1 **Distinguishing the impacts of natural and anthropogenic aerosols on global gross**

2 **primary productivity through diffuse fertilization effect**

4 Hao Zhou [1,2], Xu Yue [3*], Yadong Lei [4], Chenguang Tian[1,2], Jun Zhu [3], Yimian Ma[1,2],

5 Yang Cao [1,2], Xixi Yin [3], Zhiding Zhang [3]

7 [1] Climate Change Research Center, Institute of Atmospheric Physics (IAP), Chinese

8 Academy of Sciences (CAS), Beijing 100029, China

9 [2] University of Chinese Academy of Sciences, Beijing, China

10 [3] Jiangsu Key Laboratory of Atmospheric Environment Monitoring and Pollution

11 Control, Collaborative Innovation Center of Atmospheric Environment and Equipment

12 Technology, School of Environmental Science and Engineering, Nanjing University of

13 Information Science & Technology (NUIST), Nanjing, 210044, China

14 [4] State Key Laboratory of Severe Weather & Key Laboratory of Atmospheric Chemistry

15 of CMA, Chinese Academy of Meteorological Sciences, Beijing, 100081, China

16 *Correspondence to*: Xu Yue (yuexu@nuist.edu.cn)

18 **Abstract**

19 Aerosols can enhance ecosystem productivity by increasing diffuse radiation. Such

20 diffuse fertilization effects (DFEs) vary among different aerosol compositions and sky

21 conditions. Here, we apply a suite of chemical, radiation, and vegetation models in

22 combination with ground- and satellite-based measurements to assess the impacts of

23 natural and anthropogenic aerosol species on gross primary productivity (GPP) through

24 DFE during 2001-2014. Globally, aerosols increase GPP by 8.9 Pg C yr^{-1} at clear skies

25 but only 0.95 Pg C yr^{-1} at all skies. Anthropogenic aerosols account for 41% of the total

26 GPP enhancement though they contribute only 25% to the increment of diffuse

27 radiation. Sulfate/nitrate aerosols from anthropogenic sources make dominant

28 contributions of 33% (36%) to aerosol DFE at all (clear) skies, followed by the ratio of

29 18% (22%) by organic carbon aerosols from natural sources. In contrast to other species,

30 black carbon aerosols decrease global GPP by 0.28 (0.12) Pg C yr^{-1} at all (clear) skies.

31 Long-term simulations show that aerosol DFE is increasing 2.9% yr^{-1} at all skies mainly

32 because of a downward trend in cloud amount. This study suggests that the impacts of

33 aerosols and cloud should be considered in projecting future changes of ecosystem

34 productivity under varied emission scenarios.

36 Keywords: Diffuse fertilization effect, gross primary productivity, anthropogenic

37 aerosols, natural aerosols, YIBs model

38

1 Introduction

39

40   Diffuse light enhances plant photosynthesis more efficiently than direct light (Gu et

41   al., 2002;Alton et al., 2007;Mercado et al., 2009;Jing et al., 2010;Cirino et al.,

42   2014;Zhou et al., 2021b;Zhou et al., 2021c). The cause for such difference is that diffuse

43   light can penetrate into the deep canopy and enhance light use efficiency

44   (LUE=GPP/PAR, gross primary production per photosynthetically active radiation) of

45   shaded leaves (Roderick et al., 2001;Gu et al., 2003;Rap et al., 2015). However, direct

46   light is absorbed only by sunlit leaves and much of it is wasted because these leaves are

47   usually at the light saturated conditions (Gu et al., 2002;He et al., 2013). As a result,

48   increasing the diffuse radiation can help promote canopy photosynthesis through the

49   diffuse fertilization effect (DFE).

50   Atmospheric aerosols can alter the quality of sunlight reaching Earth's surface by

51   absorbing and scattering solar insolation (Zhou et al., 2021a). The aerosol-induced

52   radiative impacts on terrestrial ecosystem productivity have been investigated in both

53   observational and modeling studies (Table 1). Observations found unexpected decline

54   of atmospheric carbon dioxide in 1990s, which was attributed to the increase of

55   vegetation carbon uptake owing to the massive eruption of Mt. Pinatubo in 1991

56   (Roderick et al., 2001). Sulfate aerosols from volcanic eruption almost doubled diffuse

57   radiation at the clear sky, leading to the enhancement of plant productivity by 23% at

58   Harvard forests in 1992 (Gu et al., 2003). With the development of ground-based

59   instruments and satellite remote sensing, more observational data have been applied to

60   detect the aerosol DFE. Strada et al. (2015) estimated aerosol DFE on plant productivity

61   using aerosol optical depth (AOD) from satellite cloudless observations at 10 flux sites,

62   and found that aerosols enhance GPP by 13% in midday hours under high AOD

63   conditions (>0.4) for deciduous and mixed forests. Similarly, Ezhova et al. (2018) found

64   that aerosols increase clear-day diffuse fraction from 0.11 to 0.27 at five remote sites in

65   Eurasia, leading to the enhancement of site-level GPP by 6-14%.

66   In contrast to the large benefits at clear days, the aerosol DFE is limited at cloudy

67   days. Kanniah et al. (2013) explored cloud direct radiative effects on canopy

68   productivity using observed carbon fluxes and radiation in tropical savannas, and found

69   that thick cloud masked aerosol DFE and reduced GPP by 26%. Cirino et al. (2014)

70   also found that aerosol DFE cannot increase plant photosynthesis under cloudy

71   conditions. These studies indicated that aerosol DFE is subject to sky conditions and

72   aerosol loading, because the potential benefits from DFE can be offset or even reversed

73   by simultaneous reductions in direct radiation caused by thick cloud or high aerosol

74   loading (Alton, 2008;Cirino et al., 2014;Yue and Unger, 2017;Zhou et al., 2021a).

75   Although observational studies directly estimate site-level aerosol DFE, they are not

76   able to reveal regional or global aerosol DFE due to the limited spatiotemporal coverage.

