributed to the
282 following reasons. First of all, instead of the El Niño impacts, observed aerosols can

283 be affected by other factors including East Asian winter monsoon (Yang et al., 2016a),
284 Arctic Oscillation (Zhang et al., 2019) and Pacific Decadal Oscillation (Zhao et al.,
285 2017), whereas the modeled changes are purely caused by the El Niño impacts
286 through the imposed SST perturbation. Secondly, the time coverage of near-surface
287 PM_{2.5} observations is limited in China and only one extreme El Niño event
288 (2015/2016) was analyzed in previous El Niño-PM_{2.5} studies (e.g., Chang et al., 2016),
289 which is not fully representative of the impact of general El Niño events.

290 Although the spatial patterns of the SD and LD El Niño influences on the DJF
291 PM_{2.5} concentrations in China resemble each other, the magnitudes of the influences
292 are different. Central-eastern China experiences more reductions in near-surface PM_{2.5}
293 concentrations during SD El Niño, with the concentration decreases of more than 2.4
294 $\mu\text{g m}^{-3}$ (15% relative to the climatological mean), which is much larger than the 0.6
295 $\mu\text{g m}^{-3}$ (5%) during LD El Niño. In southern China, the spatial coverage of the
296 increase in PM_{2.5} concentration shrinks more during SD than LD El Niño events
297 relative to the CLIM, but the intensities of the anomalies triggered by the two
298 temporal types of El Niño events are similar. Moreover, SD El Niño induces a larger
299 increase in PM_{2.5} concentrations in northeastern China by 1.2 $\mu\text{g m}^{-3}$ (10%) than that
300 of 0.6 $\mu\text{g m}^{-3}$ (5%) during LD El Niño events.

301 The PM_{2.5} burden and near-surface concentration anomalies triggered by the El
302 Niño events with short and long durations are basically the same in spatial distribution
303 but with different magnitudes (Fig. 5). For example, the reduction in aerosol burden is
304 much larger in central-eastern China during the SD El Niño events than during the LD
305 El Niño events, with maximum negative anomalies, respectively, reaching -1.6 and
306 -0.6 mg m^{-2} . Overall, SD El Niño events yield stronger impacts on aerosol pollution in
307 China than LD El Niño events, especially in central-eastern China with negative
308 pollution anomalies.

309 **3.2 Mechanisms of SD and LD El Niño impacts on aerosols**

310 Since aerosols and their precursor gas emissions are prescribed at the same rates
311 in the control (CLIM) and sensitivity (SD/LD) simulations, changes in meteorological
312 factors such as circulation and precipitation play dominant roles in altering aerosol
313 concentrations by influencing the regional transport and wet removal of aerosols
314 (Yang et al., 2016a). Previous studies also suggested that aerosol variations during
315 ENSO events were controlled by ENSO-related circulation and precipitation changes
316 (Yu et al., 2019; Zhao et al., 2018). Here, we examine the atmospheric circulation and

317 precipitation anomalies and the associated aerosol processes during the SD and LD El
318 Niño to explore the mechanisms of the two types of El Niño effects on aerosols in
319 China.

320 Both the SD and LD El Niño events trigger negative anomalies in sea level
321 pressure (SLP) over the eastern China and East China Sea and positive anomalies
322 over the Philippine Sea and Sea of Okhotsk (not shown), leading to anomalous
323 cyclonic and anticyclonic circulations over these regions, respectively (Figs. 6a and
324 6b). At 850 hPa, the anomalous cyclonic circulation over the East China Sea causes
325 anomalous northerly winds over central-eastern China, enhancing the prevailing
326 northwesterly winds in winter. The enhanced winds favor the aerosol dispersion,
327 which explains the decrease in PM_{2.5} concentrations over central-eastern China during
328 El Niño events relative to the climatological mean. In addition, at 500 hPa, most areas
329 over China have an anomalous low pressure (Figs. 6d and 6e), which increases the
330 atmospheric instability and strengthens the aerosol vertical mixing and dispersion
331 over central-eastern China. Over southern China, the aerosol variations are
332 significantly affected by the regional transport of particles from South and Southeast
333 Asia. During El Niño events, anomalous southwesterly winds at the northwest edge of
334 the anomalous anticyclone over the Philippine Sea bring aerosols from South and
335 Southeast Asia to southern China, contributing to the aerosol increases in southern
336 China relative to the climatological mean (Figs. 6a and 6b). In the northeastern China,
337 anomalous southeasterly winds associated with the anomalous anticyclonic circulation
338 over Sea of Okhotsk weaken the wintertime prevailing northwesterly winds, giving
339 rise to the aerosol increases in the northeastern China during El Niño events. In
340 addition, the anomalous anticyclone brings aerosols from South and Southeast Asia to
341 northeastern China that will be discussed next, contributing to the aerosol pollution in
342 northeastern China during El Niño events.

343 Compared to LD El Niño events, the negative anomaly of SLP over the East
344 China Sea during the SD El Niño events is stronger and extends deeply into the
345 central-eastern China, resulting in anomalous northerly winds over central-eastern
346 China and southeasterly winds over northeastern China in the lower atmosphere (Fig.
347 6c). The wind anomalies intensify the aerosol dispersion in central-eastern and
348 accumulation in northeastern China, leading to a stronger effect of El Niño with short
349 duration on the aerosol variation in China. Furthermore, the anomalous northerly
350 winds in both lower atmosphere and 500 hPa over southern China (Figs. 6c and 6f)

351 are un conducive to the regional transport of aerosols from South and Southeast Asia
352 to central-eastern China.

353 To further verify the model simulations in capturing atmospheric circulation
354 anomalies during SD and LD El Niño events, the wind fields are compared with those
355 from ERA5 reanalysis data. The anomalous atmospheric circulation patterns in the
356 latest SD El Niño event (2015/2016) and LD El Niño event (1986/1987) relative to
357 the climatological mean (1950-2017) from the ERA5 are shown in Fig. 7. Overall, the
358 SD and LD El Niño-induced anomalous atmospheric circulations over China
359 simulated in E3SM are in consistent with the reanalysis data. Both of them show the
360 anomalous northerly winds over central-eastern China at 850 hPa during SD El Niño
361 compare to LD El Niño. In addition, obvious anomalous cyclone at 500 hPa over most
362 of China can be seen in both E3SM and ERA5.

