1	Fire-climate interactions through aerosol radiative effect in a global chemistry-climate-
2	vegetation model
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16 Abstract

17 Fire emissions influence radiation, climate, and ecosystems through aerosol radiative effects. 18 These can drive rapid atmospheric and land surface adjustments which feed back to affect fire 19 emissions. However, the magnitude of such feedback remains unclear on the global scale. Here, we 20 quantify the impacts of fire aerosols on radiative forcing and the fast atmospheric response through 21 direct, indirect, and albedo effects based on the two-way simulations using a well-established 22 chemistry-climate-vegetation model. Globally, fire emissions cause a reduction of $0.565 \pm 0.166W$ 23 m^{-2} in net radiation at the top of the atmosphere with dominant contributions by aerosol indirect 24 effect (AIE). Consequently, terrestrial surface air temperature decreases by 0.061 ± 0.165 °C with 25 coolings of >0.25°C over eastern Amazon, western U.S., and boreal Asia. Both aerosol direct effect 26 (ADE) and AIE contribute to such cooling while the aerosol albedo effect (AAE) exerts an offset 27 warming, especially at high latitudes. Land precipitation decreases by 0.180 ± 0.966 mm month⁻¹ 28 $(1.78 \pm 9.56\%)$ mainly due to the inhibition in central Africa by AIE. Such rainfall deficit further 29 reduces regional leaf area index (LAI) and lightning ignitions, leading to changes in fire emissions. 30 Globally, fire emissions reduce by 2%-3% because of the fire-induced fast responses in humidity, 31 lightning, and LAI. The fire aerosol radiative effects may cause larger perturbations to climate 32 systems with likely more fires under global warming.

33

34 Short summary

We quantify the impacts of fire aerosols on climate through direct, indirect, and albedo effects. In atmosphere-only simulations, we find global fire aerosols cause surface cooling and rainfall inhibition over many land regions. These fast atmospheric perturbations further lead to a reduction in regional leaf area index and lightning activities. By considering the feedback of fire aerosols on humidity, lightning, and leaf area index, we predict a slight reduction in fire emissions.

- 40
- 41 Keywords: Fire emissions; radiative effect; climate feedback; ModelE2-YIBs model
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- 43

44 1 Introduction

45 Fire occurs all year round in both hemispheres, burning about 1% of the Earth's surface and emitting roughly 2-3 Pg (=10¹⁵ g) carbon into atmosphere every year (van der Werf et al., 2017). 46 47 Fire activities are strongly influenced by fuel availability, ignition/suppression, and climate 48 conditions (Flannigan et al., 2009). The fuel type, continuity, and amount affect fire occurrence and 49 spread probability (Flannigan et al., 2013). Lightning discharge is the most important natural source 50 of fire ignition (Macias Fauria and Johnson, 2006). Human activities affect fire patterns by adding 51 ignition sources or by suppressing processes (Andela et al., 2017). Compared to the above factors, 52 climate shows a more dominant role in modulating fire activities through the changes of fuel 53 moisture and spread conditions (Flannigan and Harrington, 1988).

54 Fire exerts prominent impacts on Earth systems and human society through various processes. 55 Biomass burning emits a large amount of trace gases and aerosol particles into the troposphere, 56 affecting air quality at the local and downwind regions (Yue and Unger, 2018). In situ observations 57 showed that about one-third of the background particles in the free troposphere of North America 58 were originated from biomass burning (Hudson et al., 2004). Extremely intense fires can even inject 59 aerosols into stratosphere, where the particles were transported globally (Yu et al., 2019). Fire-60 induced air pollution can reduce global terrestrial productivity of unburned forests (Yue and Unger, 61 2018), leading to weakened carbon uptake by ecosystems. The global transport of fire air pollution 62 also causes large threats to public health by increasing the risks of diseases and mortality (Liu et al., 63 2015). It is estimated that fire-induced particulate matter causes more than 33,000 deaths globally 64 each year (Chen et al., 2021).

65 Aerosols from fires can cause substantial impact on climate via radiative effect owing to their 66 different optical and chemical properties (Xu et al., 2021). Aerosol radiative effect represents the 67 fast atmospheric adjustment or response before changing global mean surface air temperature (TAS). 68 First, aerosols scatter and/or absorb solar radiation through aerosol direct effect (ADE), leading to 69 altered energy budget and climate variables (Carslaw et al., 2010). There is no agreement on the 70 sign of ADE of biomass burning aerosols at the global scale. Some studies (Heald et al., 2014; Veira 71 et al., 2015; Zou et al., 2020) predicted positive forcing while others (Ward et al., 2012; Jiang et al., 72 2016; Grandey et al., 2016) yielded negative forcing (-0.2 to 0.2 W m⁻²), mainly because of the 73 large uncertainties in the absorption of fire-emitted black carbon (BC) (Carslaw et al., 2010; IPCC,

74 2014). Second, aerosols can serve as cloud condensation nuclei (CCN) or ice nuclei to affect the 75 microphysical properties of cloud. Such aerosol indirect effect (AIE) further influences climate 76 system through the changes of cloud albedo and lifetime (Twomey, 1974; Albrecht, 1989). Globally, 77 fire aerosols account for $\sim 30\%$ of the total CCN (Andreae *et al.*, 2004) and the overall negative AIE 78 of fire aerosol is stronger than the ADE in magnitude (Liu et al., 2014; Ward et al., 2012; Jiang et 79 al., 2016). Third, deposition of fire-emitted BC aerosols reduces surface albedo and promotes 80 ice/snow melting, which is called aerosol albedo effect (AAE) (Hansen and Nazarenko, 2004; 81 Warren and Wiscombe, 1980). Compared with other two effects, the AAE shows more regional 82 characteristics (Kang et al., 2020). These fire-induced disturbance in radiative fluxes further alter 83 meteorological and hydrologic variables, which in turn affect fire activities through the changes in 84 fuel moisture and weather conditions.