77   On the global scale, studies using varied models showed that aerosol DFE enhances

78   global GPP by 4.9 Pg C yr^{-1} (Chen and Zhuang, 2014), 1-2% (Strada and Unger, 2016)

79   and 1.0±0.2 Pg C yr^{-1} (Yue and Unger, 2018) at different periods. Rap et al. (2018)

80   specifically explored DFE from biogenic aerosols and found that biogenic aerosols

81   enhance global NPP by 1.23 Pg C yr^{-1}. Regionally, Matsui et al. (2008) applied a land

82   surface model and estimated that aerosol DFE decreased net primary production (NPP)

83   by 0.09% in 2000 but increased NPP by 0.5% in 2001 over eastern U.S., because the

84   cloud optical depth was about half in 2001 relative to 2000. At the same region, Keppel-

85   Aleks and Washenfelder (2016) estimated sulfate aerosol DFE using Community Earth

86   System Model and found that the reductions of sulfate aerosols by 3.0±0.6% yr^{-1} led to

87   reductions of 0.6% yr^{-1} in diffuse radiation and 0.07% yr^{-1} in regional GPP during 1995-

88   2013. In Amazon, fire aerosols are estimated to play varied DFEs among different

89   studies (Rap et al., 2015;Moreira et al., 2017;Yue and Unger, 2018;Malavelle et al.,

90   2019). For example, Rap et al. (2015) found that fire aerosols enhance NPP by 1.4-2.8%

91   while Moreira et al. (2017) estimated that fire aerosols enhance GPP by 27%. Such

92   differences are mainly attributed to the high aerosol loading in Moreira et al. (2017) for

93   September 2010, but much lower loading in Rap et al. (2015) for the 10-year (1998-

94   2007) averages. Although these studies assessed the DFE of total aerosols or the

95   specific species (e.g., sulfate, fire, or biogenic), the individual DFEs of natural and

96   anthropogenic aerosols on global terrestrial productivity remain unclear.

97   In this study, we explore the impacts of natural and anthropogenic aerosol DFE on

98   global GPP during 2001-2014 using both multi-source observations and a series of well-

99   validated models. A chemical transport model (CTM) is used to predict changes of

100   natural and anthropogenic aerosol concentrations. A radiative transfer model is applied

101   to calculate the perturbations in direct and diffuse PAR caused by aerosols. A global

102   dynamic vegetation model is used to quantify changes of global GPP caused by aerosol

103   DFE. The main objectives are (1) to distinguish the DFEs of natural and anthropogenic

104   aerosols on global GPP and (2) to explore the different characteristics of aerosol DFEs

105   for varied species.

106

107   **2 Methods**

108   **2.1 Chemical transport model**

109   The Goddard Earth Observing System coupled with Chemistry (GEOS-Chem,

110   http://geos-chem.org) is a three-dimensional (3-D) CTM for simulating atmospheric

111   compositions and air quality (Bey et al., 2001). Global anthropogenic emissions during

112   2001-2014 are from the Community Emissions Data System (CEDS) inventory

113   (http://www.globalchange.umd.edu/ceds/). The CEDS inventory has been used as

114   anthropogenic emissions in the Coupled Model Intercomparison Project Phase 6

115   (CMIP6), and this emission database relies on existing energy consumption datasets

116   and regional or country-specific inventories to produce trends over recent decades

117   (Hoesly et al., 2018). The specific emission species include aerosols (black carbon,

118   organic carbon), aerosol precursors and reactive compounds (SO_2, NO_x, NH_3, CH_4, CO,

119   and non-methane volatile organic compounds (VOCs)) (Supplementary Table 1). To

120   estimate modeling uncertainties due to emission inventories, the Emissions Database

121   for Global Atmospheric Research (EDGAR) inventory vision 4.3.1

122   (https://edgar.jrc.ec.europa.eu/) during 2001-2010 is also used as alternative

123   anthropogenic emissions for GEOS-Chem model. For natural emissions, the Global

124   Fire Emission Database (GFED) version 4 inventory is used to represent emissions

125   from open fires (http://www.globalfiredata.org/). Biogenic VOC emissions are

126   calculated using the Model of Emissions of Gases and Aerosols from Nature (MEGAN

127   v2.1) (Guenther et al., 2012). Natural emissions of sea salt (Jaeglé et al., 2011), dimethyl

128   sulfate (Breider et al., 2017), volcanic SO_2 (Fisher et al., 2011) and NH_3 are from the

129   Global Emissions InitiAtive (GEIA, http://www.geiacenter.org/). In this study, GEOS-

130   Chem version 12.0.0 is used to simulate concentrations of natural and anthropogenic

131   aerosols at a horizontal resolution of 4˚×5˚ and 47 vertical layers. The CTM is driven

132   with assimilated meteorology from the Modern-Era Retrospective analysis for

133   Research and Applications, version 2 (MERRA2).

134

135   **2.2 Radiative transfer model**

136   The Column Radiation Model (CRM) is the standalone version of the radiative

137   transfer module used by the NCAR Community Climate Model

138   (http://www.cesm.ucar.edu/models/). In this model, aerosol direct radiative effects

139   including absorbing and scattering processes are calculated at 20 vertical layers from

140   surface to 0.5 hPa at hourly intervals (Yue and Unger, 2017). The CRM utilizes aerosol

141   profiles of all species simulated by GEOS-Chem, including sulfate, nitrate, black

142   carbon (BC), organic carbon (OC), dust (clay and silt) and sea salt (coarse and

143   accumulation modes). Aerosol optical parameters (e.g. single scattering albedo,

144   extinction coefficients, and asymmetric parameters) are adopted from Yue and Liao

145   (2012) for sea salt, Yue et al. (2010) for mineral dust, and the RegCM4 model for other

146   species (Giorgi et al., 2012). In this study, the CRM is used to simulate aerosol-induced

147   perturbations in surface radiative fluxes including diffuse and direct PAR. The model

148   is driven with hourly 1˚×1˚ meteorology from MERRA-2 reanalyses, and 3-hourly

149 cloud cover and liquid water path from CERES SYN1deg (http://ceres.larc.nasa.gov).

150

151 **2.3 Dynamic vegetation model**

152 The Yale Interactive terrestrial Biosphere (YIBs) model is a process-based vegetation

153 model that dynamically simulates tree growth and leaf area changes (Yue and Unger,

154 2015). The model uses the well-established leaf photosynthesis (Farquhar et al., 1980)

155 and stomatal conductance schemes (Ball et al., 1987). The canopy is divided into sunlit

156 and shaded portions to separate photosynthetic responses to diffuse and direct light

157 (Spitters et al., 1986). We distinguish light absorption between sunlit (receiving both

158 diffuse and direct light) and shaded leaves (receiving only diffuse light), and derive

159 canopy photosynthesis as the sum of that from sunlit and shaded leaves:

160 $A_{total} = A_{sunlit} \times F_{sunlit} + A_{shaded} \times (1 - F_{sunlit})$ (1)

161 where A_{sunlit} and A_{shaded} are the photosynthesis of sunlit and shaded leaves,

162 respectively. The fraction of sunlit leaf area F_{sunlit} is calculated as:

163 $F_{sunlit} = e^{-kL}$ (2)

164 Here, L is leaf area index (LAI) at one canopy layer and k is extinction coefficient

165 defined as $0.5/cos\alpha$ (solar zenith α).

166 Simulated GPP by YIBs model were validated using ground-based observations at

167 145 sites and yielded an average correlation coefficient of 0.76 for all sites (Yue and

168 Unger, 2015). The simulated global GPP also shows reasonable spatiotemporal

169 variations compared with satellite retrievals (Yue et al., 2015). Recently, the model

170 joined the multi-model ensemble project of TRENDY to provide the estimates of global

171   carbon budget (Friedlingstein et al., 2020). In this study, the YIBs is used to isolate

172   impacts of aerosol-induced PAR changes on GPP on the global scale. The model is

173   driven with 1˚×1˚ meteorological forcing from MERRA-2 reanalyses and PAR (both

174   diffuse and direct) simulated by CRM model.