363 In addition to the regional transport prompted by anomalous atmospheric
364 circulations, El Niño can influence aerosol wet removal through perturbing
365 precipitation. As described in Figs. 8a and 8b, the spatial patterns of winter
366 precipitation anomalies in China during SD and LD El Niño events are similar, with
367 positive anomalies located along the southeastern coastal areas due to the additional
368 moisture transport by anomalous southwesterly winds over the South China Sea.
369 However, the two types of El Niño events differ in the magnitude of precipitation
370 anomalies. In central-eastern China, precipitation decreases during SD El Niño events,
371 compared to LD El Niño events, whereas precipitation increases over eastern coastal
372 areas and northeastern China (Fig. 8c). This is linked to the anomalous cyclonic
373 circulation over central-eastern China (Fig. 6c), which hinders moisture from South
374 China Sea to central-eastern China but brings in moisture from Sea of Japan to
375 northeastern China. Over Pearl River Delta, precipitation decreases during SD El
376 Niño events but increases during LD El Niño events, which is also associated with the
377 anomalous northerly winds and corresponding impact on moisture transport over this
378 region. In general, aerosol wet deposition decreases in central-eastern China and
379 increases over southern and northeastern China during El Niño events (Figs. 8d and
380 8e). With short duration but strong intensity, El Niño events have larger impacts on
381 aerosol wet removal than those with long duration (Fig. 8f). However, the wet
382 removal shows a positive relationship with the aerosol concentration, which should be
383 a negative relationship in theory if other conditions remain unchanged. Water vapor
384 can accelerate the chemical transformation of secondary aerosols (Yang et al., 2015),

385 but the primary aerosols showed the same spatial differences in near-surface
386 concentration as PM_{2.5} (not shown), indicating that change in water vapor is also not
387 the main reason for the aerosol changes in China during El Niño events. Therefore,
388 the differences in aerosols triggered by El Niño events with different durations are
389 primarily due to the impact of changes in atmospheric circulation on the accumulation
390 and transport of aerosols rather than the impact of precipitation on aerosol removal.

391 Both the accumulation/dispersion of local aerosols within China and regional
392 transport of aerosols from South and Southeast Asia can contribute to the aerosol
393 changes in China during El Niño events. With emissions of aerosols and precursor
394 gases in South and Southeast Asia turned off, the decrease pattern of PM_{2.5} over
395 central-eastern China does not change (Fig. 9), suggesting that
396 accumulation/dispersion of local aerosols dominates the aerosol change over this
397 region during El Niño events. Over southern China, the increase of PM_{2.5} burden is
398 weakened when the South and Southeast Asian emissions are turned off, indicating
399 that regional transport of aerosols from South and Southeast Asia have a large
400 contribution to the aerosol variation over this region. It is interesting that, without
401 emissions from South and Southeast Asia, both near-surface concentration and
402 column burden of PM_{2.5} in northeastern China decrease during El Niño events relative
403 to the climatological mean, but the change reverses to increase when the South and
404 Southeast Asian emissions are considered. It indicates that the aerosol enhancements
405 in northeastern China during the El Niño events are most likely influenced by aerosol
406 transport from South and Southeast Asia due to anomalous southeasterly winds at the
407 eastern edge of the anomalous cyclonic circulation in eastern China (Fig.6c), which
408 warrants further analysis using a source-receptor model such as CAM5-EAST (Ren et
409 al., 2020; Yang et al., 2020).

410 **3.3 Quantitative impacts on regional PM_{2.5} concentrations and haze** 411 **days**

412 Figure 10 summarizes the simulated probability density distributions of PM_{2.5}
413 concentrations, regional mean PM_{2.5} concentrations and number of haze days in DJF
414 over the sub-regions in China, including the North China Plain (NCP, 35–41°N,
415 114–120°E), Sichuan Basin (SCB, 28–33°N, 103–108°E), Yangtze River Delta (YRD,
416 29–34°N, 118–121.5°E), Pearl River Delta (PRD, 21.5–25°N, 111–116°E), Northeast
417 Plain (NEP, 41–48°N, 120–130°E), the Yunnan-Guizhou Plateau (YGP, 23–27°N,
418 100–110°E), and the Fenwei Plain (FWP, 33–35°N, 106–112°E and 35–38°N,

419 110–114°E) from CLIM, SD and LD simulations. Haze days are defined as days with
420 daily near-surface PM_{2.5} concentrations above the 90th percentile of the CLIM PM_{2.5}
421 concentrations in each sub-region of China.

422 During El Niño events, DJF mean near-surface PM_{2.5} concentrations decrease
423 over NCP, SCB, YGP, and FWP regions and increase over PRD and NEP in both SD
424 and LD, compared to CLIM. Although the PM_{2.5} concentrations show an increase in
425 SD and a decrease in LD over YRD region, the changes are statistically insignificant
426 in this region (Fig. 5). SD El Niño events have a stronger modulation on aerosols in
427 China than LD El Niño events. Over the regions with concentration decreases (NCP,
428 SCB, YGP, and FWP), regional mean near-surface PM_{2.5} concentration in LD is lower
429 than CLIM by 0.24 μg m⁻³, while the reduction reaches 1.22 μg m⁻³ in SD, about 5
430 times as that of LD. Over the regions with concentration increases (PRD and NEP),
431 the PM_{2.5} increase in SD relative to CLIM is 0.74 μg m⁻³, which is also higher than
432 the 0.56 μg m⁻³ in LD.

433 Similar to the PM_{2.5} concentration, the modulation of SD El Niño events on haze
434 days are 2-3 times as high as that of LD El Niño events. During LD El Niño events,
435 the number of haze days in DJF at NCP, SCB and FWP is reduced by 1.14, 0.73 and
436 1.53 days, respectively, compared to the climatological mean, while the decrease in
437 haze days during SD El Niño events is more substantial (1.87, 2.13 and 2.87 days).
438 The probability density distributions of PM_{2.5} concentrations over NCP, SCB and
439 FWP in SD and LD also shift to the left, relative to CLIM (Figs. 10b, 10g, and 10h).
440 Consistent with the stronger modulation of SD El Niño events discussed above, the
441 shift in SD is more than that in the LD simulation. In addition, YRD, PRD and NEP
442 regions all have increases in haze days in DJF during SD and LD, relative to CLIM.
443 Similarly, during SD and LD El Niño events, the probability density distributions of
444 high values of PM_{2.5} concentrations over YRD, PRD and NEP slightly shift to the
445 right relative to CLIM (Figs. 10c-e). The number of haze days in DJF over YGP
446 decreases during SD El Niño events by 1.4 days, but there is a slight increase of 0.4
447 days during LD El Niño events, likely due to the opposite aerosol changes in the
448 eastern and western parts of YGP region (Fig. 5). There are more (fewer) haze days in
449 both SD and LD than in CLIM over YRD, PRD and NEP (NCP, SCB and FWP),
450 which is inconsistent with the simulated greater (less) precipitation over these regions
451 caused by El Niño events. It further indicates that anomalies in precipitation are not
452 the most dominant factor modulating winter haze days in China during El Niño events,

453 but rather the anomalous aerosol accumulation/dispersion and transport due to
454 anomalous atmospheric circulation.