85 The impacts of fire-induced rapid adjustments on fire activity at the global scale have not been 86 fully assessed. While observations revealed fire-induced perturbations to regional climate (Bali et 87 al., 2017; Zhuravleva et al., 2017), its feedback to fire activities are difficult to be isolated from the 88 influences of background climate. Models provide unique tools to explore fire-climate interactions 89 resulting from aerosol radiative effect especially at the regional to global scales. However, they are 90 not routinely included in most of Earth system models. The IPCC sixth assessment report (AR6) 91 did not provide a quantitative assessment of such feedback as well (IPCC, 2021). In this study, we 92 explore the impacts of fire aerosol radiative effect on climate and the consequent feedbacks to fire 93 emissions by using a well-established fire parameterization coupled to a chemistry-climate-94 vegetation model ModelE2-YIBs (Yue and Unger, 2015). The main objectives are (1) to isolate the 95 radiative effects of fire aerosols through ADE, AIE, and AAE processes and (2) to quantify the 96 feedback of fire-induced rapid adjustments to fire emissions.

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98 2 Data and methods

99 **2.1 Data**

We use the emissions from Global Fire Emission Database version 4.1s (GFED4.1s) to validate the simulated fire emissions. The GFED4.1s provides monthly fire emission fluxes of various air pollutants based on satellite retrieval of area burned from the Moderate Resolution Imaging Spectroradiometer (MODIS) (van der Werf *et al.*, 2017). Area burned in GFED4.1s is mainly 104 derived from the MODIS burned area product (Giglio et al., 2013), taking into account "small" fires 105 outside the burned area maps based on active fire detections (Randerson et al., 2012). The gridded 106 fire emission dataset has a spatial resolution of $0.25^{\circ} \times 0.25^{\circ}$ and is available for every month from 107 July 1997. To compute anthropogenic ignition and suppression effects (see section 2.3), we use a 108 downscaled population density dataset from Gao (2017, 2020). Monthly sea surface temperature 109 (SST) and sea ice concentration (SIC) obtained from Hadley Centre Sea Ice and Sea Surface 110 Temperature (HadISST) dataset (Rayner et al., 2003) are used as the boundary conditions for the 111 climate model.

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113 2.2 ModelE2-YIBs model

The chemistry-climate-vegetation model ModelE2-YIBs is used to simulate the two-way coupling between fire aerosols and climate systems. The ModelE2-YIBs is composed of the NASA Goddard Institute for Space Studies (GISS) ModelE2 model (Schmidt *et al.*, 2014) and the Yale Interactive terrestrial Biosphere Model (YIBs) (Yue and Unger, 2015). The GISS ModelE2 is a global climate-chemistry model with a horizontal resolution of $2^{\circ} \times 2.5^{\circ}$ latitude by longitude and 40 vertical layers extending to the stratosphere (0.1hPa). The dynamics and physics codes are executed every 30 minutes and the radiation code is calculated every 2.5 hours.

121 The gas-phase chemistry scheme considers 156 chemical reactions among 51 species, including NO_x-HO_x-O_x-CO-CH₄ chemistry and different species of volatile organic compounds. 122 123 Aerosol species in ModelE2 include sulfate, nitrate, ammonium, sea salt, dust, BC, and organic 124 carbon (OC), which are interactively calculated and tracked for both mass and number 125 concentrations. Heterogeneous chemistry on dust surfaces and NO_x-dependent secondary organic 126 aerosol production from isoprene and terpenes is included in the model (Bauer et al., 2007b; 127 Tsigaridis and Kanakidou, 2007). The thermodynamic gas-aerosol equilibrium module is used to calculate the phase partitioning of the H₂SO₄/HSO $\frac{-}{4}$ /SO $\frac{-2}{4}$ -HNO₃/NO $\frac{-}{3}$ -NH₃/NH $\frac{+}{4}$ 128 -HCL/CL⁻-Na⁺-Ca²⁺-Mg²⁺-K⁺-H₂O system (Metzger et al., 2006; Bauer et al., 2007a). The 129 130 aerosol microphysical scheme is based on the quadrature method of moments, which incorporates 131 nucleation, gas-particle mass transfer, new particle formation, particle emissions, aerosol phase 132 chemistry, condensational growth, and coagulation (Bauer et al., 2008). The residence time of 133 aerosol species varies greatly in space and time due to different removal rates. Turbulent dry

deposition is determined by resistance-in-series scheme, which is closely coupled to the boundary
layer scheme and implemented between the surface layer (10 m) and the ground (Koch *et al.*, 2006).
The wet deposition consists of several processes including scavenging within and below cloud,
evaporation of falling rainout, transportation along convective plumes, and detrainment and
evaporation from convective plumes (Koch *et al.*, 2006; Shindell *et al.*, 2006).