175

176   **2.4 Model simulations**

177   We perform 2 GEOS-Chem runs, as well as 22 CRM and YIBs runs to isolate aerosol

178   direct radiative impacts on GPP at different sky conditions (Supplementary Table 2).

179   The GEOS-Chem runs GC_ALL and GC_NAT are driven with the same meteorology

180   and emissions except that the former includes all source of emissions while the latter

181   excludes anthropogenic emissions. The differences between GC_ALL and GC_NAT

182   represent aerosol concentrations contributed by anthropogenic sources. Both GC_ALL

183   and GC_NAT runs provide 3-D concentrations of different aerosol types including

184   sulfate, nitrate, OC, BC, dust and sea salt. The CRM runs aim to calculate aerosol-

185   induced PAR changes using aerosol profiles simulated by GEOS-Chem. These runs can

186   be divided into two groups, with CLD runs (all-sky conditions) forced with observed

187   cloud profiles while CLR runs (clear-sky conditions) forced without any cloud coverage.

188   CRM_ALL and CRM_NAT are driven with aerosol profiles of all species from

189   GC_ALL and GC_NAT, respectively. The impacts of individual aerosol species on PAR

190   are isolated with individual aerosol profiles from either GC_ALL or GC_NAT. For

191   example, OC from GC_ALL and cloud amounts from CERES SYN1deg are used to

192   drive CRM (CRM_ALL_OCCLD) so as to isolate the impacts of OC aerosols on PAR

193   at all-sky conditions. For each of CRM runs, the predicted diffuse and direct PAR are

194   used as input for YIBs model to simulate GPP changes caused by aerosol DFEs.

195

196   **2.5 Observations for model evaluations**

197   We use site-level measurements of carbon fluxes from the FLUXNET2015 product

198   (http://fluxnet.fluxdata.org/) to validate model GPP and its responses to diffuse/direct

199   radiation. We select 10 sites providing simultaneous observations of diffuse radiation

200   and GPP at half-hourly time interval for at least 8 years. On the global scale, observed

201   AOD is retrieved from the Moderate Resolution Imaging Spectroradiometer (MODIS,

202   https://modis.gsfc.nasa.gov) and GPP is derived using global OCO-2-based SIF product

203   (Li and Xiao, 2019). The all-sky and clear-sky shortwave radiation are adopted from

204   CERES SYN1deg (http://ceres.larc.nasa.gov) to validate the CRM radiative transfer

205   model.

206

207   **3 Results**

208   **3.1 Model evaluations**

209   The YIBs model simulates reasonable spatial pattern of GPP compared to

210   observations (Figure S1) with a high correlation coefficient (R) of 0.88 ($p<0.01$) and a

211   low normalize mean bias (NMB) of -2.3%. Similarly, modeled AOD from GEOS-Chem

212   model reproduces the observed spatial pattern from MODIS product with high R of

213   0.78 (p<0.01), though overestimates the mean AOD by 21.7% in eastern China and

214   37.9% in southern Africa while underestimates AOD by 35.7% in Amazon, 25.2% in

215 Central Africa and 53.4% in southeast Asia, leading to a global NMB of -25.8%.

216 The CRM model driven with aerosol concentrations from GEOS-Chem shows

217 similar patterns of shortwave radiation to the satellite observations (Figure S2). The

218 simulations match observations well with high R of 0.98 and low NMB of 4.1% at all-

219 sky conditions, and show even better performance with R of 1 and NMB of 3.7% at

220 clear skies. Moreover, the CRM model simulates reasonable aerosol direct radiative

221 effects compared to multiple radiative transfer models as shown in Yue and Unger

222 (2018).

223 We then compared the simulated and observed GPP responses to direct (diffuse

224 fraction <0.2) and diffuse radiation (diffuse fraction >0.8) (Figure 1). Observations and

225 simulations show that diffuse light can increase GPP more efficiently than direct

226 radiation as shown by the higher GPP-PAR slopes at diffuse conditions. Similar results

227 were achieved by Mercado et al. (2009) and Yue and Unger (2018) using the same

228 methods. The diffuse fertilization efficiency, percentage changes in GPP per unit diffuse

229 PAR, is estimated to be 0.45-0.7% W^{-1} m^2 for observations and 0.3-0.69% W^{-1} m^2 for

230 simulations. As a result, the YIBs model can reasonably reproduce varied light-response

231 curves so as to isolate GPP responses to direct and diffuse radiation.

232

233 **3.2 Changes of PAR by aerosols**

234 Appearance of aerosols on average decreases total surface PAR by 1.52 W m^{-2} at all

235 skies and 2.73 W m^{-2} at clear skies on the global scale. Under all-sky conditions,

236 aerosols increase diffuse PAR by 1.26 W m^{-2} (Figure 2a) but decrease direct PAR by

237   2.78 W m^{-2} (Figure S3a). These changes are larger in clear-sky conditions with increase

238   of diffuse PAR by 4.98 W m^{-2} (Figure 2d) and decrease of direct PAR by 7.71 W m^{-2}

239   (Figure S3d). The cause of smaller PAR changes at all skies is that cloud tends to

240   weaken aerosol radiative forcing by amplifying absorption and diminishing scattering

241   (Paulot et al., 2018).

242   At all-sky conditions, natural aerosols dominate aerosol-induced PAR changes by

243   increasing diffuse PAR of 0.93 W m^{-2} (Figure 2b) and decreasing direct PAR of 2.05 W

244   m^{-2} (Figure S3b). As a comparison, anthropogenic aerosols induce much smaller

245   changes of diffuse radiation by 0.33 W m^{-2} and direct radiation of -0.72 W m^{-2} (Figures

246   2c and S3c). Natural aerosols mainly influence PAR fluxes in northern Africa owing to

247   large amount of dust aerosols, while anthropogenic aerosols dominate PAR changes in

248   eastern China, India, and eastern U.S due to the large anthropogenic emissions. At clear-

249   sky conditions, natural aerosols increase diffuse PAR by 3.79 W m^{-2} (Figure 2e) and

250   decrease direct PAR by 5.84 W m^{-2} (Figure S3e), and anthropogenic aerosols on average

251   increase diffuse PAR by 1.19 W m^{-2} and decrease direct PAR by 1.88 W m^{-2}.