455 **3.4 Historical increase in SD El Niño events**

456 Many studies have suggested an increase in the variability of El Niño events
457 under greenhouse warming (Cai et al., 2018; Grothe et al., 2020). However, few
458 studies have shown the historical changes in El Niño with different durations, which
459 would further impact aerosol concentrations and haze days in China.

460 Here we show the occurrence of SD and LD El Niño events since the
461 preindustrial era in Fig. 11. The number of SD El Niño events fluctuated but has
462 increased significantly during the past few decades, especially after 1940s. The
463 occurrence of SD El Niño increased from one event per fifteen years during
464 1941–1955 to four events per fifteen years during 2001–2015, with the increase at
465 confidence level of 92%, while LD El Niño events did not present a significant trend
466 in the historical period. Wu et al. (2019) found that the duration of El Niño is mainly
467 influenced by the timing of onset, associated with the early onset of delayed negative
468 oceanic feedback as well as the fast adjustments of the tropical Indian and Atlantic
469 Oceans to the tropical Pacific Ocean warming. It is conjectured that the onset timing
470 of El Niño events gets earlier under greenhouse forcing, but the detailed analysis is
471 out of the scope of this study. Nevertheless, because the frequency of the El Niño
472 events with short duration increased significantly, the modulation by El Niño events
473 on wintertime aerosols in China has intensified in the past few decades.

474

475 **4. Conclusion and discussions**

476 As a prominent climate phenomenon, El Niño triggers atmospheric circulation
477 and precipitation anomalies on a global scale, thus having important effects on haze
478 days and aerosol pollution in China. In this study, the impacts of different temporal
479 types of El Niño events with short and long duration on aerosols in China are
480 simulated using the state-of-the-science E3SM model.

481 For both SD and LD El Niño events, their changes to the DJF mean PM_{2.5}
482 concentrations have similar spatial distributions over China, relative to the
483 climatological mean. The anomalous anticyclonic circulation over the Sea of Okhotsk
484 weakens the prevailing northwesterly winds in DJF in northeastern China and
485 enhances the accumulation of locally emitted aerosols, along with the anomalous

486 southeasterly winds at the eastern edge of the anomalous cyclonic circulation in
487 eastern China that intensifies the aerosol transport from South and Southeast Asia to
488 northeastern China. The near-surface $PM_{2.5}$ concentration in northeastern China
489 increases by 1-2 $\mu\text{g m}^{-3}$ during El Niño events relative to the climatological conditions.
490 In southern China, the anomalous anticyclonic circulation over the Philippine Sea
491 facilitates the transport of aerosols from South and Southeast Asia to southern China
492 and thus the near-surface $PM_{2.5}$ concentrations in southern China increase by 1-2 μg
493 m^{-3} . The decrease in near-surface $PM_{2.5}$ concentrations in central-eastern China is
494 mainly controlled by the enhanced northerly winds from the anomalous cyclonic
495 circulation over eastern China and the East China Sea, leading to the dispersion of
496 local aerosols, while precipitation change has little effect on aerosols here. Compared
497 to LD El Niño events, due to the anomalous cyclonic circulation over eastern China,
498 SD El Niño events exhibit a stronger reduction (1-2 $\mu\text{g m}^{-3}$) in near-surface $PM_{2.5}$
499 concentrations over central-eastern China and a larger increase (0.6 $\mu\text{g m}^{-3}$) in
500 northeastern China. Overall, El Niño with short duration has a stronger modulation on
501 wintertime aerosols in China than El Niño with long duration.

502 Compared with CLIM, mean near-surface $PM_{2.5}$ concentrations in DJF decrease
503 over NCP, SCB, YGP and FWP regions and increase over PRD and NEP in both SD
504 and LD, but the decrease over these regions in SD El Niño events reaches 1.22 $\mu\text{g m}^{-3}$,
505 about 5 times as large as that of LD. Similarly, both SD and LD El Niño events induce
506 less (more) haze days in DJF than CLIM over NCP, SCB and FWP (YRD, PRD and
507 NEP). However, the decreases in haze days in DJF at NCP, SCB and FWP during SD
508 El Niño events are 2-3 times more than that during LD El Niño events.

509 We also found that the occurrence frequency of SD El Niño events increased
510 from one event per fifteen years during 1941–1955 to four events per fifteen years
511 during 2001–2015, whereas LD El Niño events did not exhibit a significant trend in
512 the historical period. In particular, seven SD El Niño events have occurred since the
513 1990s, but no LD El Niño event occurred. Compared to LD El Niño events, SD El
514 Niño events have a greater impact on wintertime aerosols over China. Therefore, the
515 impact of El Niño events on wintertime aerosols in China has intensified in the past
516 few decades due to their short durations.

517 Our results of the important effect of SD El Niño events and its recent
518 intensification are of great significance for the understanding of El Niño on China's
519 haze pollution, alleviating air pollution, and coping with climate change. The

520 simulated spatial patterns of aerosol changes during El Niño events resemble those in
521 previous studies (Feng et al., 2016; Yu et al., 2019; Zhao et al., 2018). However, there
522 are still some inadequacies remaining to be improved. Natural aerosols including dust
523 and sea salt were not considered in this study. The EAMv1 model largely
524 underestimated PM_{2.5} concentration in China related to the lack of nitrate and
525 ammonium aerosols and other model biases. We also found that, during El Niño
526 events, more aerosols from South and Southeast Asia can be transported to
527 northeastern China, leading to an increase in aerosol concentrations over there. Thus,
528 more in-depth analysis is needed in future studies. In addition, during the cooling
529 phase of ENSO, La Niña events may also have various durations and can have
530 different impacts on air pollutions in China, which merits further investigation. Since
531 that PM_{2.5} is more harmful to human health than PM₁₀, in this study, we focused on
532 PM_{2.5} rather than PM₁₀, which is largely contributed by natural dust aerosol. The
533 impacts of El Niño on dust will be investigated in our future work.
534

535 ***Data availability***

536 The E3SMv1 model is available at <https://e3sm.org/> (last access: 25 May 2021). Our
537 results can be made available upon request. Hourly observations of PM_{2.5}
538 concentrations over China can be derived from the China National Environmental
539 Monitoring Centre (<http://www.cnemc.cn>, last access: 25 May 2021). Monthly
540 ERA5 reanalysis data can be downloaded at
541 <https://www.ecmwf.int/en/forecasts/datasets/reanalysis-datasets/era5> (last access: 25
542 May 2021)

543

544 ***Author contributions***

545 YY designed the research. LZ performed the model simulations and analyzed the data.
546 All the authors discussed the results and wrote the paper.