139 In ModelE2, gases can be converted to aerosols through chemical reactions, while aerosols 140 affect photolysis and provide reaction surface for gases. For example, the formation of sulfate 141 aerosols is driven by modeled oxidants (Bell et al., 2005), and the chemical production of nitrate 142 aerosols is dependent on nitric acid and gaseous ammonia (Bauer et al., 2007b). Moreover, the 143 disturbances of aerosols on climate systems via direct, indirect, and albedo effects are considered in 144 ModelE2. Aerosol optical parameters are calculated by the Mie scattering theory using complex 145 refractive index depending on chemical speciation and particle size. The first AIE is estimated by 146 the prognostic treatment of cloud droplet number concentration, which is a function of species-147 dependent contact nucleation, auto-conversion, and immersion freezing (Menon et al., 2008; Menon 148 et al., 2010). The AAE of BC is considered by estimating the decline of surface albedo as a function 149 of aerosol concentrations at the top layer of snow or ice (Koch and Hansen, 2005). We note that 150 average BC deposition to snow estimated by measurement-based average scavenging ratios is 151 used as a climatological proxy to the physical process of BC deposition (Hansen and Nazarenko, 152 2004). The latter involves size resolved and meteorologically dependent BC deposition fluxes, 153 as would be found in a chemical transport model, but is not used here due to computational 154 constraints. More detailed descriptions of ModelE2 can be found in Schmidt et al. (2014). It has 155 been extensively evaluated for meteorological and chemical variables against observations, 156 reanalysis products and other models, and widely used for studies of climate systems, atmospheric 157 components, and their interactions (Schmidt et al., 2014).

YIBs is a process-based vegetation model that dynamically simulates tree growth and terrestrial carbon fluxes with prescribed fractions of nine plant functional types (PFTs), including deciduous broadleaf forest, evergreen needleleaf forest, evergreen broadleaf forest, tundra, shrubland, C_3/C_4 grassland, and C_3/C_4 cropland. Essential biological processes such as photosynthesis, phenology, autotrophic and heterotrophic respiration are considered and parameterized using the state-of-the-art schemes (Yue and Unger, 2015). Dynamic daily leaf area 164 index (LAI) is estimated based on carbon allocation which is updated every 10 days and prognostic 165 phenology which is dependent instantaneously on temperature and drought conditions. Simulated 166 tree height, phenology, gross primary productivity and LAI agree well with site-level observations 167 and/or satellite retrievals (Yue and Unger, 2015). The YIBs model joined the dynamic global 168 vegetation model inter-comparison project TRENDY and showed reasonable performance of 169 carbon fluxes against available observations (Friedlingstein et al., 2020). In the coupled model, 170 ModelE2 provides meteorological drivers to YIBs, which feeds back to alter land surface water and 171 energy fluxes through changes in stomatal conductance, surface albedo, and LAI. By incorporating 172 YIBs into ModelE2, the new coupled model ModelE2-YIBs can simulate interactions between 173 terrestrial ecosystems and climate systems through the exchange of water and energy fluxes, and 174 chemical components (Yue and Unger, 2015; Yue et al., 2017).

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176 **2.3 Fire parameterization**

We implemented the active global fire parameterization from Pechony and Shindell (2009) into
ModelE2-YIBs model. The parameterization considers key fire-related processes including fuel
flammability, lightning and human ignitions, and human suppressions. Flammability is a unitless
metric indicating conditions favorable for fire occurrence, and is calculated using vapor pressure
deficit (VPD, hPa), precipitation (R, mm day⁻¹), and LAI (m² m⁻²) as follows:

$$Flam = VPD \times LAI \times e^{-C_R \times R}$$
(1)

183 Here, LAI represents vegetation density and is dynamically calculated by YIBs model. c_R is a 184 constant set to 2. VPD is a vital indicator of flammability conditions:

185
$$VPD = e_s \times \left(1 - \frac{RH}{100}\right)$$
(2)

186 where e_s is the saturation vapor pressure and RH is surface relative humidity. e_s can be 187 calculated by Goff-Gratch equation:

 $e_s = e_{st} \times 10^Z \tag{3}$

189 where e_{st} is 1013.246 hPa and

190
$$Z = a \times \left(\frac{T_s}{T} - 1\right) + b \times \log \frac{T_s}{T} + c \times \left(10^{d^{\left(1 - \frac{T_s}{T}\right)}} - 1\right) + f \times \left(10^{h^{\left(\frac{T_s}{T} - 1\right)}} - 1\right)$$
(4)

Here, a, b, c, d, f and h are constants set to -7.90298, 5.02808, -1.3816×10⁻⁷, 11.344, 8.1328×10⁻³ and -3.49149, respectively. T_s is boiling point of water and equal to 373.16 K. VPD and LAI in Eq. (1) are calculated in half-hourly and daily time step, respectively, while 30-day running average precipitation is employed to avoid unrealistically huge flammability fluctuations. It should be noted that the response of flammability to abovementioned factors may not be instantaneous, but may occur over time. For example, a reduction in precipitation in one season at a given location may reduce foliage growth and hence reduce the fuel available for combustion in another season.