252   We further explore the contributions of individual aerosol species to the changes of

253   diffuse and direct PAR at all skies (Figures S4 and S5). On the global scale, sulfate and

254   nitrate aerosols increase diffuse PAR by 0.57 W m^{-2}, accounting for 51% of aerosol-

255   induced diffuse PAR changes. Meanwhile, diffuse PAR is increased 0.05 W m^{-2}, 0.37

256   W m^{-2} and 0.25 W m^{-2} by the scattering effects of OC, dust, and sea salt aerosols.

257   However, BC aerosols reduce diffuse PAR by 0.06 W m^{-2} due to the strong absorption.

258   The changes of direct PAR caused by all aerosol species are negative, especially that

259   by sulfate and nitrate (-0.97 W m^{-2}), dust (-0.86 W m^{-2}), and sea salt (-0.5 W m^{-2}).

260   Generally, natural aerosols dominate changes of diffuse and direct PAR owing to the

261   large contributions from dust and sea salt aerosols. However, sulfate, nitrate, and BC

262   aerosols from anthropogenic sources dominate the changes of diffuse and direct PAR

263   over eastern China, Indian and eastern U.S.

264

265   **3.3 DFE by natural and anthropogenic aerosols**

266   We quantify the percentage changes of global GPP caused by aerosol DFE. On the

267   global scale, aerosol DFE increases GPP by 0.65 % (0.95 ± 0.13 Pg C yr^{-1}) at all skies

268   (Figure 3a). Relatively high enhancements (>2%) in GPP are found over middle

269   latitudes (20-50˚N) following the changes of diffuse PAR (Figure 2a). The DFE of

270   natural aerosols increases global GPP by 0.38% (0.56 ± 0.1 Pg C yr^{-1}), mainly over

271   Middle East and northern Africa due to dust aerosols (Figure S6a and S7g). The DFE

272   of anthropogenic aerosols increases global GPP up to 0.27 % (0.39 ± 0.04 Pg C yr^{-1}),

273   especially over populous regions including northeast China, Middle East and

274   contiguous U.S (Figure S6b).

275   At clear skies, aerosol DFE increases global GPP up to 7.8% (8.91 ± 0.26 Pg C yr^{-1})

276   (Figure 3c), which is around 9.5 times of that at all skies (Figure 3a). In most regions,

277   aerosol DFE increases GPP by more than 4%, with the maximum enhancement of 22.7%

278   in East Asia. The DFE of natural aerosols enhances global GPP by 4.6%, with large

279   changes over Amazon, center Africa, boreal Asia, and North America (Figure S6c).

280   Meanwhile, anthropogenic aerosols increase GPP by 3.2%, mainly located at eastern

281   U.S, Europe, boreal Asia, India and East Asia (Figure S6d).

282   We further quantify the contributions of anthropogenic aerosols to the total aerosol

283   DFE. Although cloud masks aerosol DFE and significantly reduces GPP enhancement,

284   the contributions of anthropogenic aerosols remain similar between all-sky (Figure 3b)

285   and clear-sky (Figure 3d) conditions. Relatively high contributions (>50%) are located

286   at low-mid latitudes including North America, Europe, and eastern China. Low

287   contributions (<50%) are found at other regions such as Africa, South America, and

288   Australia. On the global scale, anthropogenic aerosols on average contribute to 41% of

289   the total aerosol DFE at all-sky conditions (Figure S6a and S6b). Anthropogenic

290   aerosols dominate DFE over 30.5% of land grids at all skies, but only 19.5% at clear

291   skies (Figure 3b and 3d). The most significant differences are located at boreal Europe

292   where the anthropogenic aerosols make dominant contributions to DFE at clear skies

293   while the natural species dominate at all skies.

294

295   **3.4 DFE by individual aerosol species**

296   We isolate the DFE of individual aerosol species on global GPP (Figures S7 and

297   S8). At all-sky conditions, sulfate and nitrate aerosols averagely increase GPP by 0.79

298   Pg C yr^{-1}, to which anthropogenic sources contribute 0.58 Pg C yr^{-1} (Figure S7f). OC

299   aerosols increase global GPP by 0.47 Pg C yr^{-1}, to which natural sources contribute 0.32

300   Pg C yr^{-1} (Figure S7c). As the dominant natural species, dust and sea salt are generated

301   from non-vegetated areas. They can increase GPP of downwind land regions by 0.17

302   Pg C yr^{-1} (Figure S7g) and 0.06 Pg C yr^{-1} (Figure S7h), respectively. Different from the

303   above species, BC aerosols lead to negative impacts on GPP up to -0.28 Pg C yr^{-1}

304   globally due to the strong absorbing radiative effects. Regionally, such negative effects

305   are prominent over center Africa from biomass burning (Figure S7a) and eastern China

306   from anthropogenic emissions (Figure S7b).

307   At clear-sky conditions, scattering aerosols show larger DFE compared to the all-

308   sky conditions. sulfate and nitrate aerosols increase global GPP by 5.18 Pg C yr^{-1}, which

309   is 6.6 times of that at all skies. The DFE of OC aerosols also largely increase to 2.89 Pg

310   C yr^{-1}, in which 2.21 Pg C yr^{-1} is from natural sources. Dust and sea salt aerosols lead

311   to positive impacts on global GPP by 1.32 Pg C yr^{-1} and 0.48 Pg C yr^{-1}, respectively. In

312   contrast, BC aerosols reduce global GPP by 0.12 Pg C yr^{-1}, much weaker than the

313   magnitude of 0.28 Pg C yr^{-1} at all skies. Such change manly follows the larger diffuse

314   absorption by BC aerosols at all skies (0.06 W m^{-2}) than that at clear skies (0.02 W m^{-2}

315   ).

316   We then identify the aerosol species making the dominant contributions to the total

317   aerosol DFE (Figure 4). At all-sky conditions, sulfate and nitrate aerosols lead the DFE

318   at 65% grids (Figure 4a) and account for 44.7% of the total absolute GPP changes

319   (Figure 4c). The secondary contribution is from OC aerosols, which account for 26.7%

320   of the total DFE. Dust and sea salt aerosols contribute to the total DFE by 9.5% and

321   3.4%, respectively (Figure 4c). BC aerosols exert negative DFE, the absolute value of

322   which is equivalent to 15.8% of the total DFE. Regionally, sulfate and nitrate aerosols

323   lead DFE in eastern China, India, eastern U.S., and Europe, while dust aerosols

324   dominate DFE at Middle East (Figures 4e and 4f). At clear-sky conditions, the

325   percentage contributions of sulfate and nitrate aerosols to the total DFE further increase

326   to 51.8% on the global scale (Figure 4d). OC, dust, and sea salt aerosols show

327   comparable contributions to DFE as that at all skies. However, the absolute ratios by

328   BC aerosols significantly reduce to 1.2%, because BC-induced DFE is limited while

329   DFE of other species is significantly strengthened at clear skies (Figure S8).

330   We further explore the interannual variations of GPP changes caused by aerosol

331   DFE from natural and anthropogenic sources (Figure 5). At all-sky conditions, aerosol

332   DFE significantly ($p < 0.05$) increases by 2.89% yr^{-1} (24.6 Tg C yr^{-2}) on the global scale

333   (Figure S9a). Such enhancement is mainly located in northeastern China, India, central

334   Africa, and Europe (Figure S10a). Natural aerosols lead to a positive trend of 4.7% yr^{-}

335   ^{1} in the global GPP (22 Tg C yr^{-2}), which is six times of the trend of 0.67% yr^{-1} (2.6 Tg

336   C yr^{-2}) from anthropogenic aerosols (Figure 5a). At clear-sky conditions, aerosol DFE

337   increases by only 0.4% yr^{-1} (Figure S9b), much lower than that at all skies (Figure S9a).