547

548 ***Competing interests***

549 The authors declare that they have no conflict of interest.

550

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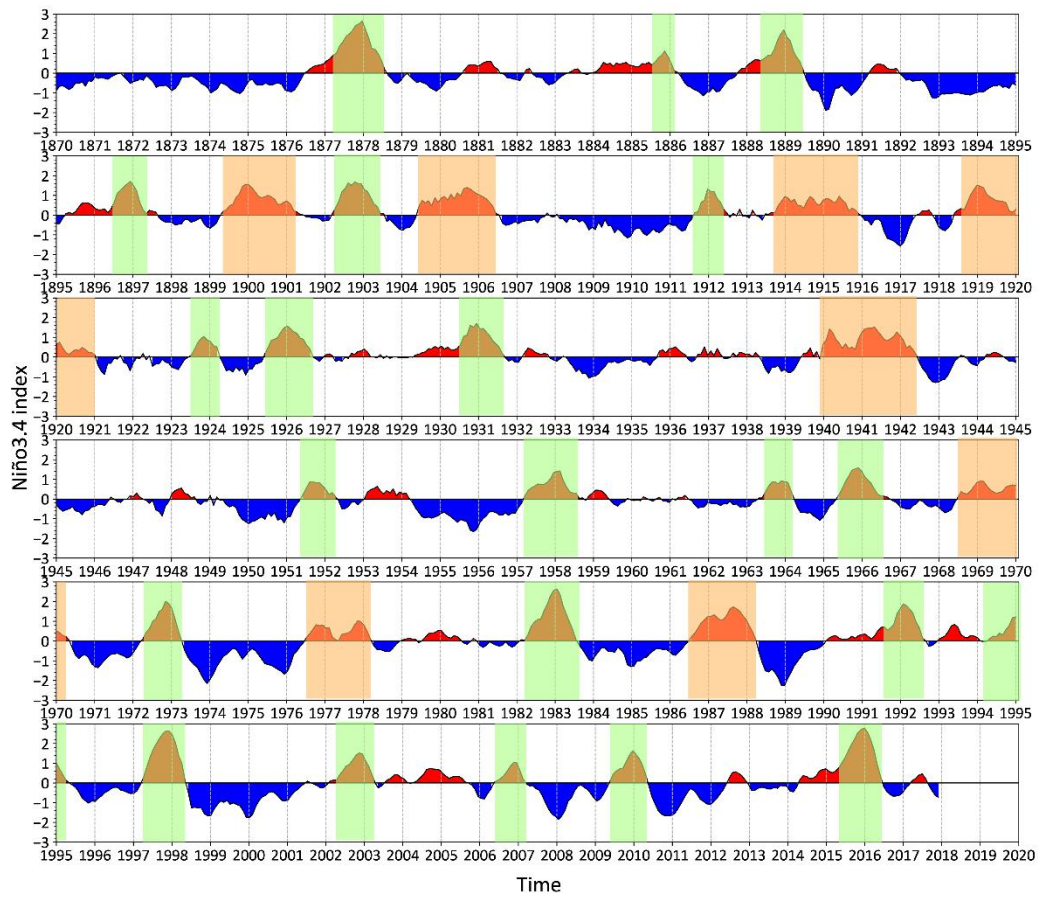
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796 **Table 1.** Experimental design.

Experiments	Model Configuration
CLIM	Climatological SST
SD	Climatological SST + $\Delta\text{SST}_{\text{SD El Niño}}$
LD	Climatological SST + $\Delta\text{SST}_{\text{LD El Niño}}$
SD_emis	Same as SD but turn off the emissions from South and Southeast Asia
CLIM_emis	Same as CLIM but turn off the emissions from South and Southeast Asia

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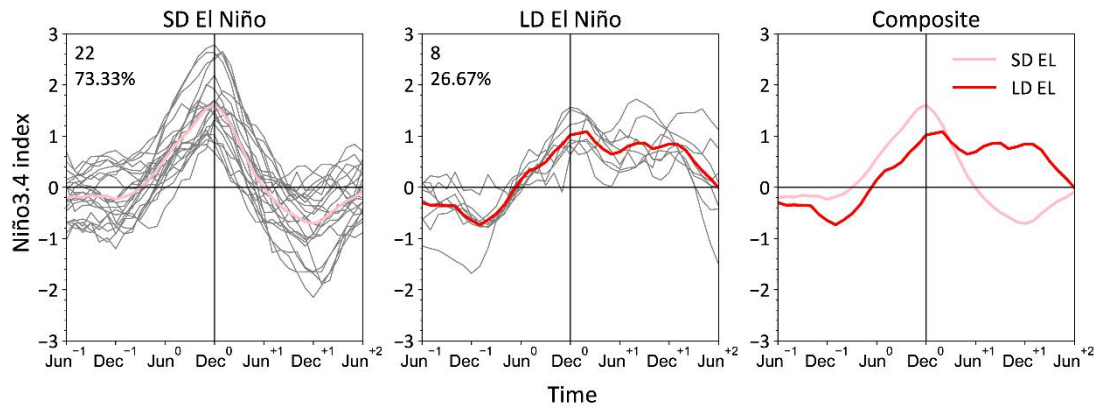
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800 **Figure 1.** Time series of the Niño3.4 index (°C) based on the merged Hadley-NOAA/OI SST

801 dataset for 1870-2017. The time series were detrended and smoothed with a 3-month running

802 average filter. Highlighted slots illustrate the SD (green) and LD (orange) El Niño events.