198 Natural and anthropogenic ignition rate determines whether the fire can actually occur. If the 199 ignition rate is zero, the resulting fire emissions will be zero, regardless of flammability. The natural 200 ignition rate I_N depends on cloud-to-ground lightning (CoGL) rate, which is simulated by ModelE2 201 following the parameterization of Price and Rind (1994):

202
$$I_{N} = CoGL = \begin{cases} 3.44 \times 10^{-5} \times H^{4.9} \text{ over land} \\ 6.4 \times 10^{-4} \times H^{1.73} \text{ over ocean} \end{cases}$$
(5)

203 where H is the cloud depth (unit: km).

Humans influence fire activity by adding ignition sources and suppressing fire events, the rates of which increase with population and to some extent counteract each other. The anthropogenic ignition rate I_A (number km⁻² month⁻¹) is calculated as follows (Venevsky *et al.*, 2002):

207 $I_A = k(PD) \times PD \times \alpha$ (6)

where PD is population density (number km⁻²). $k(PD) = 6.8 \times PD^{-0.6}$ stands for ignition potentials of human activity, assuming that people in scarcely populated areas interact more with the natural ecosystems and therefore produce more ignition potential. α is the number of potential ignitions per person per month and set to 0.03.

In principle, the successful suppression of fires is dependent on early detection. It is reasonably assumed that fires are detected earlier and suppressed more effectively in highly populated areas. Therefore, the fraction of non-suppressed fires F_{NS} can be expressed as:

215 $F_{NS} = c_1 + c_2 \times exp(-\omega \times PD)$

(7)

where c_1 , c_2 and ω are constants and set to 0.05, 0.95 and 0.05, respectively. The selection of constant values in Eq. (7) is done in a heuristic way, due to lack of quantified data globally. It assumes that up to 95% of fires is suppressed in the densely populated regions but only 5% in unpopulated areas.

220 With the calculation of flammability (Flam), ignition (I_N and I_A), and non-suppression (F_{NS}), 221 the fire count density N_{fire} (unit: number km⁻²) at a specific time step can be derived as:

222	$N_{fire} = Flam \times (I_N + I_A) \times F_{NS}$	(8)
	$N_{\text{fire}} = \Gamma_{\text{fam}} \wedge (\Gamma_{\text{N}} + \Gamma_{\text{A}}) \wedge \Gamma_{\text{NS}}$	(0)

- 223 Finally, fire emissions of trace gases and particulate matters (FireEmis) are calculated as:
- 224

 $FireEmis = N_{fire} \times EF$

(9)

225 Here, EF is the PFT-specific emission factor of an air pollutant such as BC, OC, NO_x, NH₃, SO₂, 226 CO, Alkenes and Paraffin. For each species, simulated gridded emissions are grouped by dominant 227 PFT and compared to annual total emissions from GFED4.1s over the same grids. The EF is then 228 calibrated to minimize the root-mean-square error between the simulated and GFED data for all 229 land grids. Such calibration adjusts only the global total amount of fire emissions without changing 230 the spatiotemporal pattern predicted by the parameterization. The EF is the intrinsic attribution of 231 wildfire emissions that should not vary greatly with climatic conditions. The fire-emitted minerals 232 or dust-like materials are not implemented in the current model, given that these species is not 233 included in the current GFED4.1s.

234 Compared to fire indexes, such as Canadian Fire Weather Index system (Wagner, 1987), this 235 fire parameterization shows advantages in integrating the effects of meteorology, vegetation, natural 236 ignition, and human activities (both ignition and suppression) on fires. Furthermore, it is physically 237 straightforward and has been validated based on global observations (Pechony and Shindell, 2009; 238 Hantson et al., 2020). In ModelE2-YIBs, fire emissions are affected by environmental factors 239 following above parameterizations. In turn, the radiative effects of fire-emitted aerosols feed back 240 to affect those climatic and ecological factors. Note that the changes in the environmental factors 241 may result in changes to fire emissions later. We consider only the fire emissions at surface due 242 to the large uncertainties in depicting fire plume height (Sofiev et al., 2012; Ke et al., 2021). The 243 fire emissions include both primary aerosols and trace gases, the latter of which react with other 244 species to form the secondary aerosols. These particles could be transported across the globe by the 245 three-dimensional atmospheric circulation and eventually removed through either dry or wet 246 deposition.

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248 2.4 Simulations

We perform four groups of sensitivity experiments (Table 1) with the ModelE2-YIBs model to quantify the fire-climate interactions through different radiative processes. The first group with suffix 'AD' considers only the ADE. The second (third) group with suffix 'AD AI' ('AD AA') 252 considers both ADE and AIE (ADE and AAE). The fourth group with suffix 'AD AI AA' includes 253 all three aerosol radiative effects (ADE, AIE, and AAE). Within each group, two runs are performed 254 with (YF) or without (NF) fire emissions. For YF simulations, fire-induced aerosols including 255 primarily emitted and secondarily formed are dynamically calculated based on fire parameterization 256 (see section 2.3) and atmospheric transport. These fire emissions cause radiative perturbations and 257 the consequent fast atmospheric adjustments, which feed back to influence fire emissions. For NF 258 simulations, fire emissions are calculated offline at each step without perturbing the climate system, 259 which can be considered that there is no fire emission. By comparing the climatic variables from 260 the YF and NF runs in the first group, we isolate the impacts of fire aerosols on climate through 261 ADE. By comparing the climatic effects from the first and second (third) groups, we isolate the AIE 262 (AAE) of fire aerosols. By comparing the climatic variables from YF and NF runs in the fourth 263 group, the overall effect (ADE+AIE+AAE) is obtained. Besides, the differences of fire emissions 264 between simulations of "YF AD AI AA" and "NF AD AI AA" represent the feedback of fire 265 aerosol-induced environmental perturbations. Note that fire-emitted gas-phase species also perturb 266 radiation via atmospheric absorption and/or feedback from rapid adjustment; these perturbations are 267 far less than aerosol forcing and could be ignored.