338   Both the DFE trends from natural and anthropogenic aerosols are limited (Figure 5b).

339   The contrast of DFE trends between different sky conditions is related to the changes

340   of cloud amount, which shows a significant reduction trend of 0.38% yr^{-1} in 2001-2014

341   (Figure S9c), especially over Amazon and eastern U.S. (Figure S10d). The reduction of

342   cloud helps increase or maintain aerosol DFE at all-sky conditions (Figure S10c). The

343   trend of all-sky aerosol DFE is mainly contributed by dust aerosols from natural sources,

344   which increases by 4.75% yr^{-1} during 2001-2014 (Figure 5c). The trend of clear-sky

345   aerosol DFE is mainly attributed to sulfate and nitrate aerosols, which increase by 0.44%

346   yr^{-1} during 2001-2014 (Figure 5d).

347

4 Discussion

4.1 Factors influencing aerosol DFE

We quantified the impacts of sky conditions, emission sources, and aerosol species on terrestrial ecosystem productivity through aerosol DFE. In our simulations, aerosols increase global GPP by 8.91 Pg C yr^{-1} at clear skies but only 0.95 Pg C yr^{-1} at all skies. Similarly, Cohan et al. (2002) and Yue and Unger (2017) found aerosol DFE was limited at cloudy skies. Cloud can mask aerosol DFE by modifying both the quantity and quality of aerosol radiative perturbations (Yu et al., 2006). First, cloud weakens the impacts of aerosols on both direct and diffuse radiation (Figure 2 and S3) by reducing the total sunlight available for the extinction by aerosols (Kinne, 2019). Therefore, the smaller changes in diffuse PAR by aerosols at all skies (Figure 2) result in lower DFE than that at clear skies. Second, cloud significantly reduces direct radiation and limits the potential of increasing GPP by diffuse radiation. Observations have shown an optimal diffuse fraction of 0.4-0.6 to enhance GPP for most plant types (Zhou et al., 2021c). A further increase of diffuse fraction above the optimal range will dampen GPP due to the reduced photosynthesis of sunlit leaves. Appearance of cloud has provided an environment with high diffuse fraction that aerosols may have limited benefits or even negative effects for GPP (Yue and Unger, 2017). Such relationship also explains why the decreasing trend of global cloud amount contributes to an increased aerosol DFE (Figure 5a).

Anthropogenic aerosols account for ~25% of the total aerosol-induced

369   enhancement of diffuse radiation (Figure 2), while they contribute 41% to the total

370   aerosol DFE at both all and clear sky conditions (Figure 3). The higher efficiency of

371   anthropogenic aerosols in increasing GPP is partly associated with their geographic

372   distribution. Regionally, anthropogenic aerosols take a leading role in DFE over North

373   America, Europe, India, and eastern China, consistent with the estimations by Strada

374   and Unger (2016). On the other hand, natural aerosols dominate DFE at the tropical

375   regions. Observations have revealed higher optimal diffuse fraction at higher latitudes,

376   where the higher solar zenith angle induces larger fraction of shading leaves (Zhou et

377   al., 2021c). As a result, the same amount of diffuse radiation increased by anthropogenic

378   aerosols results in higher GPP enhancement at the middle latitudes than natural aerosols

379   at low latitudes. Furthermore, a dominant fraction of natural aerosols is contributed by

380   dust and sea salt, which increase diffuse radiation over the barren land or open ocean

381   with little forest coverage (Figure 2). In contrast, most anthropogenic aerosols locate at

382   populous regions covered with dense vegetation. Consequently, the diffuse radiation by

383   anthropogenic aerosols has more interactions with ecosystems than that from natural

384   sources.

385   Different aerosol species induce varied DFEs to global GPP. Sulfate and nitrate

386   dominate the aerosols-induced GPP changes (Figure 4) because their strong scattering

387   effects (Gu et al., 2003) largely increase diffuse radiation (Figure S7 and S8). Keppel-

388   Aleks and Washenfelder (2016) estimated that the regional reductions of sulfate

389   aerosols decreased diffuse radiation by 0.6% yr^{-1} and GPP by 0.07% yr^{-1} in eastern U.S.

390   during 1995-2013. Such negative trends of GPP can also be found over the same region

391   in our clear-sky simulations (Figure S10b). However, the global ΔGPP shows limited

392   trends at clear skies (Figure 5b) because the enhanced SO_2 emissions in China at the

393   same period (Hoesly et al., 2018) increased sulfate loading, promoted local GPP (Figure

394   S10b), and offset the negative ΔGPP in eastern U.S. In our simulations, OC aerosols

395   promote global GPP by 0.47 Pg C yr^{-1}. Such magnitude is much lower than the estimates

396   of 0.76-1.61 Pg C yr^{-1} for the same aerosol species by Rap et al. (2018). The main cause

397   of such discrepancy is related to the predicted aerosol concentrations and radiative

398   effects in two studies (Zhou et al., 2021a). Dust and sea salt aerosols increase regional

399   GPP over arid and coastal regions due to the large local emissions (Yue et al., 2010;Yue

400   and Liao, 2012). At all skies, dust exerts a large DFE over North Africa and Middle

401   East (Figure 3a) because of the low cloud coverage (Figure S11). However, such high

402   GPP ratio shows limited contributions (Figure 4) to global total ΔGPP because of the

403   extremely low baseline GPP in arid regions. Different from above species, BC exerts

404   negative impacts on direct and diffuse PAR owing to strong absorbing properties

405   (Kvalevåg and Myhre, 2007). As a result, BC aerosols always decrease GPP with

406   stronger dampening effects at all skies (Figures 4c and 4d) when the light availability

407   is much smaller than that at clear skies.

408

409   4.2 Uncertainties

410   Our simulations are subject to limitations and uncertainties. First, biases in aerosol

411   profiles may influence the derived aerosol DFE. We used the chemical transport model

412   GEOS-Chem to predict aerosol concentrations and identify contributions from natural

413 and anthropogenic sources. Evaluations showed that GEOS-Chem underestimated

414 global AOD by 25.8%, especially over Amazon, central Africa, and boreal Asia (Figure

415 S1) where natural aerosols dominate. Such bias in part explains why Rap et al. (2018)

416 estimated that biogenic aerosols increased global NPP by 1.23 Pg C yr^{-1} with hotspots

417 over Amazon and central Africa, while our study derived only a moderate enhancement

418 of 0.32 Pg C yr^{-1} by natural OC aerosols. In contrast, simulated AOD is overestimated

419 in eastern China where anthropogenic sources dominate. In this study, CEDS emission

420 inventory is used for anthropogenic emissions. Here, we used another emission

421 database (EDGAR) to assess the uncertainties of DFE from anthropogenic aerosols.