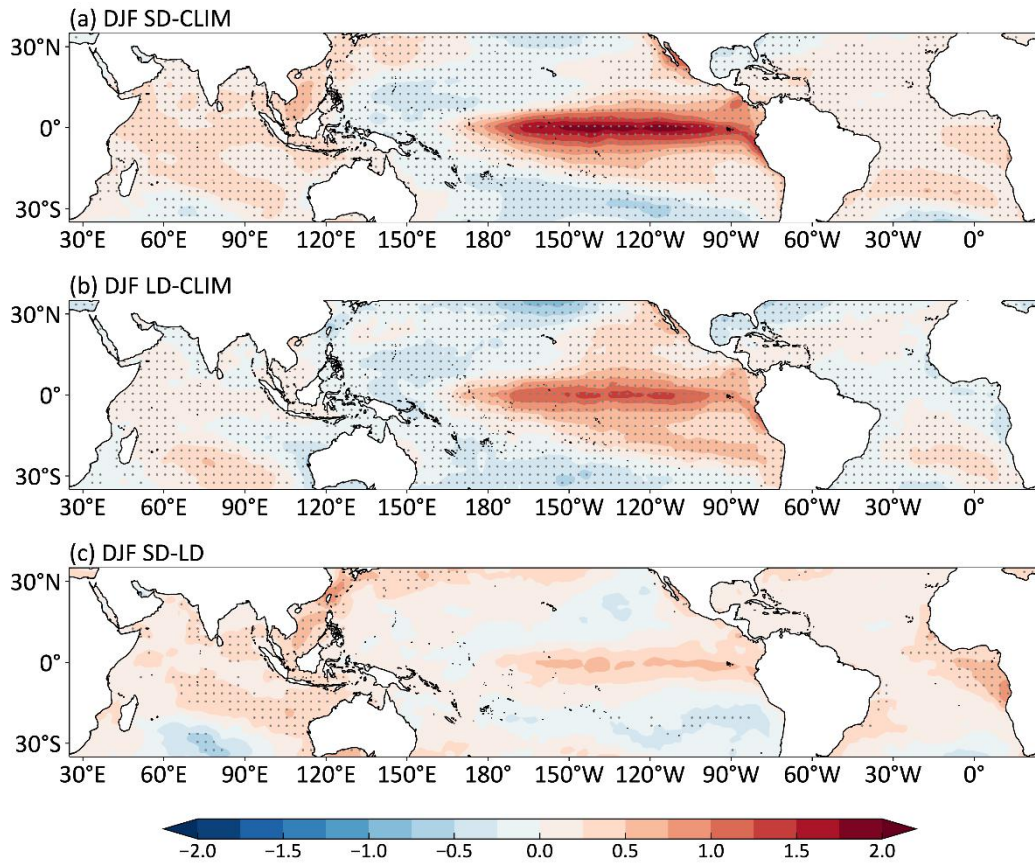
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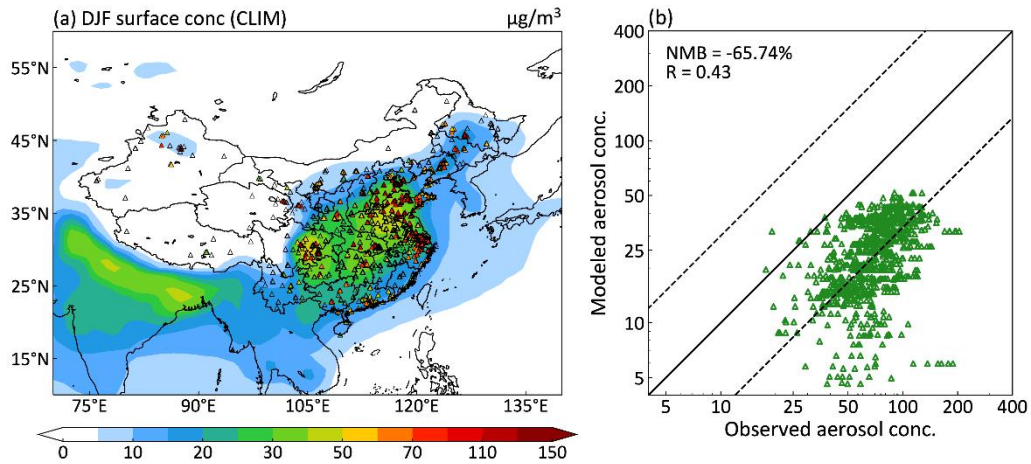
806 **Figure 2.** Time series of the Niño3.4 index (°C) overlaid from Jun⁻¹ to Jun⁺² for (left) SD and
 807 (middle) LD El Niño events during 1870-2017. The individual and composite events are shown by
 808 thin gray and bold red curves, respectively. The total number and percentage of events are shown
 809 at the upper left corner of each panel. A comparison of the composite time series of Niño3.4 index
 810 for SD and LD events is shown in the right panel.



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Figure 3. Composite differences in DJF mean SST (°C) between SD (a) / LD (b) El Niño events and climatological mean over 1870-2017 and between SD and LD (c) El Niño events. Differences that are statistically significant at 95% from a two-tailed T-test are stippled.

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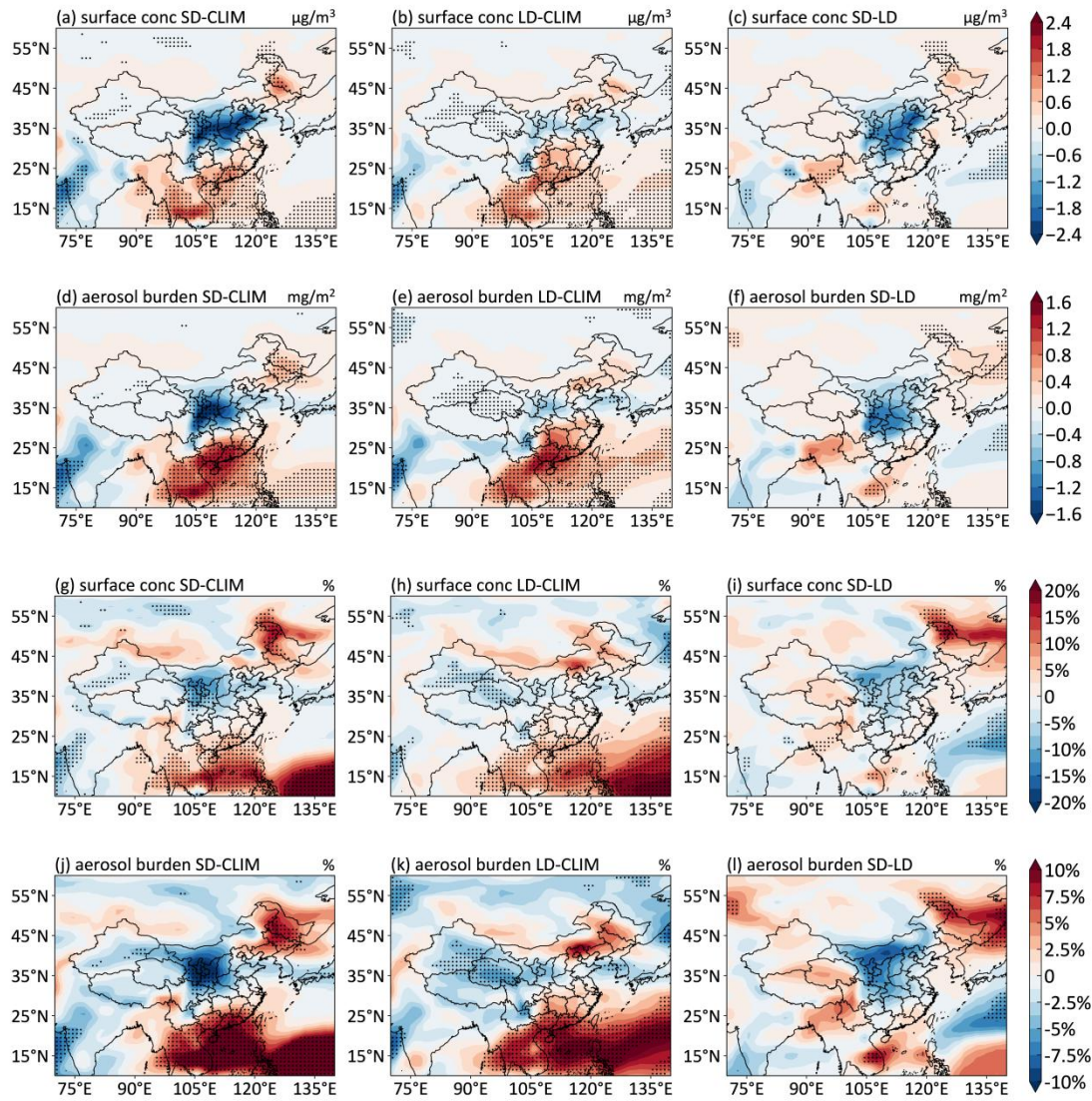