268 For each simulation, climatological mean CO₂ concentrations, SST/SIC, and population 269 density during 1995-2005 are used as boundary conditions to drive the model. Such configuration ignores the year-to-year variability in climate systems, which may cause significant changes in 270 271 annual fire emissions (Burton et al., 2020). Each simulation is integrated for 25 years with the first 272 5 years spinning up and the last 20 years averaged. A two-tailed Student t-test is performed to assess 273 90% confidence levels of the predicted radiative and climatic responses (p < 0.1). The global mean 274 or sum value is depicted in the form of mean value \pm standard deviation. In this study, downward 275 (upward) radiative/heat fluxes are defined as positive (negative). Given that the model is driven by 276 prescribed SST and SIC, only the rapid adjustments of atmospheric variables are taken into account 277 and we mainly focus on climate changes over land grid. The radiative effect simulated with such 278 model configuration is termed the effective radiative forcing (IPCC, 2014).

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280 3 Results

281 3.1 Model evaluation

282 Simulated fire emissions of BC and OC show hotspots in the tropics, such as Amazon, Sahel, 283 central Africa, and Southeast Asia (Fig. S1). The large tropical fire emissions are related to abundant 284 vegetation and/or distinct dry seasons. Compared to GFED4.1s data, ModelE2-YIBs slightly 285 underestimates boreal fire emissions especially over northern Asia and North America. On the global scale, fire releases 1.85 ± 0.01 Tg (1 Tg = 10^{12} g) C year⁻¹ of BC and 16.8 ± 0.92 Tg C year⁻¹ 286 ¹ of OC in ModelE2-YIBs, close to the 1.86 Tg C year-¹ of BC and 16.4 Tg C year-¹ of OC estimated 287 by GFED4.1s. In general, ModelE2-YIBs reasonably captures the spatial distribution of fire 288 289 emissions, with high spatial correlations of 0.67 (p < 0.01) for BC and 0.58 (p < 0.01) for OC, and 290 low normalized mean biases of 0.6% for BC and 2.4% for OC against satellite-based observations.

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3.2 Fire-induced radiative perturbations

Fig. S2 shows the fire-induced changes in Aerosol Optical Depth (AOD) at 550nm. Fire emissions largely enhance surface aerosols especially over tropical regions. Hotspots are located in southern Africa and South America with regional enhancement larger than 0.05. In addition, large enhancement is also found at boreal high latitudes (> 0.01). At the global scale, fires enhance AOD by 0.006 ± 0.001 with 0.010 ± 0.001 over land.

298 Fire aerosols cause large perturbations in net radiation at top of atmosphere (TOA). Globally, the net radiation at TOA decreases 0.565 ± 0.166 W m⁻² by fire aerosols (Fig. 1a). Regionally, 299 300 negative changes are predicted over central Africa, western South America, western North America 301 and the boreal high latitudes. Diagnosis shows that fire-induced AIE dominates the reduction of 302 TOA flux with a global value of -0.440 ± 0.264 W m⁻² (Fig. 1c), accounting for 78% of the total 303 TOA radiative effect by fire aerosols. The spatial correlation coefficient is 0.62 over land grids 304 between the perturbations by all aerosol effects and that by AIE alone. Compared to AIE, the 305 changes in TOA radiative fluxes are much smaller for fire ADE (-0.058 \pm 0.213W m⁻², Fig. 1b) and AAE (-0.016 \pm 0.283 W m⁻², Fig. 1d) with limited perturbations on land. 306

Fire aerosols decrease net shortwave radiation reaching the surface up to 9 W m⁻² in central Africa and 7 W m⁻² in Amazon (Fig. 2a), where biomass burning emissions are most intense (Fig. S1). Such pattern is in general consistent with the changes of TOA fluxes (Fig. 1a), leading to an average reduction of -1.227 ± 0.216 W m⁻² in the shortwave radiation over global land. The fireinduced ADE alone reduces land surface shortwave radiation by 0.654 ± 0.353 W m⁻² with the 312 maximum center in Amazon (Fig. S3a). As a comparison, the fire-induced AIE causes a smaller reduction of -0.553 ± 0.518 W m⁻² with the hotspot in central Africa (Fig. S3c). The net effect of 313 AAE $(0.263 \pm 0.551 \text{ W m}^{-2})$ by fire aerosols is positive mainly because fire AAE reduces surface 314 315 albedo and increase shortwave radiation over Tibetan Plateau and boreal high latitudes (Fig. S3e). 316 However, the magnitude of AAE is much smaller compared to that of ADE and AIE.