422 The new simulations showed that anthropogenic aerosols increased global GPP by 0.31

423 Pg C yr^{-1} (Figures S12-S13), lower than the value of 0.39 Pg C yr^{-1} predicted with

424 CEDS inventory (Figure 3). The spatial patterns of dominant species for aerosol DFE

425 (Figure S14) remain similar to that from CEDS under both sky conditions.

426   Second, uncertainties in the radiative transfer may cause biases to aerosol DFE.

427 Although the CRM was fully validated with observations (Figure S2), simulated aerosol

428 radiative effects showed large differences compared to other studies. For example, Chen

429 and Zhuang (2014) found that aerosols increased surface diffuse PAR by 5.2 W m^{-2}

430 using another radiative transfer model. In our simulations, we estimated that aerosols

431 increased diffuse PAR by only 1.26 W m^{-2}. As a result, the GPP enhancement by aerosol

432 DFE is 0.95 ± 0.13 Pg C yr^{-1} in our study, much lower than the value of 4.9 Pg C yr^{-1}

433 in Chen and Zhuang (2014) though the latter study also considered aerosol-induced

434 changes in temperature and soil moisture. However, the aerosol radiative effects are

435   likely overestimated in Chen and Zhuang (2014), which predicted total (direct + diffuse)

436   reductions of 21.9 W m^{-2} in surface solar radiation by aerosols; such magnitude is much

437   higher than the multi-model ensemble estimate of -6.3 W m^{-2} at clear skies (Yu et al.,

438   2006). As a comparison, our simulations showed a reduction of 5.8 W m^{-2} in surface

439   shortwave radiation, much closer to the ensemble estimates by Yu et al. (2006).

440   Third, we neglected the climatic responses to aerosol radiative effects. Surface

441   temperature and relative humidity is altered in response to radiative changes caused by

442   aerosols (Jing et al., 2010;Cirino et al., 2014). The increase of relative humidity can

443   increase plant photosynthesis owing to the enhancement of water use efficiency (Lu et

444   al., 2017;Wang et al., 2021), but the impacts of cooling on photosynthesis are dependent

445   on whether local background temperature is over the optimal temperature (Farquhar et

446   al., 1980). Moreover, the changes in cloud from aerosol indirect effects were not

447   considered in this study. Cloud can significantly influence aerosol DFE because of its

448   strong scattering effects (Figure 3). The perturbations in cloud can further influence

449   surface temperature, precipitation, and radiation (Zhu et al., 2019), leading to more

450   complex impacts on terrestrial ecosystem productivity. However, these interactive

451   effects by aerosols need to be resolved using earth system models that implement fully

452   coupled atmospheric chemistry, radiation, land biosphere, and climate feedbacks.

453

454   4.3 Implications

455   Our study reveals that aerosol DFE can enhance global GPP by 0.95 Pg C yr^{-1} at

456   all skies and as high as 8.91 Pg C yr^{-1} at clear skies. The natural and anthropogenic

457   aerosols make comparable contributions globally but with distinct spatial patterns. The

458   DFE, as well as the climatic effects, suggests that aerosols play important roles in

459   mitigating global warming through direct (cooling) and indirect (more carbon

460   assimilation) processes. Although the reductions of aerosols may weaken the DFE, the

461   associated reductions of cloud amount due to reduced aerosol-cloud interactions may

462   induce more benefits to ecosystems. Furthermore, reductions of black carbon aerosols

463   help relieve both climate warming and GPP inhibitions. Our results suggest that aerosol

464   DFE should be considered in projecting future changes in terrestrial ecosystem

465   productivity especially for different emission scenarios.

466

467   **Acknowledgements**

468   This work was jointly supported by the National Key Research and Development

469   Program of China (grant no. 2019YFA0606802) and Jiangsu Science Fund for

470   Distinguished Young Scholars (grant no. BK20200040).

471

472   **Data availability**

473   The simulated GPP and diffuse PAR caused by natural and anthropogenic aerosols on

474   this paper are publicly available via Zenodo (http://doi.org/10.5281/zenodo.5115314).

475

476   **Author contributions**

477   X.Y. conceived the study; X.Y., H.Z. and Y.D.L. designed the research and performed

478   simulations; H.Z. completed data analysis and the first draft; X.Y. reviewed and edited

479   the manuscript; C.G.T, J.Z., Y.M.M., Y.C. X.X.Y and Z.D.Z were responsible for data

480   collection processes.

481

Competing interests

The authors declare no competing interests.

484

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670   **Table 1** Summary of previous studies about aerosol DFE

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Period	Region	Method	Species	Results[a]	References
2000-2001	Eastern United States	Model	Aerosols	Aerosol DFE decreases NPP by 0.71 g C m^{-2} (-0.09%) in 2000 but increases NPP by 5 g C m^{-2} (0.5%) in 2001	Matsui et al. (2008)
1960-1999	Global	Model	Cloud and aerosols	DFE enhances the land carbon sink by approximately one quarter during 1960-1999	Mercado et al. (2009)
2002-2003	Amazon	Flux.obs	Smoke aerosols	The increase of CO_2 uptake under high AOD is due to DFE (80%) and decreased temperature (20%)	Doughty et al. (2010)
2007 July–August	Northwest China	Flux.obs	Cloud and aerosols	Cloud dominates DFE, but aerosols lead to negative carbon uptake	Jing et al. (2010)
2003-2010	Global	Model	Aerosols	Aerosol DFE enhances GPP by 4.9 Pg C yr^{-2}, NPP by 3.8 Pg C yr^{-2}, and NEP by 3.9 Pg C yr^{-2}	Chen and Zhuang (2014)
1999-2009	Amazon	Flux.obs	Cloud and fire aerosols	Low AOD and cloud cover lead to relatively larger photosynthetic efficiency than high aerosol loading and thick cloud	Cirino et al. (2014)
1998-2007	Amazon	Model	Fire aerosols	Fire aerosols enhance diffuse radiation by 3.4-6.8% and NPP by 1.4-2.8%	Rap et al. (2015)
2003-2012	Eastern United States	Flux.obs	Aerosols	High AOD (>0.6) enhances plant productivity for forests, but causes negative effects for croplands and grasslands.	Strada et al. (2015)
2000	Global	Model	Aerosols	Aerosol DFE increases global GPP by 1-2%	Strada and Unger (2016)
1995-2013	United States	Model	Sulfate aerosols	The reductions of sulfate aerosols lead to decreased diffuse light by 0.6% yr^{-1} and GPP by 0.07% yr^{-1}	Keppel-Aleks and Washenfelder (2016)
2010	Amazon	Model	Fire aerosols	Fire aerosols increase GPP by 27%, plant respiration by 10% and decrease soil respiration by 3%	Moreira et al. (2017)
2010 2050	Boreal North America	Model	Fire aerosols	Fire aerosols increase NPP by 8 Tg C yr^{-1} at 2010s and 14 Tg C yr^{-1} at 2050s due to increased diffuse radiation of 2.6 W m^{-2} (1.7%) and 4.0 W m^{-2} (2.3%)	Yue et al. (2017)
2009-2011	China	Model	Aerosols	Aerosols increase NPP by 1.6±0.5% at all-sky and 35±0.9% at clear-sky	Yue and Unger (2017)