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820 **Figure 4.** Spatial distributions (a) and scatter plots (b) of observed and simulated DJF mean
821 near-surface PM_{2.5} concentrations ($\mu\text{g m}^{-3}$) from the CLIM experiment. Solid line represents 1:1
822 ratio and dashed lines mark 1:3 and 3:1 ratios. The observed concentrations are derived from the
823 CNEMC in December 2014-February 2015. The normalized mean deviation (NMB) and the
824 correlation coefficient (R) between observations and simulation are shown in the upper left corner
825 of the right panel. $\text{NMB} = 100\% \times \sum (M_i - O_i) / \sum O_i$, where M_i and O_i are the simulated and
826 observed values at the site i , respectively.

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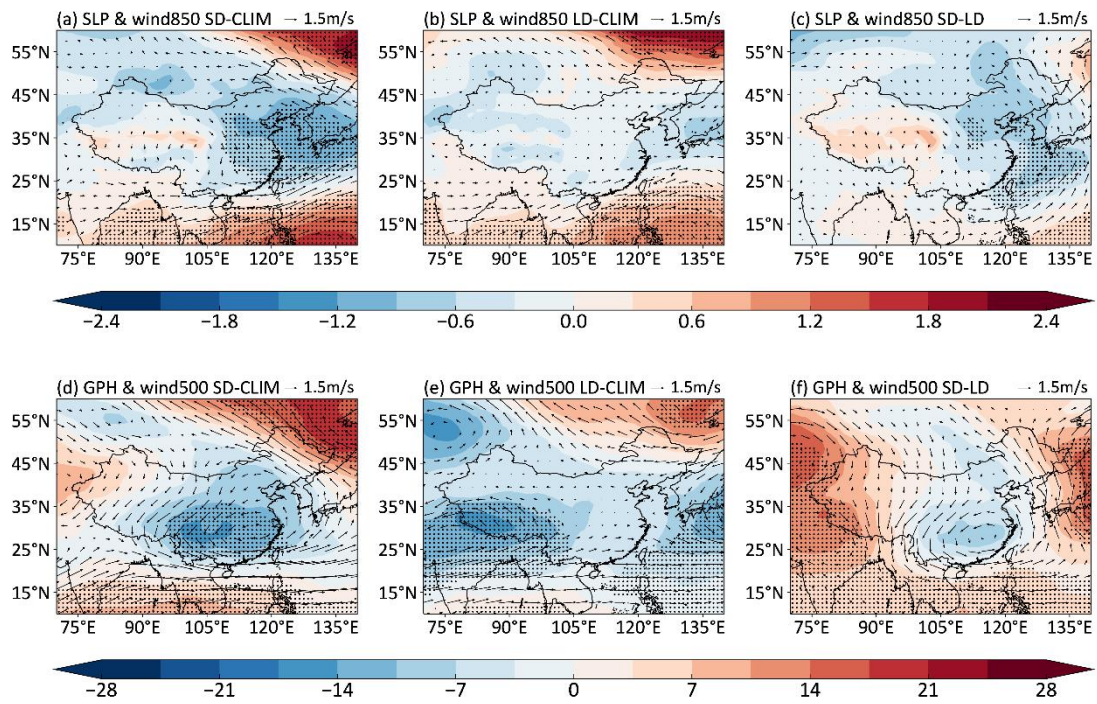


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830 **Figure 5.** The absolute (a-f) and relative (h-l) composite differences in DJF mean
 831 near-surface PM_{2.5} concentrations ($\mu\text{g m}^{-3}$) and aerosol column burdens (mg m^{-2}) between SD and
 832 CLIM (a, d, g, j), LD and CLIM (b, e, h, k), and SD and LD (c, f, i, l). The stippled areas indicate
 833 statistical significance with 90% confidence from a two-tailed T-test.

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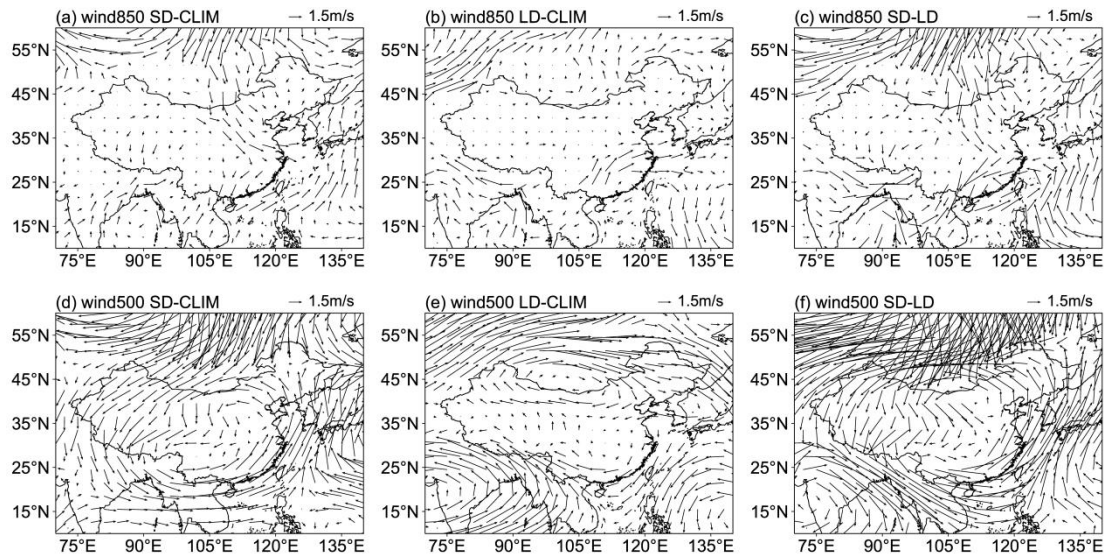


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838 **Figure 6.** Composite differences in DJF mean sea level pressure (SLP, shaded; units: hPa) and
 839 wind at 850 hPa (WIND850, vector; units: m s^{-1}) (top panels) and geopotential height at 500 hPa
 840 (GPH500, shaded; units: m) and wind at 500 hPa (WIND500, vector; units: m s^{-1}) (bottom panels)
 841 between SD and CLIM (a, d), LD and CLIM (b, e), and SD and LD (c, f). The stippled areas
 842 indicate statistical significance with 90% confidence from a two-tailed T-test.