317 Changes in surface longwave radiation (Fig. 2b) are much smaller than those in shortwave 318 radiation (Fig. 2a). Regionally, positive changes are predicted in the western U.S., eastern Amazon, 319 and South Africa, where fire-induced surface cooling (Fig. 3a) decreases the upward longwave 320 radiation. On the global scale, fire aerosols cause a decrease of 0.281 ± 0.371 W m⁻² in surface upward longwave radiation. As a result, fire aerosols induce a net atmospheric absorption of 0.191 321 \pm 0.227 W m⁻² over land grids (Fig. 2c). The reductions in surface shortwave radiation are largely 322 323 balanced by changes in heat fluxes at the surface, which shows an average decrease of 0.826 ± 0.311 324 W m⁻² in the upward fluxes over land grids (Fig. 2d). Fire ADE and AIE lead to reductions of 0.503 \pm 0.289 W m⁻² and 0.432 \pm 0.411W m⁻² in surface upward heat fluxes, respectively (Fig. S3b and 325 S3d). Changes in sensible heat account for 82.2 % of the changes in total heat reduction, much 326 327 higher than the contributions of 17.8% by latent heat fluxes (Fig. S4). Regionally, the upward 328 sensible heat decreases in the western U.S. and Amazon mainly due to fire ADE, while the upward 329 latent heat decreases in central Africa mainly by fire AIE (Fig. S5).

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3.3 Fire-induced fast climatic responses

332 In response to the perturbations in radiative fluxes, land TAS decreases 0.061 ± 0.165 °C 333 globally by fire aerosols (Fig. 3a). Such cooling is mainly located in western U.S., Amazon, and 334 boreal Asia, following the large reductions in shortwave radiation (Fig. 2a). Meanwhile, moderate 335 warming is predicted at the high latitudes of both hemispheres especially over the areas covered 336 with land ice such as Greenland and Antarctica. Sensitivity experiments show that both ADE (Fig. 337 4a) and AIE (Fig. 4c) of fire aerosols result in net cooling globally, with regional reductions of TAS 338 over boreal Asia and North America. In contrast, the fire AAE causes increases of TAS over boreal 339 Asia and North America (Fig. 4e), where the deposition of BC aerosols reduces surface albedo. 340 Consequently, the fire AAE results in a global warming of 0.054 ± 0.163 °C, which in part offsets 341 the cooling effects by the ADE and AIE of fire aerosols.

342 Meanwhile, global land precipitation decreases by 0.180 ± 0.966 mm/month ($1.78 \pm 9.56\%$) 343 with great spatial heterogeneity (Fig. 3b). Decreased precipitation is predicted over central Africa, 344 boreal North America, and eastern Siberia. In contrast, increased rainfall is predicted in western 345 U.S., eastern Amazon, and northern Asia. The reduction of precipitation is mainly contributed by 346 fire AIE, which reduces cloud droplet size and inhibits local rainfall in central Africa (Fig. 4d). 347 Consequently, latent heat fluxes are reduced to compensate the rainfall deficit in central Africa (Fig. 348 S4b).

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3.4 Fast response feedback on fire emissions

351 The fire-aerosol-induced fast response in precipitation, VPD, lightning, and LAI can feed back 352 to affect fire emissions. However, these changes may have contrasting impacts on fire activities. For 353 example, the aerosol-induced reduction of precipitation in central Africa (Fig. 3b) increases local 354 VPD (Fig. 5a) and consequently causes more fire emissions. Meanwhile, such enhanced drought 355 condition inhibits plant growth and decreases local LAI (Fig. 5c), which has negative impacts on 356 fire emissions by reducing fuel density. Furthermore, the fire AIE inhibits the development of 357 convective cloud, which limits cloud height and the number of cloud-to-ground lightning in central 358 Africa (Fig. 5b), leading to reduced ignition sources and fire emissions.

359 To illustrate the joint the impacts of fire-aerosol-induced fast climate responses, we count the 360 number out of the four factors contributing positive effects to fire emissions over land grids (Fig. 361 5d). The larger (smaller) number indicates higher possibility of increasing (decreasing) fire 362 emissions. Most of areas show neutral number of 2, indicating offsetting effects of the changes in 363 fire-prone factors. Only 13.5 % of land grids show numbers higher than 2 with sparse distribution. 364 In contrast, 32.1 % of land grids show numbers smaller than 2, especially for the grids over Siberia 365 and western U.S. where the increased rainfall (Fig. 3b) and decreased VPD (Fig. 5a) inhibit fire 366 emissions. Furthermore, the regional reductions in lightning ignition or LAI promote the inhibition 367 effects. As a result, fire emissions in YF AD AI AA slightly decrease by 31.0 ± 35.9 Gg year⁻¹ (1.7%) for BC and 493.6 \pm 566.8 Gg year⁻¹ (2.9%) for OC compared to NF AD AI AA in which 368 369 fire emissions do not perturb climate (Fig. 6).