2008-2017	Eurasia	Flux.obs	Aerosols	High aerosol loading increases GPP by 6-14% at all sites.	Ezhova et al. (2018)
2000	Global	Model	Biogenic aerosols	Biogenic aerosols enhance global NPP by 1.23 Pg C yr^{-1} due to DFE	Rap et al. (2018)
2001-2011	Global	Model	All and fire aerosols	All (fire) aerosols increase global GPP by 1.0±0.2 (0.05±0.3) Pg C yr^{-1} due to DFE	Yue and Unger (2018)
2014-2015	China	Flux.obs	Aerosols	Photosynthesis of sunlit and shaded leaves increases by 0.56% and 10.71% due to the increase AOD of 0.1	Wang et al. (2018)
2000	Amazon	Model	Fire aerosols	Fire aerosols increase NPP by 5-13 Tg C yr^{-1} due to radiative effects	Malavelle et al. (2019)
2018	Western North America	Flux.obs	Wildfire-smoke aerosols	Aerosols lead to GPP enhancement of 1.2-4.1% compared to the previous growing season	Hemes et al. (2020)
2006-2015	China	Model	Aerosols	Aerosols enhance GPP by 0.36 Pg C yr^{-1} (5%), and DFE makes the dominant contribution (59-62%)	Xie et al. (2020)

672   [a] Carbon metrics include net primary productivity (NPP), net ecosystem productivity (NEP) and

673   gross primary productivity (GPP)

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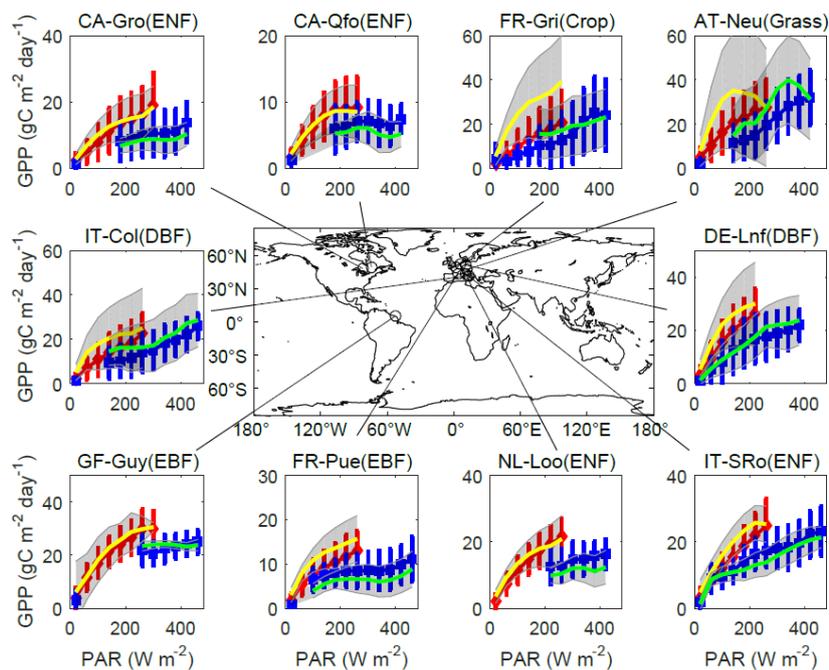
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679 **Figure 1** Simulated and observed GPP responses to direct and diffuse radiation. The comparisons
680 are performed at 10 FLUXNET sites where more than 8 years of observations are available. For
681 each site, hourly observations are divided into direct and diffuse conditions if diffuse fraction is <0.2
682 (blue squares) and >0.8 (red diamonds), respectively. The classified observations are averaged over
683 PAR bins of 40 W m^{-2} with errorbars indicating one standard deviation of GPP for each bin.
684 Similarly, simulations are also divided into direct (green) and diffuse (yellow) bins of PAR with
685 gray shading indicating one standard deviation. The plant function types include evergreen
686 broadleaf forest (EBF), evergreen needleleaf forest (ENF), deciduous broadleaf forest (DBF),
687 grassland (Grass), and cropland (Crop). The site name and vegetation type are listed on the title of
688 each panel.

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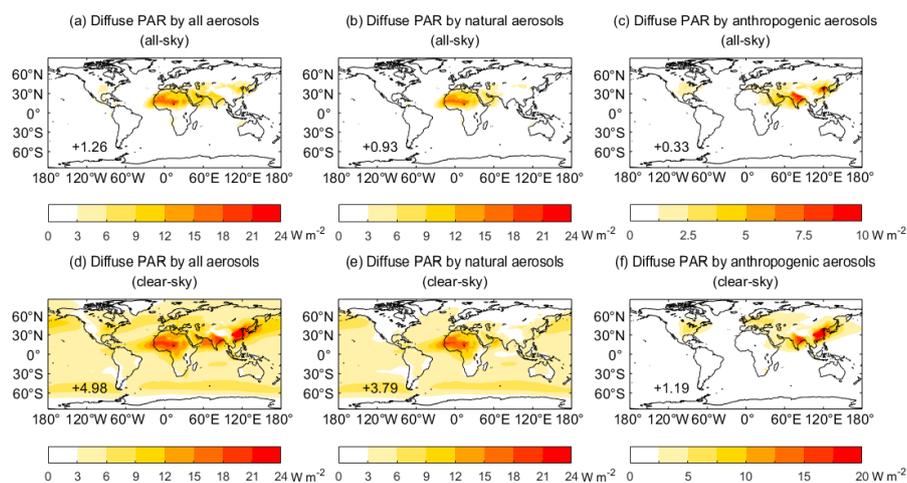
Figure 2 Global changes of diffuse PAR at surface by all, natural, and anthropogenic aerosols at all skies (a, b, c) and clear skies (d, e, f). The aerosol species include natural (BC, OC, dust, sea salt, sulfate, and nitrate) and anthropogenic (BC, OC, sulfate, and nitrate) aerosols. The total changes in PAR caused by different aerosol sources are shown on corresponding panels. Changes of diffuse PAR caused by individual aerosol species can be seen in Fig. S4. The units are W m^{-2}.
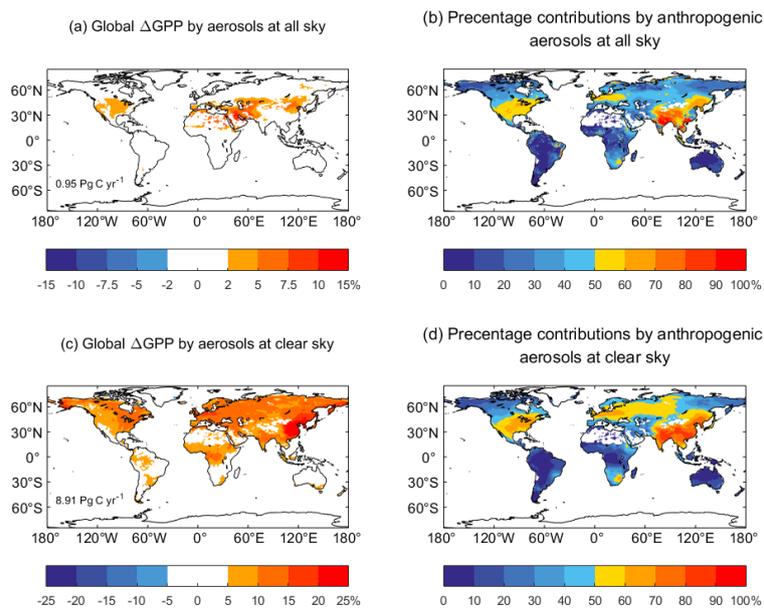
Figure 3 Percentage changes in GPP caused by aerosol diffuse fertilization effect and percentage contributions by anthropogenic aerosols at (a, b) all skies and (c, d) clear skies. The DFE of all aerosols (natural + anthropogenic) are shown on the left, and the contributions by anthropogenic aerosols alone are shown on the right.