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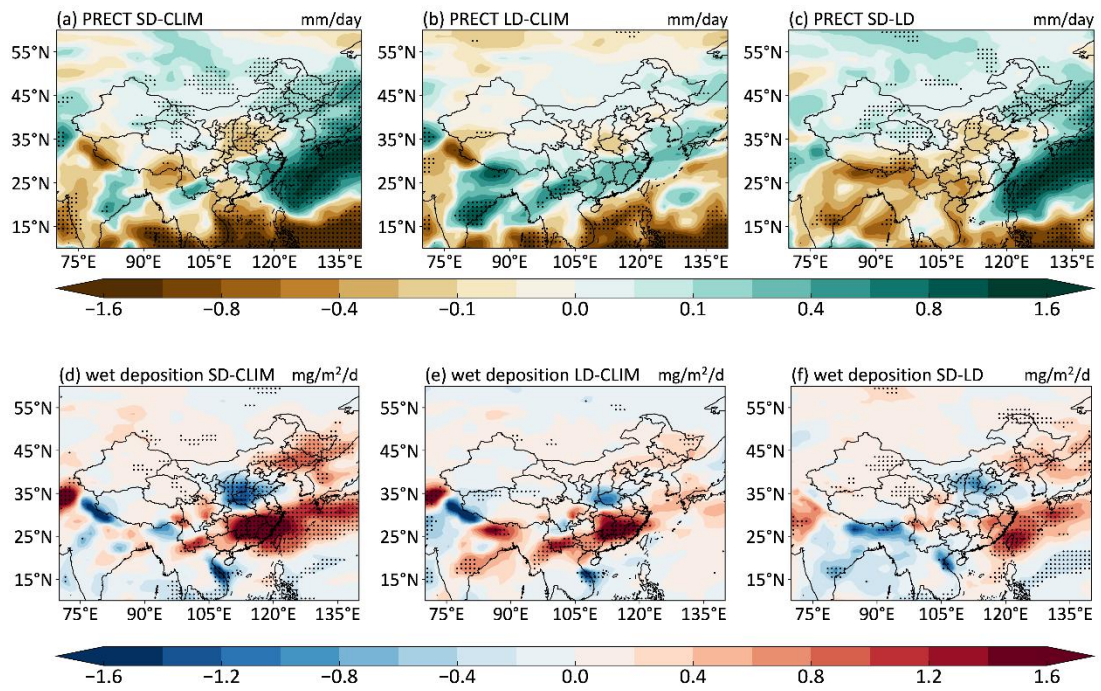


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845 **Figure 7.** Composite differences in DJF mean winds at 850 hPa (m s^{-1}) (top panels) and 500 hPa
 846 (m s^{-1}) (bottom panels) between 2015/2016 SD El Niño and climatological mean (1950-2017) (a,
 847 d), 1986/1987 LD El Niño and climatological mean (b, e), and 2015/2016 SD El Niño and
 848 1986/1987 LD El Niño (c, f) from the EAR5 reanalysis data. The data were detrended over
 849 1950-2017.

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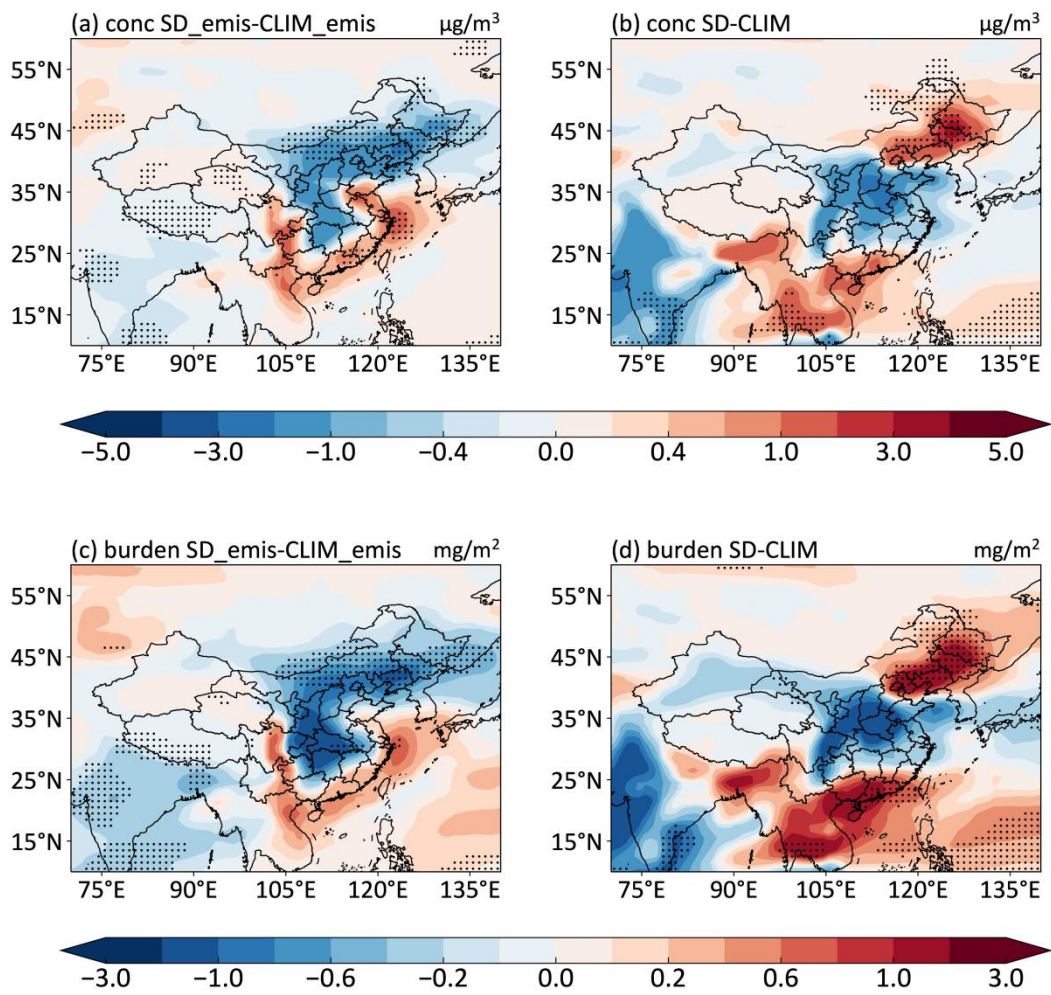


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854 **Figure 8.** Composite differences in DJF mean precipitation rate (top panels; units: mm day⁻¹) and
855 wet deposition of PM_{2.5} (bottom panels; units: mg m⁻² d⁻¹) between SD and CLIM (a, d), LD and
856 CLIM (b, e), and SD and LD (c, f). The stippled areas indicate statistical significance with 90%
857 confidence from a two-tailed T-test.