370

371 4 Conclusions and discussion

372 We used the chemistry-climate-vegetation coupled model ModelE2-YIBs to quantify fire-373 climate interactions through ADE, AIE, and AAE. Globally, fire aerosols decrease TOA net 374 radiation by 0.565 ± 0.166 W m⁻², dominated by the AIE over central Africa. Surface net solar 375 radiation also exhibits widespread reductions especially over fire-prone areas with compensations 376 from the decreased sensible and latent heat fluxes. Following the changes in radiation, land TAS decreases by 0.061 ± 0.165 °C and precipitation decreases by 0.180 ± 0.966 mm/month, albeit with 377 378 regional inconsistencies. The surface cooling is dominated by fire ADE and AIE, while the drought 379 tendency is mainly contributed by fire AIE with hotspots in central Africa. AAE also plays an 380 important role by introducing warming tendency at the mid-to-high latitudes. These fire-induced 381 fast climatic responses further affect VPD, LAI, and lightning ignitions, leading to reductions in 382 global fire emissions of BC by 2% and OC by 3%. It may seem counter-intuitive that reduced 383 precipitation would decrease wildfire emissions, while the observation-based data show that the 384 fire-precipitation correlations are not negative in all regions (Fig. S6). In this study, the inhibition 385 of precipitation in central Africa (Fig. 3b) reduces regional LAI (Fig. 5c) and decreases fuel 386 availability for fire occurrence, resulting in a positive correlation between fire and precipitation that 387 matches the observed relationship in Africa (Fig. S6). However, in North America, Eurasia, and the 388 Amazon Basin, precipitation is anti-correlated with fire emissions. These differences may reflect 389 the seasonal variation of rainfall in the different regions.

Our predicted reduction of 0.565 ± 0.166 W m⁻² in TOA radiation by fire aerosols is close to 390 391 the estimate of -0.51 W m⁻² reported by Jiang et al. (2016) and -0.59 W m⁻² of Zou et al. (2020) 392 using different models with prescribed SST/SIC and fire-induced ADE, AIE and AAE (Table 2). 393 Within such change, fire ADE alone makes a moderate contribution of -0.016 ± 0.283 W m⁻², falling 394 within the range of -0.2 to 0.2 W m⁻² from other studies. The large uncertainty of fire ADE is likely 395 related to the discrepancies in the BC absorption among climate models, which cause varied net 396 effects when offsetting the radiative perturbations of scattering aerosols. As a comparison, fire AIE 397 in our model induces a significant radiative effect of -0.440 ± 0.264 W m⁻². However, such magnitude is much smaller than previous estimates of -0.7 to -1.1 W m⁻² using different models 398 399 (Table 2). We further estimated a limited fire AAE of -0.016 ± 0.283 W m⁻², consistent with previous 400 findings showing insignificant role of AAE by fire aerosols (Ward et al., 2012; Jiang et al., 2016). 401 Our estimates of reductions in TAS and precipitation also fall within the range of previous studies

402 (Table 2).

403 Our estimates are subject to some limitations and uncertainties. First, we considered only the 404 fast climatic responses of land surface with prescribed SST and SIC in the simulations. Although 405 most of fire-induced AOD changes are located on land (Fig. S2), the air-sea interaction may cause 406 complex climatic responses to aerosol radiative effects. In a recent study, Jiang et al. (2020) 407 emphasized the role of slow feedback contributed by fire aerosols on global precipitation reduction 408 by using a coupled model. Such air-sea interaction will modify the magnitude and/or spatial pattern 409 of fast climatic responses revealed in this study, and should be explored in the future studies with 410 coupled ocean models. Second, the nonlinear effects of different radiative processes may influence 411 the attribution results. In this study, we isolate the effects of AIE and AAE by subtracting variables 412 between different groups following the approaches by Bauer and Menon (2012). However, the 413 additive perturbations from individual processes are not equal to the total perturbations with all 414 processes in one simulation. For example, the sum of three processes causes changes of TOA 415 radiation by -0.513 ± 0.324 W m⁻² (Figs 1b-1d), surface temperature by -0.037 ± 0.160 °C (Figs 4a, 416 4c, 4e), and precipitation by -1.090 ± 1.122 mm month⁻¹ (Figs 4b, 4d, 4f). These perturbations are 417 weaker than the net effects of 0.565 ± 0.166 W m⁻² (Fig. 1a) in radiation and -0.061 ± 0.165 °C in 418 temperature (Fig. 3a), but much stronger than that of -0.18 ± 0.96 mm month⁻¹ in precipitation (Fig. 419 3b) predicted by the simulation with all three processes. As a result, the nonlinear feedbacks among 420 different radiative processes may magnify or offset the final climatic responses to fire aerosols. 421 Third, considering the complex nature of fire activities, the fire parameterization in this study does 422 not incorporate all fire-related processes (e.g., the influence of wind). In addition, the simulations 423 omit several factors influencing fire emissions (e.g., moist content of fuels) and aerosol radiative 424 effects (e.g. fire plume height). For example, studies show significant impacts of plume rise on the 425 vertical distribution of fire aerosols and the consequent radiative effects (Walter et al., 2016). The 426 impacts of human activity on fire emissions are calculated as a function of population density 427 without considerations of differences in economy, education, and policies. These auxiliary factors 428 may increase the spatial heterogeneity of fire aerosol radiative effects and deserve further 429 explorations in the future studies.