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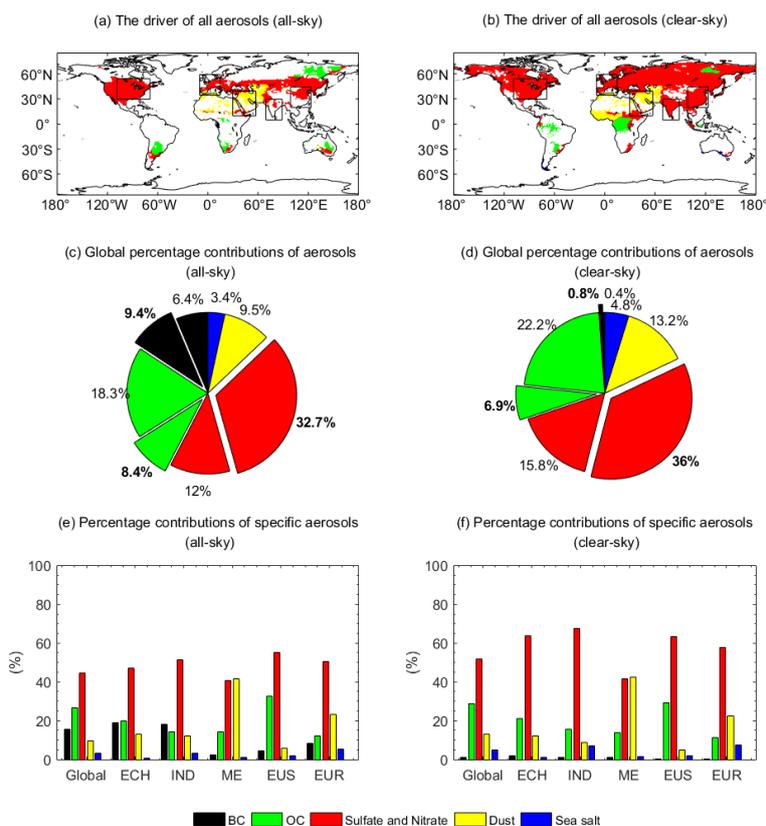
Figure 4 (a, b) Dominant aerosol species contributing to the simulated changes in GPP and the percentage contributions of aerosol species for global (c, d) and specific regions (e, f) at (a, c, e) all skies and (b, d, f) clear skies. The contributions in (c, e) and (d, f) are calculated as the ratios of absolute DFE, as BC aerosols induce negative DFE. The normal (bold) fonts in (c) and (d) represent aerosol species from natural (anthropogenic) sources. Regions with relatively high percentage changes in GPP (>1% for all-sky and >5% for clear-sky) by aerosols are shown in Figure (a) and (b). The regions include eastern China (ECH), India (INA), Middle East (ME), eastern U.S. (EUS), and Europe (EUR), which are marked as black boxes in (a) and (b). The black, green, red, yellow, and blue represent the effects of BC, OC, sulfate and nitrate, dust, and sea salt aerosols, respectively.

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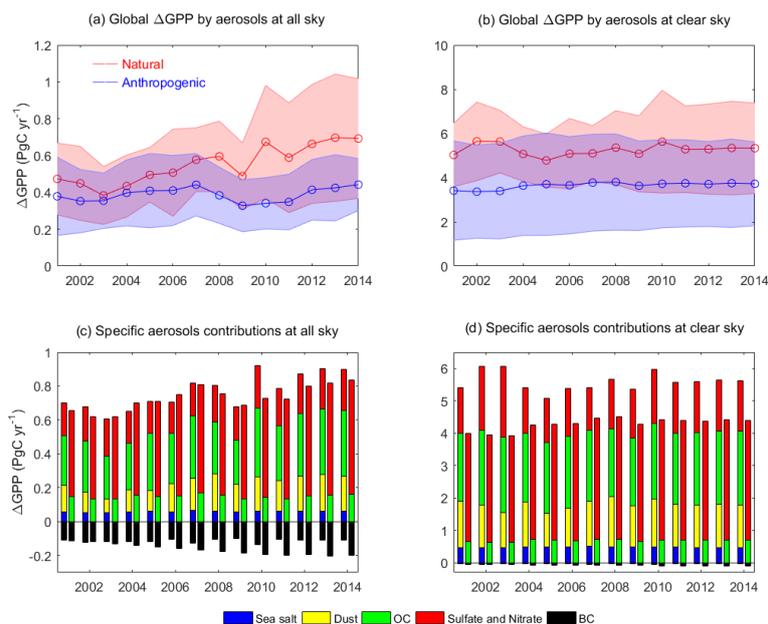
Figure 5 Interannual variations of GPP changes induced by the DFE of natural and anthropogenic aerosols at (a, c) all skies and (b, d) clear skies during 2001-2014. The left and right bars at each year in (c) and (d) represent the effects of natural and anthropogenic aerosol species, respectively. The hollow circles and shadings in (a) and (b) represent annual mean and standard deviation of aerosol-induced GPP changes. The black, green, red, yellow, and blue bars indicate the effects of BC, OC, sulfate and nitrate, dust, and sea salt aerosols, respectively.

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