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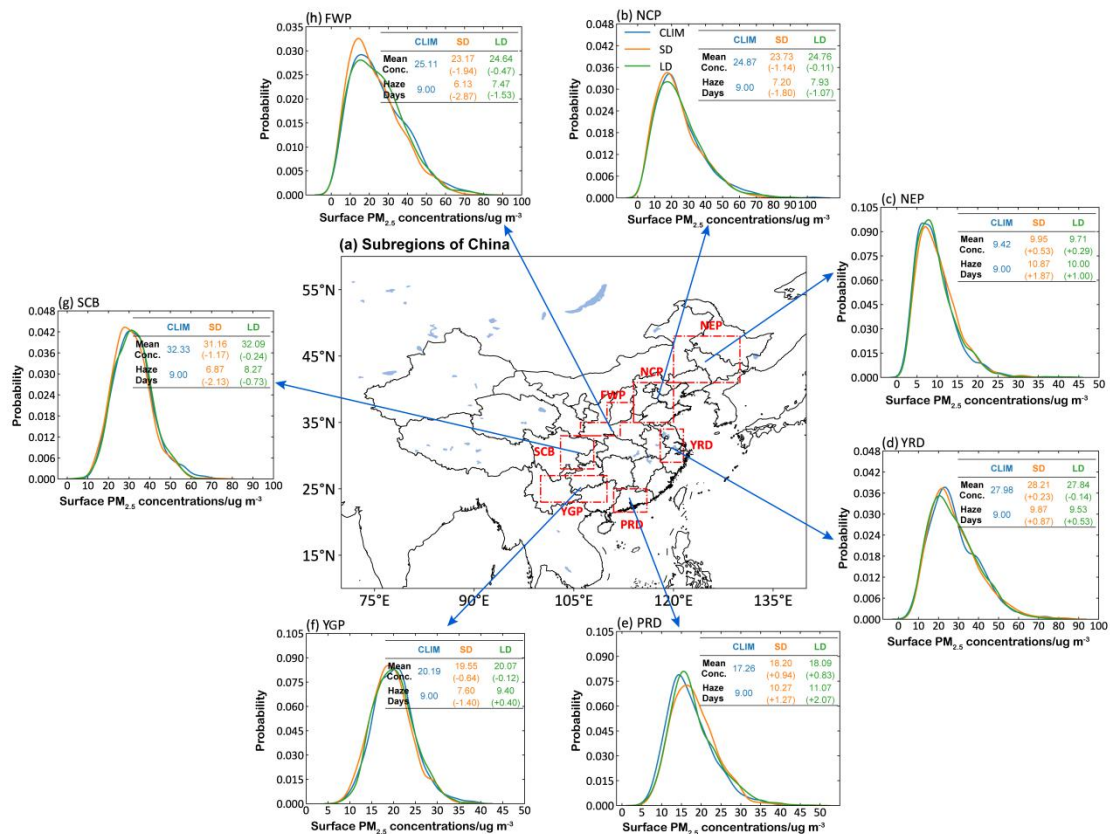
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862 **Figure 9.** Composite differences in DJF mean near-surface PM_{2.5} concentration ($\mu\text{g m}^{-3}$) and
 863 aerosol column burden (mg m^{-2}) between SD_emis and CLIM_emis (a, c) SD and CLIM (b, d).

864 The stippled areas indicate statistical significance with 90% confidence from a two-tailed T-test.

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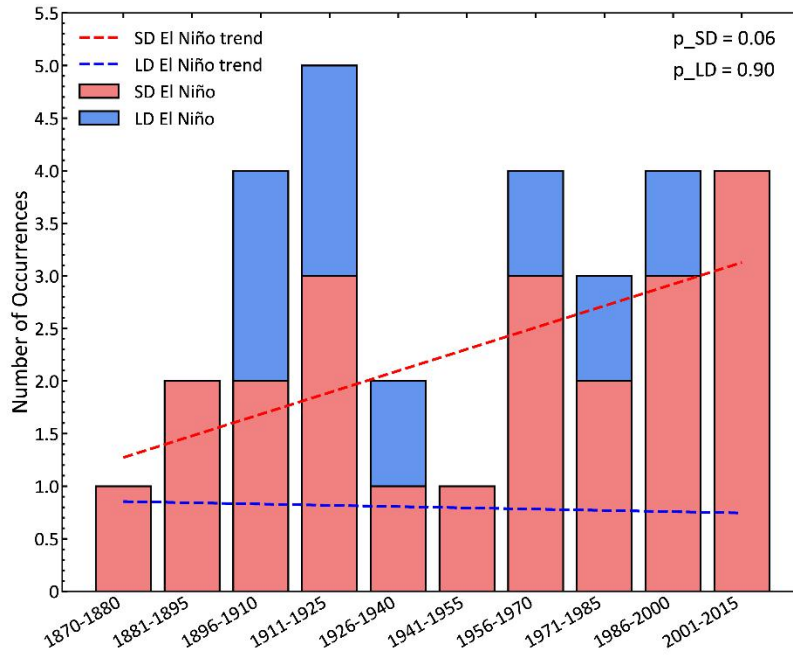


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869 **Figure 10.** (a) Subregions of China defined in this study, including the North China Plain (NCP,
 870 35–41°N, 114–120°E), the Sichuan Basin (SCB, 28–33°N, 103–108°E), the Yangtze River Delta
 871 (YRD, 29–34°N, 118–121.5°E), the Pearl River Delta (PRD, 21.5–25°N, 111–116°E), the
 872 Northeast Plain (NEP, 41–48°N, 120–130°E), the Yunnan-Guizhou Plateau (YGP, 23–27°N,
 873 100–110°E), and the Fenwei Plain (FWP, 33–35°N, 106–112°E and 35–38°N, 110–114°E). (b-h)
 874 Probability density distributions of daily near-surface PM_{2.5} concentrations ($\mu\text{g m}^{-3}$) in DJF over
 875 various subregions of China. The seasonal mean aerosol concentrations (unit: $\mu\text{g m}^{-3}$) and number
 876 of haze days (unit: day) in December-January-February (DJF) over various regions of China from
 877 CLIM, SD and LD experiments are shown in the corresponding table. Haze days are defined as
 878 days with daily average near-surface PM_{2.5} concentrations above the 90th percentile in each region.
 879 The values in brackets represent the anomalies in SD and LD relative to CLIM.

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883 **Figure 11.** Stacked histograms of the number of SD and LD El Niño events per 15 years (except
 884 1870-1880 for 10 years) during 1870-2015. The red and blue dashed lines indicate linear trends in
 885 the number of SD and LD El Niño events, respectively. Their p-values are shown in the upper
 886 right corner of the figure, which indicate the increasing trend of SD at a two-tailed T-test
 887 confidence level of 94% for 1870-2015 (87% for 1880-2015 and 92% for 1940-2015) statistical
 888 significance.

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