430 Despite these limitations, we made the first attempt to assess the two-way interaction between431 fire emissions and climate via aerosol radiative effects. Our results show that fire-emitted aerosols

432 cause negative ERF of 0.565 ± 0.166 W m⁻², which is about 20% of the anthropogenic ERF due to 433 the increased greenhouse gases and aerosols from 1950 to 2019 (IPCC, 2021). Such fire ERF largely 434 reduces regional TAS and precipitation, leading to further changes in fire emissions. Although the 435 reduction of 2% to 3% in fire emissions by the fire-climate interaction through aerosol radiative 436 effect seems limited, such change is a result of several complex feedbacks that may exert offsetting 437 effects, and the relative magnitude of individual factors may vary spatially. Both the number of 438 factors and the magnitude of their effects will determine the overall response. Furthermore, our 439 simulations reveal a strong inhibition effect of fire aerosols on LAI in central Africa due to the 440 aerosol-induced drought intensification. Such negative effects on ecosystems are inconsistent with 441 previous estimates that showed certain fertilization effects by fire aerosols (Yue and Unger, 2018), 442 mainly because the rainfall deficit overweighs the diffuse fertilization effects of aerosols. With likely 443 more fires under global warming (Abatzoglou et al., 2019), our results suggested complex and 444 uncertain perturbations by fire emissions to climate and ecosystem through fire-climate interactions.

445

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449

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- 453

454 **Competing Interests**

- 455 The authors declare that they have no conflict of interest.
- 456

457 Data availability

Hadley Centre Sea Ice and Sea Surface Temperature dataset were obtain from
<u>https://www.metoffice.gov.uk/hadobs/hadisst/</u>. Population data could be downloaded form
<u>https://cmr.earthdata.nasa.gov/search/concepts/C1739468823-SEDAC.html</u>. GFED data were
obtained from https://daac.ornl.gov/VEGETATION/guides/fire emissions v4 R1.html. Model

462 data from this study are available from the corresponding author upon request.

463 **Reference:**

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650

Fig. 1 Changes in net radiation flux at top of atmosphere due to (a) total effects, (b) ADE, (c) AIE, and (d) AAE of fire aerosols. Positive values represent the increase of downward radiation. Global average value is shown at the top of each panel. Slashes denote areas with significant (p < 0.1)

- 654 changes.
- 655





Fig. 2 Changes in (a) surface net shortwave radiation, (b) surface net longwave radiation, (c) atmospheric absorbed radiation, and (d) surface heat flux (sensible + latent) over land grids caused by fire aerosols. Positive values represent the increase of downward radiation/heat for (a, b and d) and absorption for (c). Global land average value is shown at the top of each panel. Slashes denote areas with significant (p < 0.1) changes.



Fig. 3 Changes in (a) surface air temperature and (b) precipitation over land grids caused by fire aerosols. The zonal averages of these changes are shown by the side of each panel. Global land average value is shown at the top of each panel. Slashes denote areas with significant (p < 0.1) changes.



669

670 Fig. 4 Changes in (a, c, e) surface air temperature and (b, d, f) precipitation over land grids due to

- 671 (a, b) ADE, (c, d) AIE, and (e, f) AAE of fire aerosols. Global land average value is shown at the 672 top of each panel. Slashes denote areas with significant (p < 0.1) changes.
- 673





Fig. 5 Changes in (a) vapor pressure deficit (VPD), (b) lightning ignition, and (c) leaf area index (LAI) over land grids induced by fire aerosols. Global land average value is shown at the top of each panel. Slashes denote areas with significant (p < 0.1) changes. The number of factors whose changes induced by fire aerosols cause positive feedback to fire emissions is shown in (d). Only grids with fire-emitted OC larger than 1×10^{-12} kg s⁻¹ m⁻² (colored domain in Fig. S1b) are shown in 680 (d).



682

683 **Fig. 6** Changes in fire emissions of (a) BC and (b) OC due to the fast response feedback. The changes 684 of fire emissions are calculated as the differences between YF_AD_AI_AA and NF_AD_AI_AA 685 with slashes indicating significant (p < 0.1) changes. The total emission is shown at the top of each 686 panel.

687

 Table 1. Summary of simulations using ModelE2-YIBs

	Simulation	Fires ^a	Aerosol direct effect	Aerosol indirect effect	Aerosol albedo effect
	NF_AD	No	Yes	No	No
	YF_AD	Yes	Yes	No	No
	NF_AD_AI	No	Yes	Yes	No
	YF_AD_AI	Yes	Yes	Yes	No
	NF_AD_AA	No	Yes	No	Yes
	YF_AD_AA	Yes	Yes	No	Yes
	NF_AD_AI_AA	No	Yes	Yes	Yes
_	YF_AD_AI_AA	Yes	Yes	Yes	Yes

^a All simulations predict fire emissions but the runs with NF do not feed the fire aerosols into the

690 model to perturb radiative fluxes.

693	surface climate with previous studies						
	Deference	RF	ADE	AIE	AAE	TAS	Pr
	Reference	(W m ⁻²)	(°C)	(mm month ⁻¹)			
	Ward <i>et al.</i> (2012) ^a	-0.55	0.10	-1.00	0.00		
	Heald et al. (2014)	—	-0.19				
	Veira et al. (2015)		-0.20		—		
	Grandey et al. (2016)	-1.0	0.04	-1.11	-0.1		-0.018
	Jiang et al. (2016)	-0.51	0.16	-0.70	0.03	-0.03	-0.3
	Zou et al. (2020)	-0.59	-0.003	-0.82	0.19		
	Xu et al. (2021)	-0.73	0.25	-0.98	_	-0.17	-1.2
	Yan et al. (2021)	-0.62	0.17	-0.74	-0.04	0.03	
	This study	-0.565	-0.058	-0.440	-0.016	-0.061	-0.180

Table 2. Comparison of the simulated fire-induced change in radiative forcings at TOA and

 surface climate with previous studies

^a other effects of fire-induced on radiative turbulances are considered in this paper